



Version 14

Discovering JMP

*"The real voyage of discovery consists not in seeking new
landscapes, but in having new eyes."*

Marcel Proust

JMP, A Business Unit of SAS
SAS Campus Drive
Cary, NC 27513

14.1

The correct bibliographic citation for this manual is as follows: SAS Institute Inc. 2018.
Discovering JMP® 14. Cary, NC: SAS Institute Inc.

Discovering JMP® 14

Copyright © 2018, SAS Institute Inc., Cary, NC, USA

ISBN 978-1-63526-489-0 (Hardcopy)

ISBN 978-1-63526-490-6 (EPUB)

ISBN 978-1-63526-491-3 (MOBI)

ISBN 978-1-63526-492-0 (Web PDF)

All rights reserved. Produced in the United States of America.

For a hard-copy book: No part of this publication may be reproduced, stored in a retrieval system, or transmitted, in any form or by any means, electronic, mechanical, photocopying, or otherwise, without the prior written permission of the publisher, SAS Institute Inc.

For a web download or e-book: Your use of this publication shall be governed by the terms established by the vendor at the time you acquire this publication.

The scanning, uploading, and distribution of this book via the Internet or any other means without the permission of the publisher is illegal and punishable by law. Please purchase only authorized electronic editions and do not participate in or encourage electronic piracy of copyrighted materials. Your support of others' rights is appreciated.

U.S. Government License Rights; Restricted Rights: The Software and its documentation is commercial computer software developed at private expense and is provided with RESTRICTED RIGHTS to the United States Government. Use, duplication or disclosure of the Software by the United States Government is subject to the license terms of this Agreement pursuant to, as applicable, FAR 12.212, DFAR 227.7202-1(a), DFAR 227.7202-3(a) and DFAR 227.7202-4 and, to the extent required under U.S. federal law, the minimum restricted rights as set out in FAR 52.227-19 (DEC 2007). If FAR 52.227-19 is applicable, this provision serves as notice under clause (c) thereof and no other notice is required to be affixed to the Software or documentation. The Government's rights in Software and documentation shall be only those set forth in this Agreement.

SAS Institute Inc., SAS Campus Drive, Cary, North Carolina 27513-2414.

March 2018

August 2018

SAS® and all other SAS Institute Inc. product or service names are registered trademarks or trademarks of SAS Institute Inc. in the USA and other countries. ® indicates USA registration.

Other brand and product names are trademarks of their respective companies.

SAS software may be provided with certain third-party software, including but not limited to open-source software, which is licensed under its applicable third-party software license agreement. For license information about third-party software distributed with SAS software, refer to <http://support.sas.com/thirdpartylicenses>.

Technology License Notices

- Scintilla - Copyright © 1998-2017 by Neil Hodgson <neilh@scintilla.org>.

All Rights Reserved.

Permission to use, copy, modify, and distribute this software and its documentation for any purpose and without fee is hereby granted, provided that the above copyright notice appear in all copies and that both that copyright notice and this permission notice appear in supporting documentation.

NEIL HODGSON DISCLAIMS ALL WARRANTIES WITH REGARD TO THIS SOFTWARE, INCLUDING ALL IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS, IN NO EVENT SHALL NEIL HODGSON BE LIABLE FOR ANY SPECIAL, INDIRECT OR CONSEQUENTIAL DAMAGES OR ANY DAMAGES WHATSOEVER RESULTING FROM LOSS OF USE, DATA OR PROFITS, WHETHER IN AN ACTION OF CONTRACT, NEGLIGENCE OR OTHER TORTIOUS ACTION, ARISING OUT OF OR IN CONNECTION WITH THE USE OR PERFORMANCE OF THIS SOFTWARE.

- Telerik RadControls: Copyright © 2002-2012, Telerik. Usage of the included Telerik RadControls outside of JMP is not permitted.
- ZLIB Compression Library - Copyright © 1995-2005, Jean-Loup Gailly and Mark Adler.
- Made with Natural Earth. Free vector and raster map data @ naturalearthdata.com.
- Packages - Copyright © 2009-2010, Stéphane Sudre (s.sudre.free.fr). All rights reserved.

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.

Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer in the documentation and/or other materials provided with the distribution.

Neither the name of the WhiteBox nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS "AS IS" AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT OWNER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

- iODBC software - Copyright © 1995-2006, OpenLink Software Inc and Ke Jin (www.iodbc.org). All rights reserved.

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

- Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.
- Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer in the documentation and/or other materials provided with the distribution.
- Neither the name of OpenLink Software Inc. nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS "AS IS" AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL OPENLINK OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

- bzip2, the associated library “libbzip2”, and all documentation, are Copyright © 1996-2010, Julian R Seward. All rights reserved.

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.

The origin of this software must not be misrepresented; you must not claim that you wrote the original software. If you use this software in a product, an acknowledgment in the product documentation would be appreciated but is not required.

Altered source versions must be plainly marked as such, and must not be misrepresented as being the original software.

The name of the author may not be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE AUTHOR “AS IS” AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE AUTHOR BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

- R software is Copyright © 1999-2012, R Foundation for Statistical Computing.
- MATLAB software is Copyright © 1984-2012, The MathWorks, Inc. Protected by U.S. and international patents. See www.mathworks.com/patents. MATLAB and Simulink are registered trademarks of The MathWorks, Inc. See www.mathworks.com/trademarks for a list of additional trademarks. Other product or brand names may be trademarks or registered trademarks of their respective holders.
- libopc is Copyright © 2011, Florian Reuter. All rights reserved.

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

- Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.

- Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer in the documentation and / or other materials provided with the distribution.
- Neither the name of Florian Reuter nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS “AS IS” AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT OWNER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

- libxml2 - Except where otherwise noted in the source code (e.g. the files hash.c, list.c and the trio files, which are covered by a similar license but with different Copyright notices) all the files are:

Copyright © 1998 - 2003 Daniel Veillard. All Rights Reserved.

Permission is hereby granted, free of charge, to any person obtaining a copy of this software and associated documentation files (the “Software”), to deal in the Software without restriction, including without limitation the rights to use, copy, modify, merge, publish, distribute, sublicense, and/or sell copies of the Software, and to permit persons to whom the Software is furnished to do so, subject to the following conditions:

The above copyright notice and this permission notice shall be included in all copies or substantial portions of the Software.

THE SOFTWARE IS PROVIDED “AS IS”, WITHOUT WARRANTY OF ANY KIND, EXPRESS OR IMPLIED, INCLUDING BUT NOT LIMITED TO THE WARRANTIES OF MERCHANTABILITY, FITNESS FOR A PARTICULAR PURPOSE AND NONINFRINGEMENT. IN NO EVENT SHALL DANIEL VEILLARD BE LIABLE FOR ANY CLAIM, DAMAGES OR OTHER LIABILITY, WHETHER IN AN ACTION OF CONTRACT, TORT OR OTHERWISE, ARISING FROM, OUT OF OR IN CONNECTION WITH THE SOFTWARE OR THE USE OR OTHER DEALINGS IN THE SOFTWARE.

Except as contained in this notice, the name of Daniel Veillard shall not be used in advertising or otherwise to promote the sale, use or other dealings in this Software without prior written authorization from him.

- Regarding the decompression algorithm used for UNIX files:

Copyright © 1985, 1986, 1992, 1993

The Regents of the University of California. All rights reserved.

THIS SOFTWARE IS PROVIDED BY THE REGENTS AND CONTRIBUTORS "AS IS" AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE REGENTS OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

1. Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.

2. Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer in the documentation and/or other materials provided with the distribution.

3. Neither the name of the University nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

- Snowball - Copyright © 2001, Dr Martin Porter, Copyright © 2002, Richard Boulton.

All rights reserved.

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

1. Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.

2. Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer in the documentation and/or other materials provided with the distribution.

3. Neither the name of the copyright holder nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS \“AS IS\” AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT HOLDER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

- Pako - Copyright © 2014–2017 by Vitaly Puzrin and Andrei Tuputcyn.

Permission is hereby granted, free of charge, to any person obtaining a copy of this software and associated documentation files (the “Software”), to deal in the Software without restriction, including without limitation the rights to use, copy, modify, merge, publish, distribute, sublicense, and/or sell copies of the Software, and to permit persons to whom the Software is furnished to do so, subject to the following conditions:

The above copyright notice and this permission notice shall be included in all copies or substantial portions of the Software.

THE SOFTWARE IS PROVIDED “AS IS”, WITHOUT WARRANTY OF ANY KIND, EXPRESS OR IMPLIED, INCLUDING BUT NOT LIMITED TO THE WARRANTIES OF MERCHANTABILITY, FITNESS FOR A PARTICULAR PURPOSE AND NONINFRINGEMENT. IN NO EVENT SHALL THE AUTHORS OR COPYRIGHT HOLDERS BE LIABLE FOR ANY CLAIM, DAMAGES OR OTHER LIABILITY, WHETHER IN AN ACTION OF CONTRACT, TORT OR OTHERWISE, ARISING FROM, OUT OF OR IN CONNECTION WITH THE SOFTWARE OR THE USE OR OTHER DEALINGS IN THE SOFTWARE.

Get the Most from JMP[®]

Whether you are a first-time or a long-time user, there is always something to learn about JMP.

Visit JMP.com to find the following:

- live and recorded webcasts about how to get started with JMP
- video demos and webcasts of new features and advanced techniques
- details on registering for JMP training
- schedules for seminars being held in your area
- success stories showing how others use JMP
- a blog with tips, tricks, and stories from JMP staff
- a forum to discuss JMP with other users

<http://www.jmp.com/getstarted/>

Contents

Discovering JMP

	About This Book	15
	Gallery of JMP Graphs	17
1	Learn about JMP	37
	Documentation and Additional Resources	
	Formatting Conventions	39
	JMP Documentation	40
	JMP Documentation Library	40
	JMP Help	46
	Additional Resources for Learning JMP	46
	Tutorials	46
	Sample Data Tables	47
	Learn about Statistical and JSL Terms	47
	Learn JMP Tips and Tricks	47
	Tooltips	47
	JMP User Community	48
	JMP Books by Users	48
	The JMP Starter Window	48
	Technical Support	49
2	Introducing JMP	51
	Basic Concepts	
	Concepts That You Should Know	53
	How Do I Get Started?	53
	Starting JMP	54
	Using Sample Data	56
	Understanding Data Tables	56
	Understanding the JMP Workflow	58
	Step 1: Launching a Platform and Viewing Results	59
	Step 2: Removing the Box Plot	61
	Step 3: Requesting Additional Output	61
	Step 4: Interacting with Platform Results	62
	How is JMP Different from Excel?	63

3	Work with Your Data	65
	Prepare Your Data for Graphing and Analyzing	
	Get Your Data into JMP	67
	Copy and Paste Data	67
	Import Data	67
	Enter Data	70
	Transfer Data from Excel	72
	Work with Data Tables	73
	Edit Data	74
	Select, Deselect, and Find Values	75
	View or Change Column Information	79
	Calculate Values with Formulas	81
	Filter Data	83
	Manage Data	84
	View Summary Statistics	85
	Create Subsets	89
	Join Data Tables	91
	Sort Tables	93
4	Visualize Your Data	95
	Common Graphs	
	Analyze Single Variables	97
	Histograms	97
	Bar Charts	99
	Compare Multiple Variables	102
	Scatterplots	102
	Scatterplot Matrix	106
	Side-by-Side Box Plots	109
	Overlay Plots	112
	Variability Chart	115
	Graph Builder	118
	Bubble Plots	124
5	Analyze Your Data	131
	Distributions, Relationships, and Models	
	About This Chapter	133
	The Importance of Graphing Your Data	133
	Understand Modeling Types	136
	Example of Viewing Modeling Type Results	136
	Change the Modeling Type	137
	Analyze Distributions	139

	Distributions of Continuous Variables	140
	Distributions of Categorical Variables	142
	Analyze Relationships	145
	Use Regression with One Predictor	146
	Compare Averages for One Variable	151
	Compare Proportions	154
	Compare Averages for Multiple Variables	156
	Use Regression with Multiple Predictors	160
6	The Big Picture	167
	Exploring Data in Multiple Platforms	
	Fun Fact: Linked Analyses	169
	Explore Data in Multiple Platforms	169
	Analyze Distributions	169
	Analyze Patterns and Relationships	173
	Analyze Similar Values	177
7	Save and Share Your Work	183
	Save and Re-create Results	
	Save Platform Results in Journals	185
	Example of Creating a Journal	185
	Add Analyses to a Journal	186
	Create a Project	186
	Save Files in a Project	190
	Move Files to a Project	192
	Share the Project	194
	Save and Run Scripts	194
	Example of Saving and Running a Script	195
	About Scripts and JSL	196
	Save Reports as Interactive HTML	196
	Interactive HTML Contains Data	197
	Example of Creating Interactive HTML	198
	Create a Web Report	199
	Save a Report as a PowerPoint Presentation	201
	Create Adobe Flash Versions of the Profiler, Bubble Plot, or Distribution Platforms	202
	Example of Saving an Adobe Flash Version of a Bubble Plot	202
	Create Dashboards	204
	Example of Combining Windows	205
	Example of Creating a Dashboard with Two Reports	206

8 Special Features 209

Automatic Analysis Updates and SAS Integration

 Automatically Update Analyses and Graphs 211

 Example of Using Automatic Recalc 211

 Change Preferences 214

 Example of Changing Preferences 215

 Integrate JMP and SAS 217

 Example of Creating SAS Code 218

 Example of Submitting SAS Code 218

Index 221

Discovering JMP

About This Book

Discovering JMP provides a general introduction to the JMP software. This guide assumes that you have no knowledge of JMP. Whether you are an analyst, researcher, student, professor, or statistician, this guide gives you a general overview of JMP's user interface and features.

This guide introduces you to the following information:

- Starting JMP
- The structure of a JMP window
- Preparing and manipulating data
- Using interactive graphs to learn from your data
- Performing simple analyses to augment graphs
- Customizing JMP and special features
- Sharing your results

This guide contains six chapters. Each chapter contains examples that reinforce the concepts presented in the chapter. All of the statistical concepts are at an introductory level. The sample data used in this book are included with the software. Here is a description of each chapter:

- [Chapter 2, “Introducing JMP”](#) provides an overview of the JMP application. You learn how content is organized and how to navigate the software.
- [Chapter 3, “Work with Your Data”](#) describes how to import data from a variety of sources, and prepare it for analysis. There is also an overview of data manipulation tools.
- [Chapter 4, “Visualize Your Data”](#) describes graphs and charts that you can use to visualize and understand your data. The examples range from simple analyses involving a single variable, to multiple-variable graphs that enable you to see relationships between many variables.
- [Chapter 5, “Analyze Your Data”](#) describes many commonly used analysis techniques. These techniques range from simple techniques that do not require the use of statistical methods, to advanced techniques, where knowledge of statistics is useful.
- [Chapter 6, “The Big Picture”](#) shows you how to analyze distributions, patterns, and similar values in several platforms.
- [Chapter 7, “Save and Share Your Work”](#) describes sharing your work with non JMP users in PowerPoint presentations, interactive HTML, and Adobe Flash. Saving analyses as scripts and saving work in journals and projects for JMP users are also covered.

- [Chapter 8, “Special Features”](#) describes how to automatically update graphs and analyses as data changes, how to use preferences to customize your reports, and how JMP interacts with SAS.

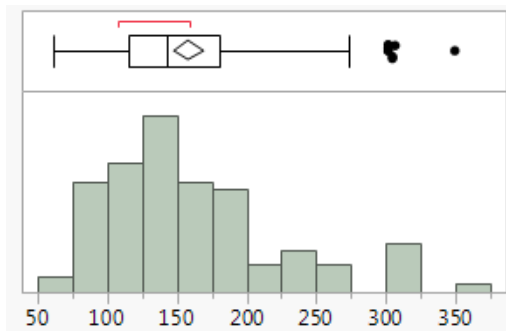
After reviewing this guide, you will be comfortable navigating and working with your data in JMP.

JMP is available for both Windows and Macintosh operating systems. However, the material in this guide is based on a Windows operating system.

Gallery of JMP Graphs

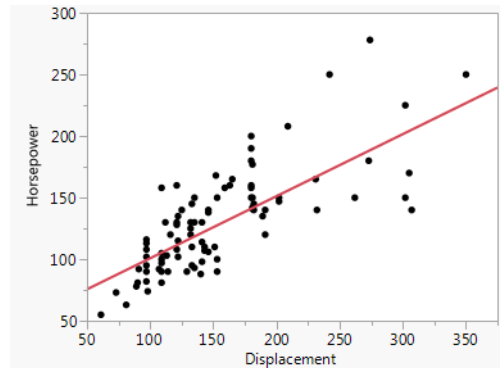
Various Graphs and Their Platforms

Here are pictures of many of the graphs that you can create with JMP. Each picture is labeled with the platform used to create it. For more information about the platforms and these and other graphs, see the documentation on the **Help > Books** menu.



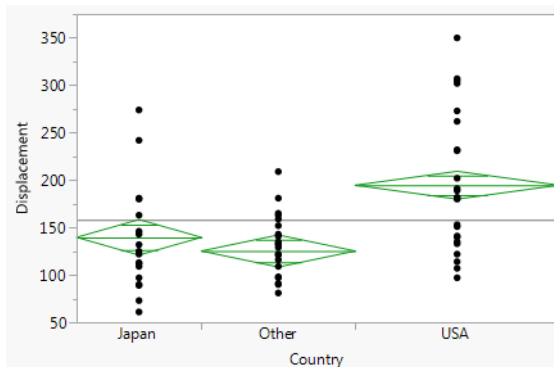
Histogram

Analyze > Distribution



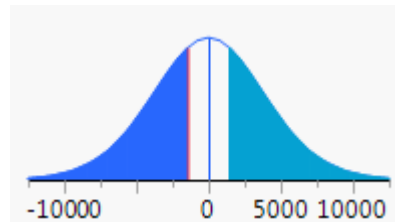
Bivariate

Analyze > Fit Y by X



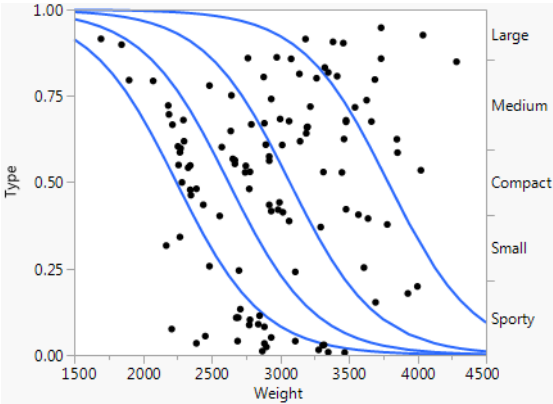
Oneway

Analyze > Fit Y by X

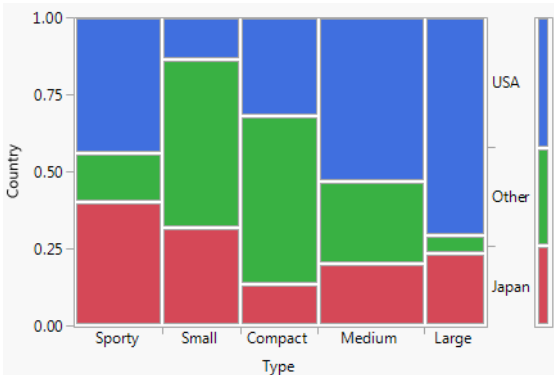


Oneway t Test

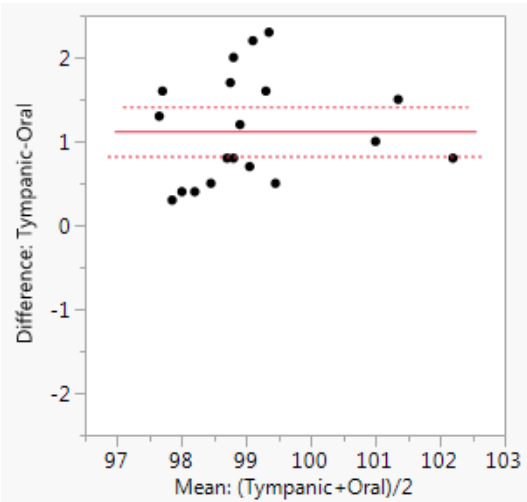
Analyze > Fit Y by X



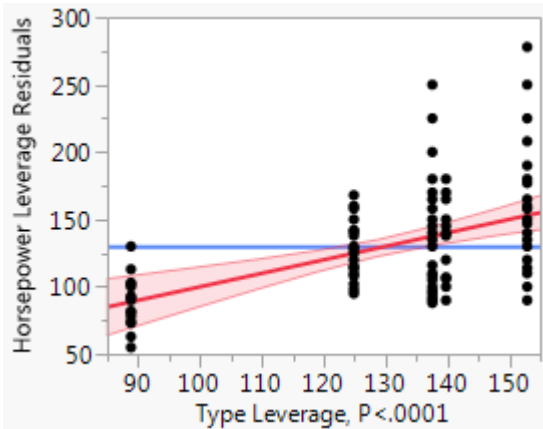
Logistic
Analyze > Fit Y by X



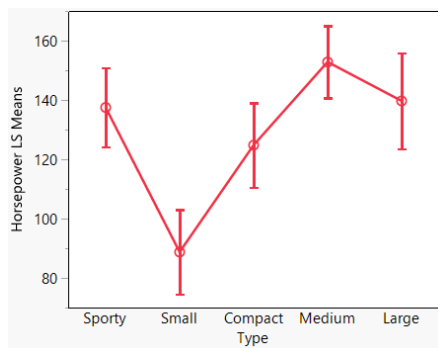
Mosaic Plot
Analyze > Fit Y by X



Matched Pairs
Analyze > Specialized Modeling > Matched Pairs

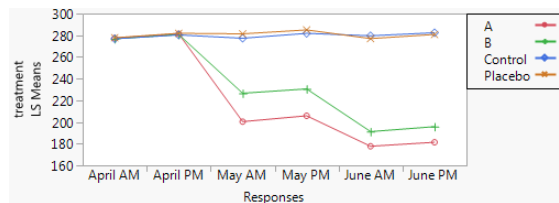


Leverage Plot
Analyze > Fit Model



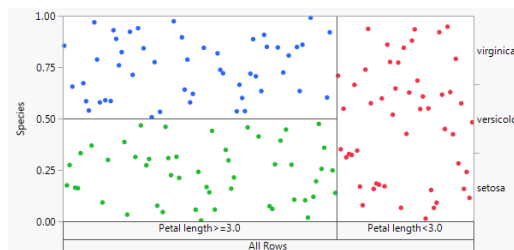
LS Means Plot

Analyze > Fit Model



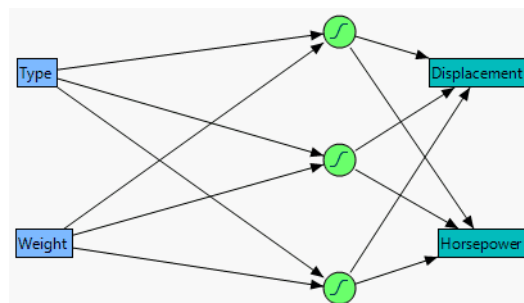
MANOVA

Analyze > Fit Model



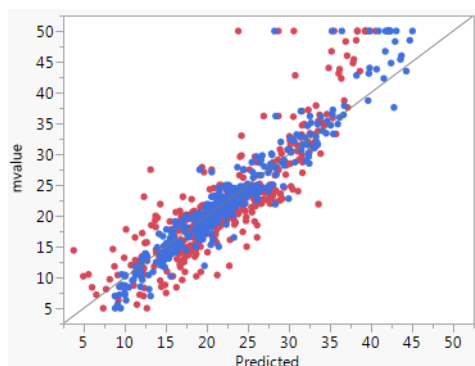
Partition

Analyze > Predictive Modeling > Partition



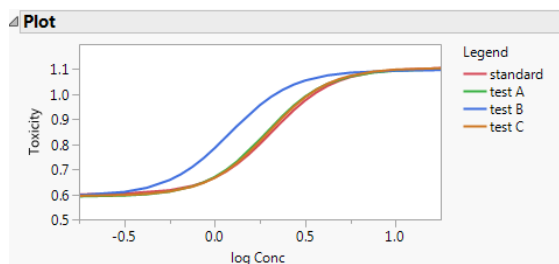
Neural Diagram

Analyze > Predictive Modeling > Neural



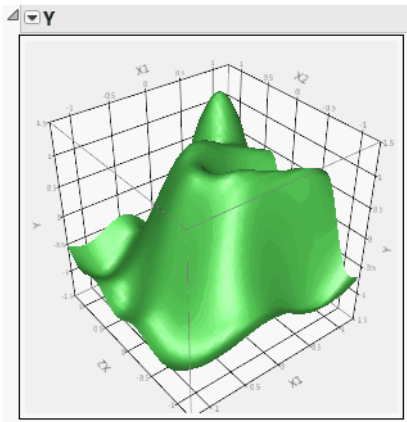
Actual by Predicted

Analyze > Predictive Modeling > Model Comparison



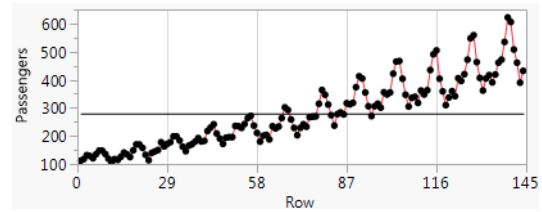
Nonlinear Fit

Analyze > Specialized Modeling > Nonlinear



Surface Profiler

Analyze > Specialized Modeling > Gaussian Process



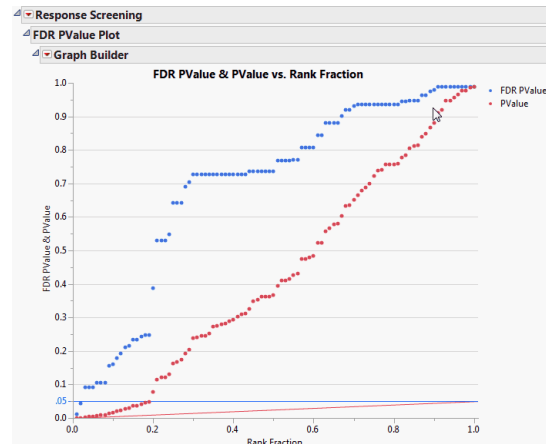
Time Series

Analyze > Specialized Modeling > Time Series

Term	Contrast	
Type	27.4115	
Model	-17.6588	
Type*Type	19.2417 *	
Type*Model	1.5953 *	
Model*Model	-1.0338 *	

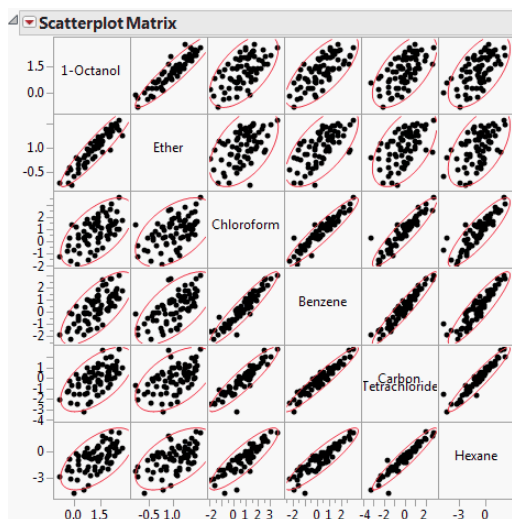
Screening

Analyze > Specialized Modeling > Specialized DOE Models > Fit Two Level Screening



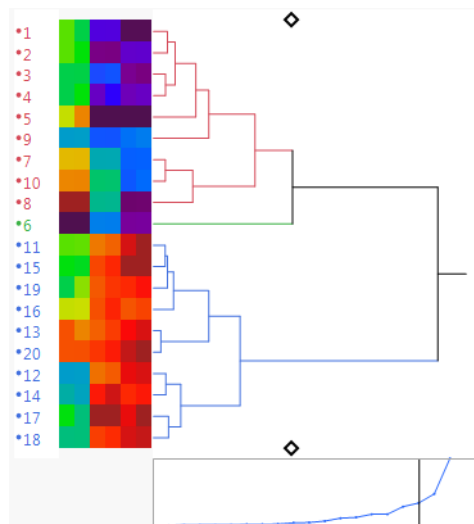
FDR pValue Plot

Analyze > Screening > Response Screening



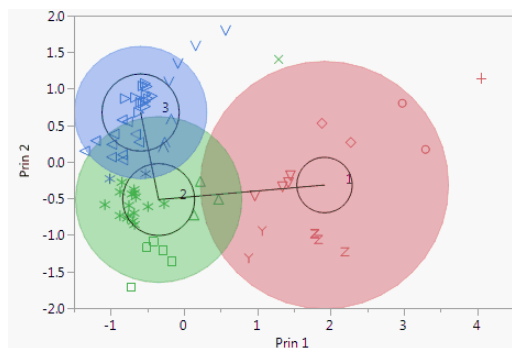
Scatterplot Matrix

Analyze > Multivariate Methods > Multivariate



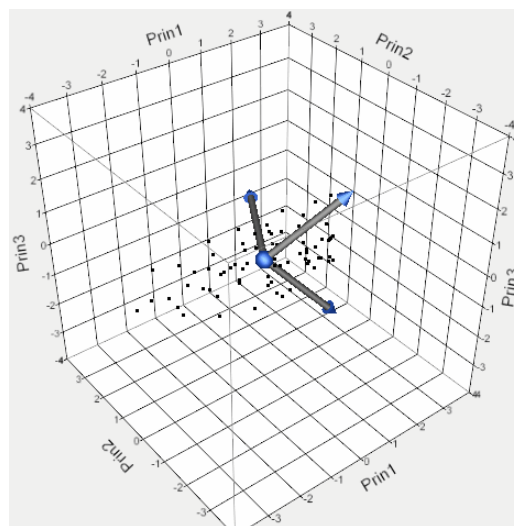
Dendrogram

Analyze > Clustering > Hierarchical Cluster



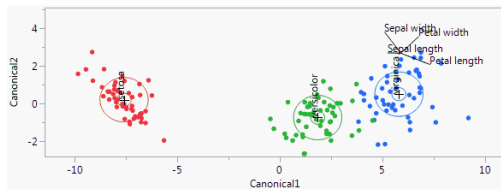
Self Organizing Map

Analyze > Clustering > K Means Cluster



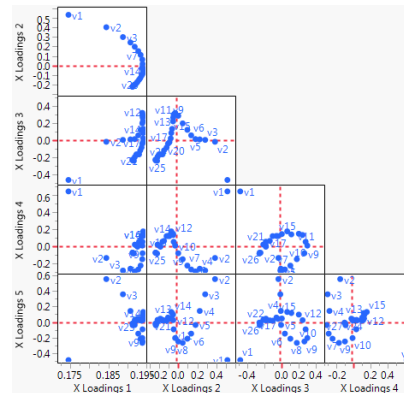
Principal Components

Analyze > Multivariate Methods > Principal Components



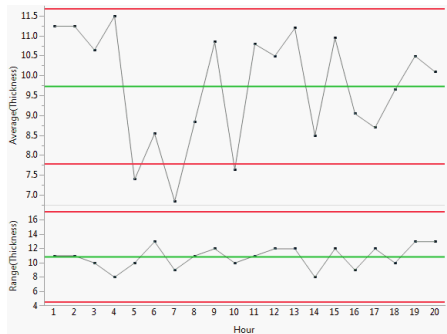
Canonical Plot

Analyze > Multivariate Methods > Discriminant



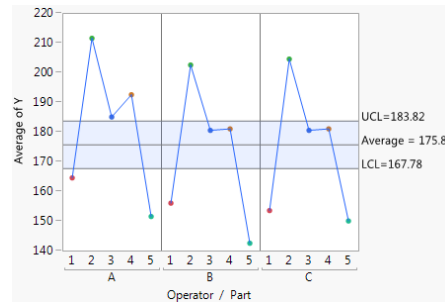
Loadings Plot

Analyze > Multivariate Methods > Partial Least Squares



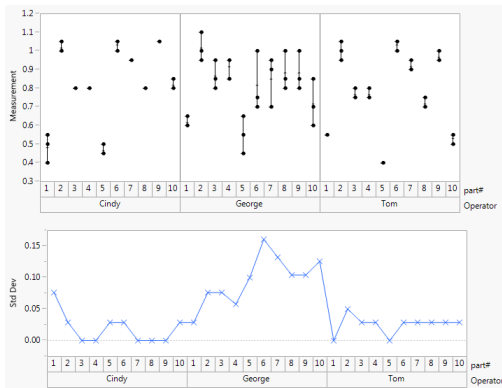
XBar and R Charts

Analyze > Quality and Process > Control Chart Builder

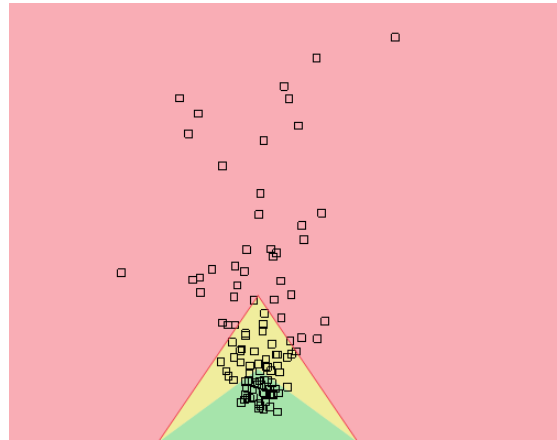


Average Chart

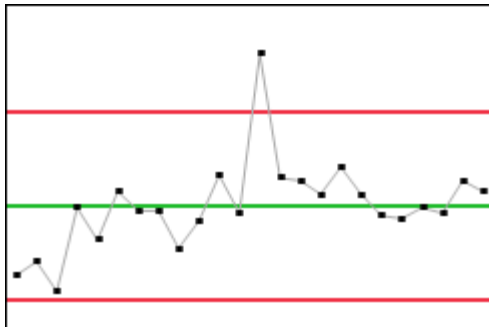
Analyze > Quality and Process > Measurement Systems Analysis

**Variability Chart**

Analyze > Quality and Process > Variability/Attribute Chart

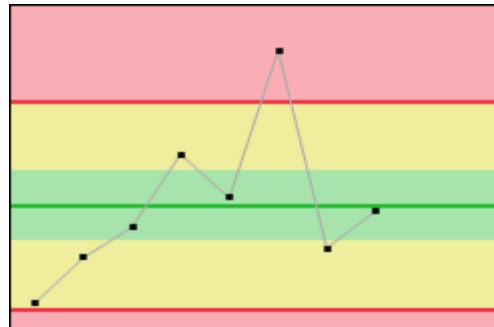
**Goal Plot**

Analyze > Quality and Process > Process Capability

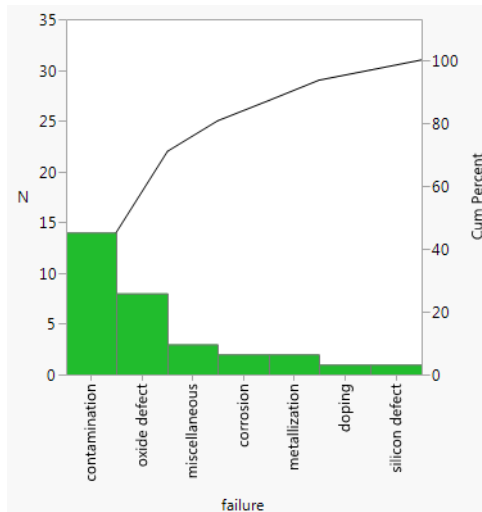
**Individual Measurement Chart**

Moving Range Chart

Analyze > Quality and Process > Control Chart > IR

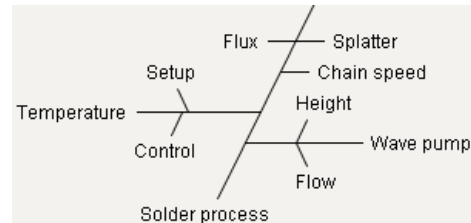
**XBar Chart**

Analyze > Quality and Process > Control Chart > XBar



Pareto Plot

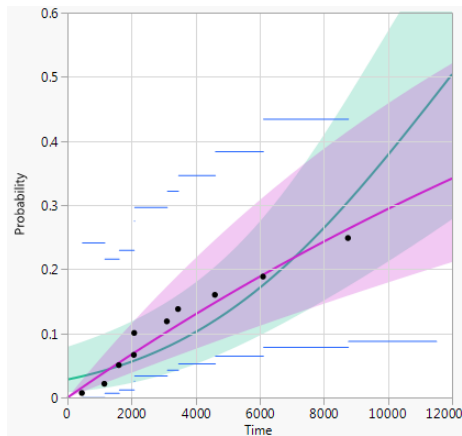
Analyze > Quality and Process > Pareto Plot



Ishikawa Chart

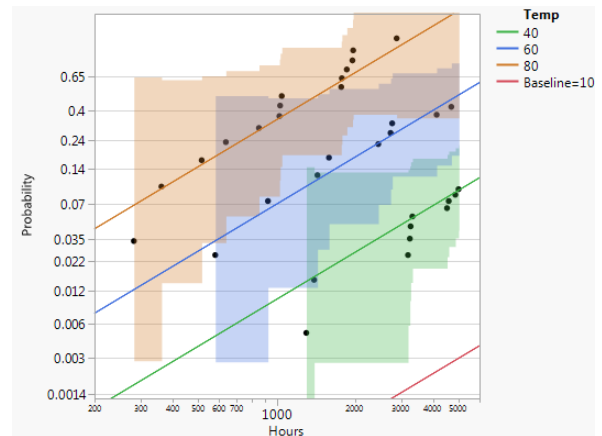
Fishbone Chart

Analyze > Quality and Process > Diagram



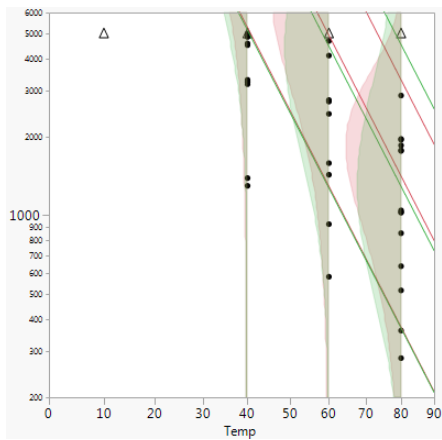
Compare Distributions

Analyze > Reliability and Survival > Life Distribution



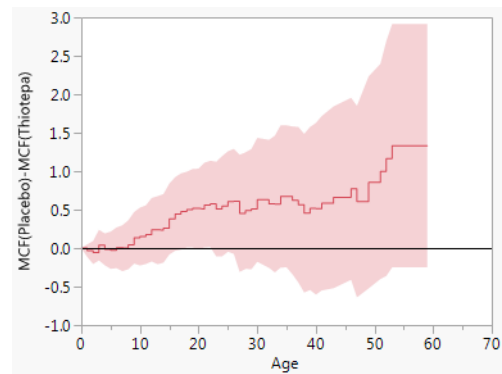
Nonparametric Overlay

Analyze > Reliability and Survival > Fit Life by X



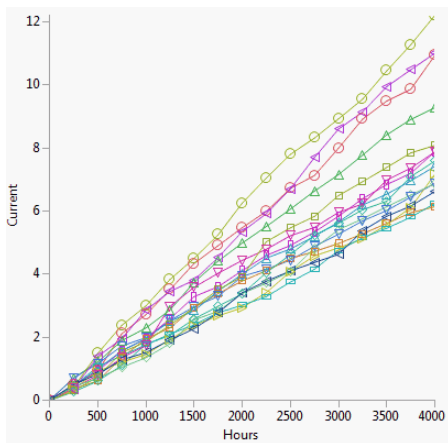
Scatterplot

Analyze > Reliability and Survival > Fit Life by X



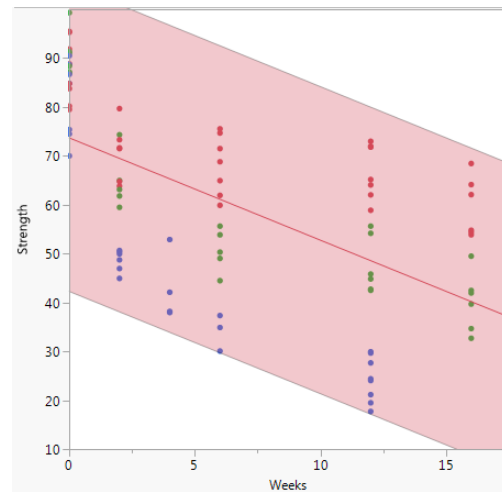
MCF Plot

Analyze > Reliability and Survival > Recurrence Analysis



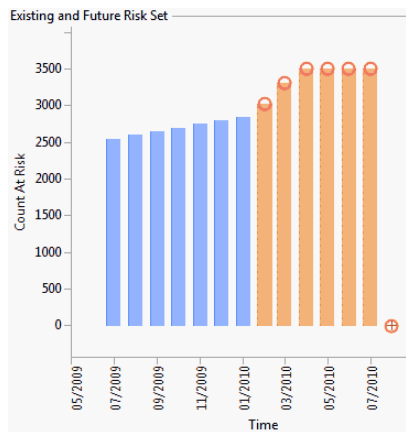
Overlay

Analyze > Reliability and Survival > Degradation



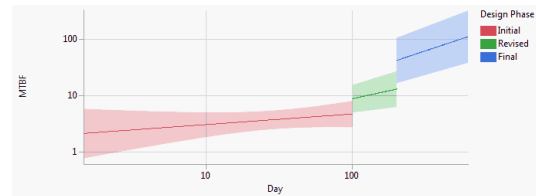
Prediction Interval

Analyze > Reliability and Survival > Destructive Degradation



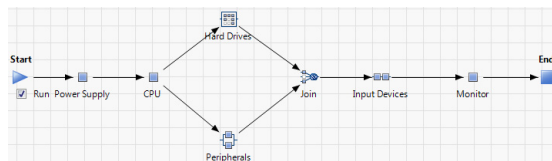
Forecast

Analyze > Reliability and Survival > Reliability
Forecast



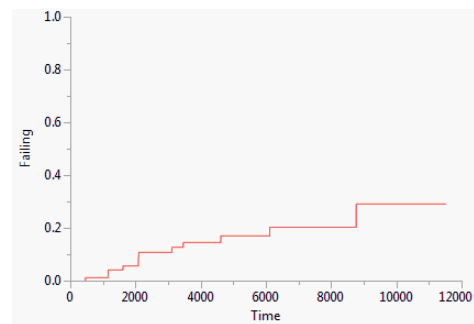
Piecewise Weibull NHPP

Analyze > Reliability and Survival > Reliability
Growth



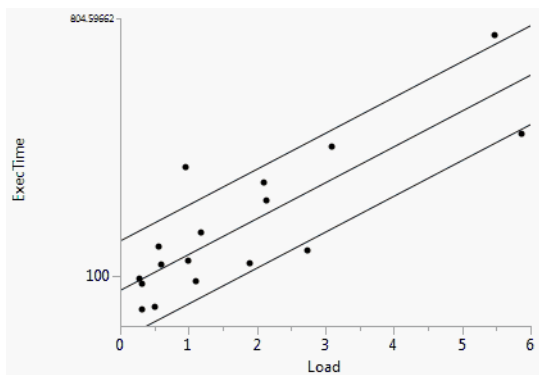
Reliability Block Diagram

Analyze > Reliability and Survival > Reliability Block
Diagram

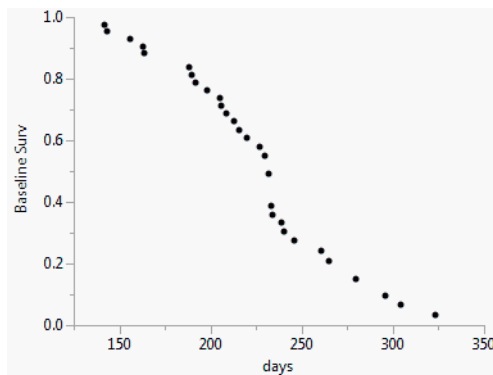


Failure Plot

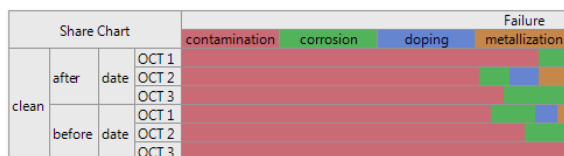
Analyze > Reliability and Survival > Survival

**Survival Quantiles**

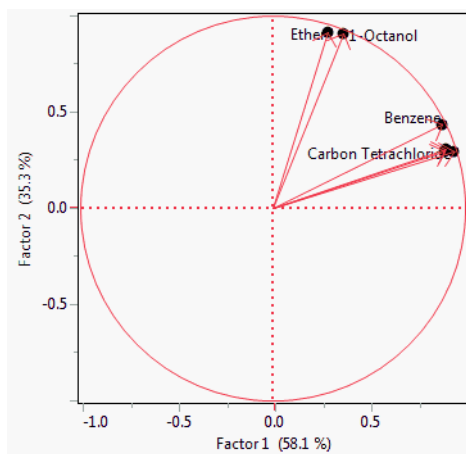
Analyze > Reliability and Survival > Fit Parametric Survival

**Baseline Survival**

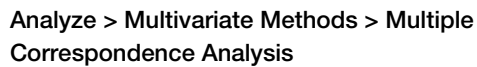
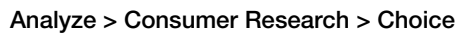
Analyze > Reliability and Survival > Fit Proportional Hazards

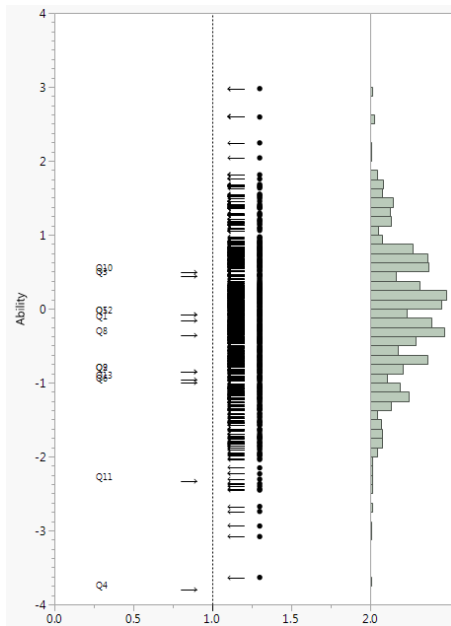
**Mixture Profiler**

Analyze > Consumer Research > Categorical

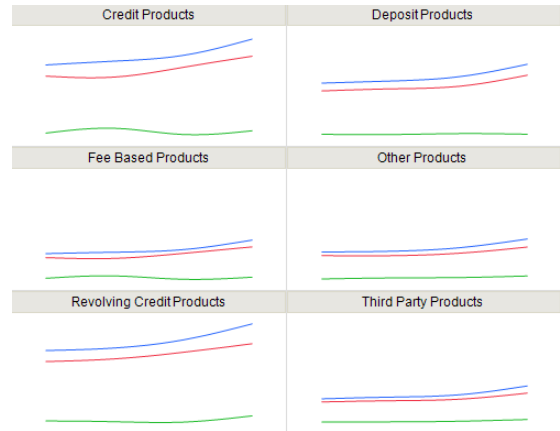
**Factor Loading Plot**

Analyze > Multivariate Methods > Factor Analysis

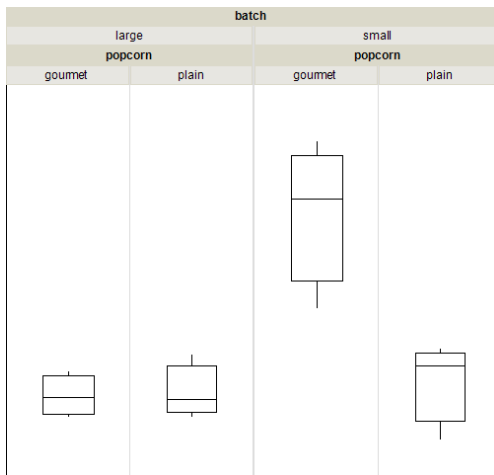


**Dual Plot**

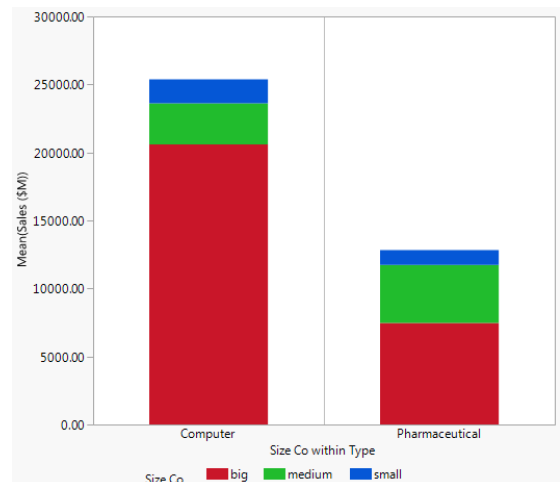
Analyze > Multivariate Methods > Item Analysis

**Line Graphs**

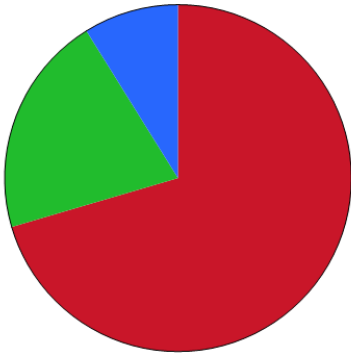
Graph > Graph Builder

**Box Plots**

Graph > Graph Builder

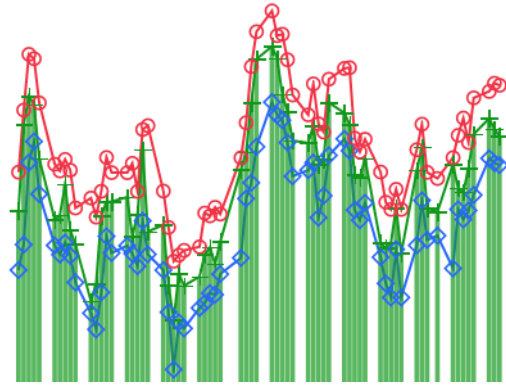
**Stacked Bar Chart**

Graph > Graph Builder



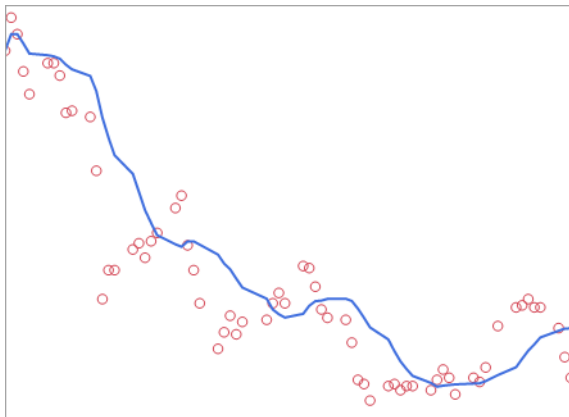
Pie Chart

Graph > Graph Builder



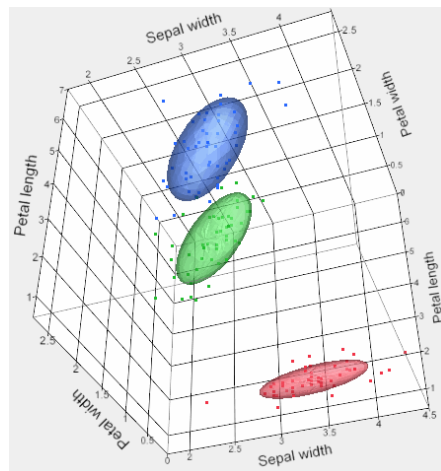
Needle and Line Chart

Graph > Graph Builder



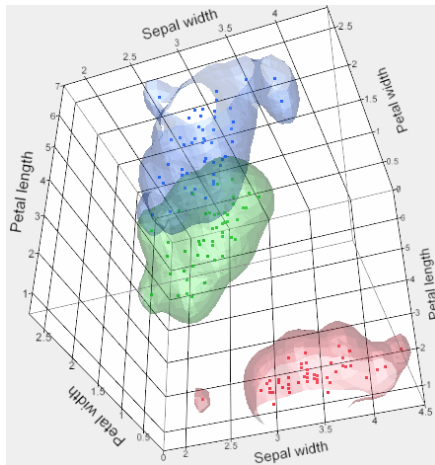
Smoother

Graph > Graph Builder

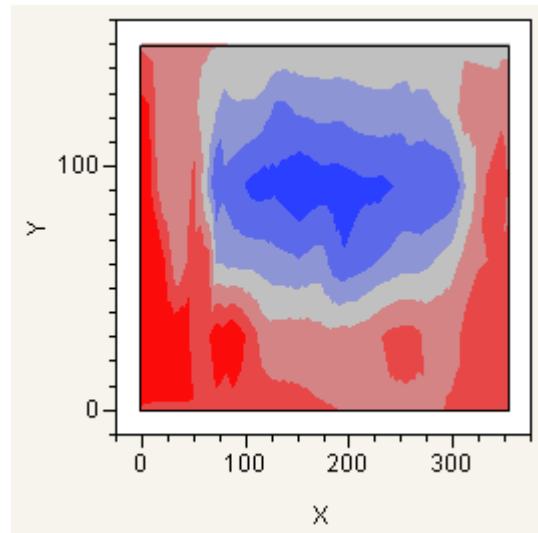


Three Dimensional Scatterplot

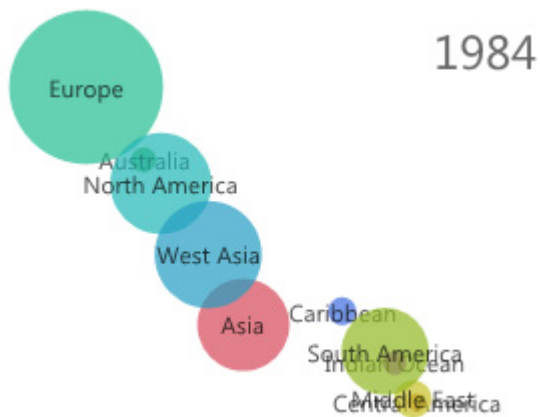
Graph > Scatterplot 3D



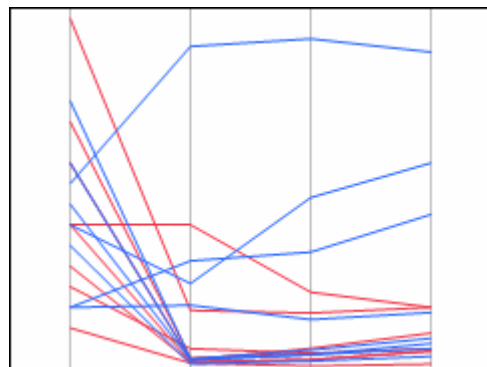
Three Dimensional Scatterplot
Graph > Scatterplot 3D



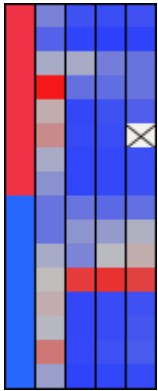
Contour Plot
Graph > Graph Builder



Bubble Plot
Graph > Bubble Plot

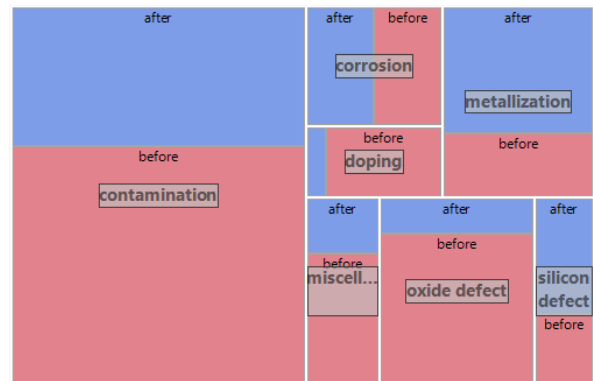


Parallel Plot
Graph > Graph Builder



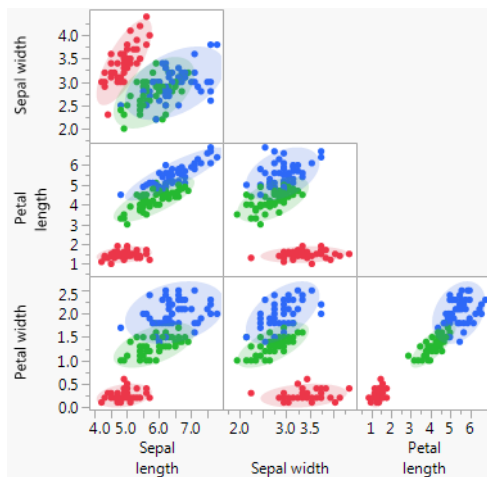
Cell Plot

Graph > Cell Plot



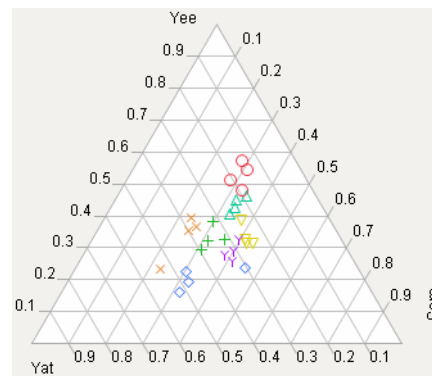
Treemap

Graph > Graph Builder



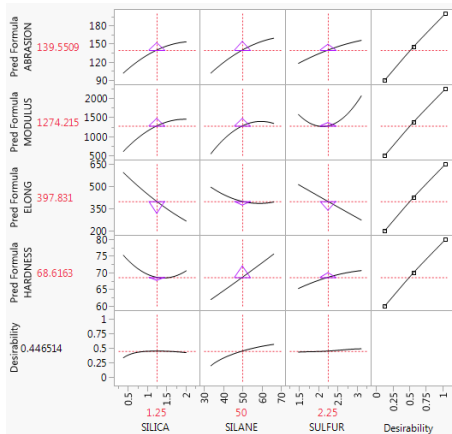
Scatterplot Matrix

Graph > Scatterplot Matrix



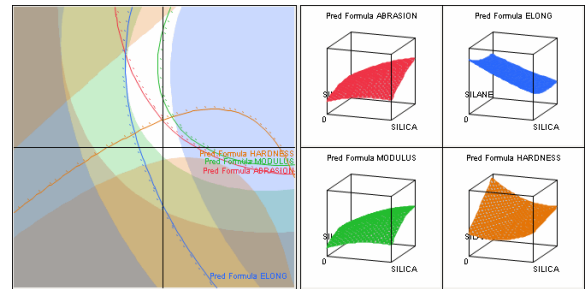
Ternary Plot

Graph > Ternary Plot



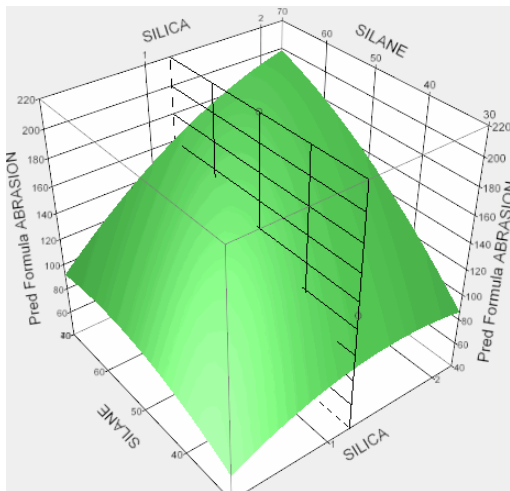
Prediction Profiler

Graph > Profiler



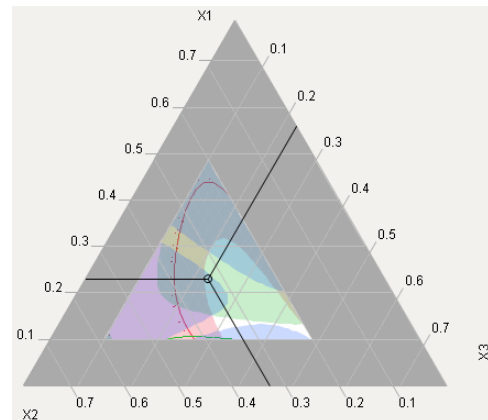
Contour Profiler

Graph > Contour Profiler



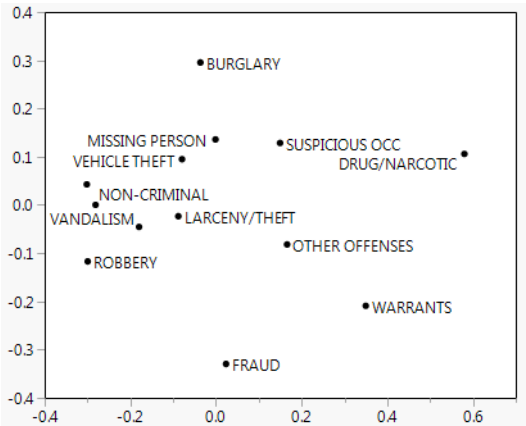
Surface Plot

Graph > Surface Plot

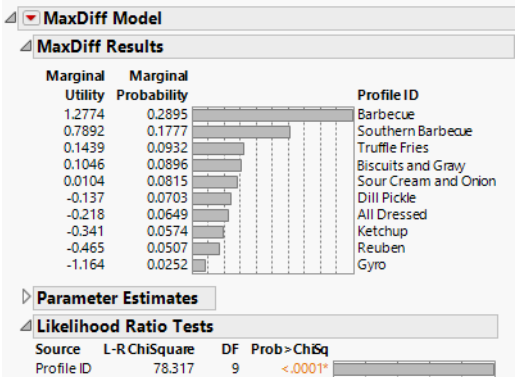


Mixture Profiler

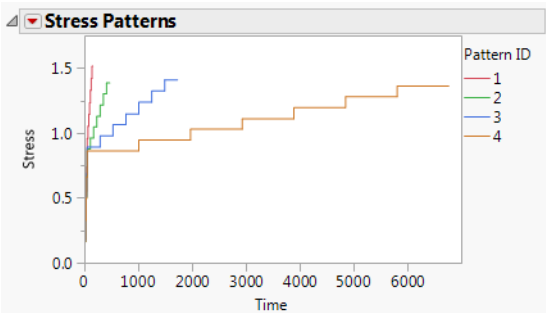
Graph > Mixture Profiler



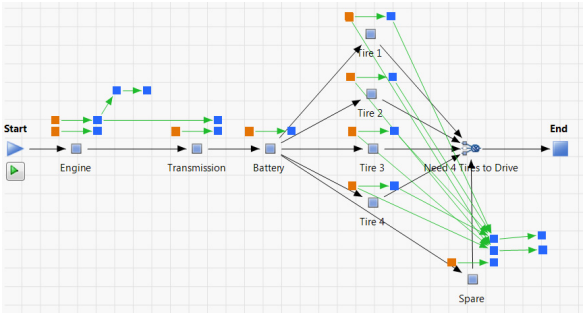
Multidimensional Scaling
Analyze > Multivariate Methods > Multidimensional
Scaling



MaxDiff
Analyze > Consumer Research > MaxDiff



Stress Patterns Plot
Analyze > Reliability and Survival > Cumulative
Damage



Repairable Systems Simulation
Analyze > Reliability and Survival > Repairable
Systems Simulation

Parameter Estimates

Cluster	Overall	sex		marital status	
		Female	Male	Married	Single
Cluster 1	0.28596	0.3764	0.6236	0.4163	0.5837
Cluster 2	0.25892	0.4067	0.5933	0.6742	0.3258
Cluster 3	0.20696	0.6794	0.3206	0.7524	0.2476
Cluster 4	0.19717	0.4706	0.5294	0.9922	0.0078
Cluster 5	0.05099	0.1840	0.8160	0.0329	0.9671

Cluster	Overall	sex	marital status
Cluster 1	0.28596		
Cluster 2	0.25892		
Cluster 3	0.20696		
Cluster 4	0.19717		
Cluster 5	0.05099		

Latent Class Analysis

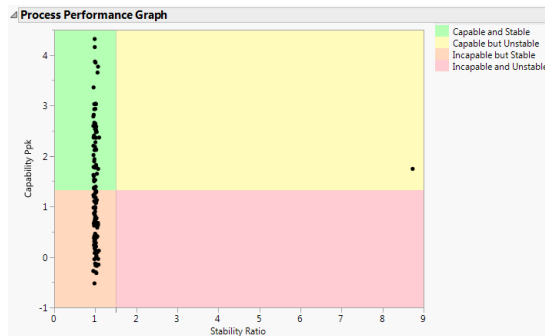
Analyze > Clustering > Latent Class Analysis

Predictor Screening

Predictor	Contribution	Portion	Banding?	
				Rank
ink pct	30.0387	0.1761		1
varnish pct	19.7704	0.1159		2
solvent pct	16.6355	0.0975		3
press	15.9053	0.0932		4
press speed	10.6091	0.0622		5
ESA Voltage	8.2611	0.0484		6
press type	5.9066	0.0346		7
roller durometer	4.6799	0.0274		8
viscosity	4.4441	0.0261		9
grain screened	4.3653	0.0256		10
unit number	4.1686	0.0244		11
ink type	4.1140	0.0241		12
ESA Amperage	4.0286	0.0236		13
solvent type	3.8420	0.0225		14
paper mill location	3.6955	0.0217		15
blade pressure	3.4523	0.0202		16
hardener	3.4171	0.0200		17
humidity	3.2859	0.0193		18
proof cut	3.2396	0.0190		19
ink temperature	2.5650	0.0150		20
type on cylinder	2.2837	0.0134		21
anode space ratio	2.1421	0.0126		22
roughness	1.7253	0.0101		23
cylinder size	1.2731	0.0075		24
caliper	1.2634	0.0074		25
paper type	1.1367	0.0067		26
current density	1.0214	0.0060		27
proof on ctd ink	0.8997	0.0053		28
blade mfg	0.8482	0.0050		29
wax	0.7817	0.0046		30
plating tank	0.7267	0.0043		31
direct steam	0.0663	0.0004		32
chrome content	0.0000	0.0000		33

Predictor Screening

Analyze > Screening > Predictor Screening



Process Screening

Analyze > Screening > Process Screening

Text Explorer for Survey Response

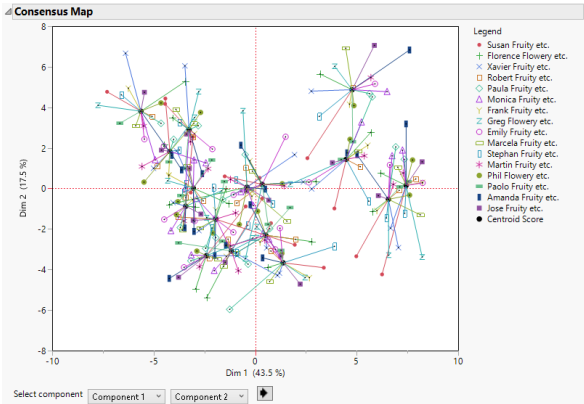
Number of Terms	Number of Cases	Total Tokens	Tokens per Case	Number of Non-empty Cases	Portion Non-empty per Case
413	194	1921	9.90206	150	0.7732

Term and Phrase Lists

Term	Count	Phrase	Count	N
cat	55	video of the cat	5	4
dogs	48	all the time	5	3
dog	46	sit in my lap	4	4
all	19	cat food	4	2
cats	17	dog food	4	2
lap	14	dogs have been	3	3
been	13	barking all	3	2
barking	12	dog barks	3	2
out	11	duck hunting	3	2
video	11	funny video	3	2
while	11	great job	3	2
food	10	last week	3	2
walk	10	stop barking	3	2
mice	9	cat in my lap	2	4
over	9	dogs do a great	2	4
down	8	hunting with the dogs	2	4
just	8	out for a sled	2	4
like	8	sled in the winter	2	4
take	8	take the huskies out	2	4
bark	7	video of a cat	2	4
huskies	7	while we were away	2	4
through	7	all the dogs	2	3
time	7	around the block	2	3
always	6	been barking all	2	3
cattle	6	job of herding	2	3
sheep	6	lap and purr	2	3
sheep	6	take the huskies	2	3

Text Explorer

Analyze > Text Explorer



Multiple Factor Analysis
Analyze > Consumer Research > Multiple Factor
Analysis

Chapter 1

Learn about JMP

Documentation and Additional Resources


This chapter includes details about JMP documentation, such as book conventions, descriptions of each JMP book, the Help system, and where to find other support.

Contents

Formatting Conventions	39
JMP Documentation	40
JMP Documentation Library	40
JMP Help	46
Additional Resources for Learning JMP	46
Tutorials	46
Sample Data Tables	47
Learn about Statistical and JSL Terms	47
Learn JMP Tips and Tricks	47
Tooltips	47
JMP User Community	48
JMP Books by Users	48
The JMP Starter Window	48
Technical Support	49

Formatting Conventions

The following conventions help you relate written material to information that you see on your screen:

- Sample data table names, column names, pathnames, filenames, file extensions, and folders appear in Helvetica font.
- Code appears in *Lucida Sans Typewriter* font.
- Code output appears in *Lucida Sans Typewriter* italic font and is indented farther than the preceding code.
- **Helvetica bold** formatting indicates items that you select to complete a task:
 - buttons
 - check boxes
 - commands
 - list names that are selectable
 - menus
 - options
 - tab names
 - text boxes
- The following items appear in italics:
 - words or phrases that are important or have definitions specific to JMP
 - book titles
 - variables
- Features that are for JMP Pro only are noted with the JMP Pro icon . For an overview of JMP Pro features, visit <https://www.jmp.com/software/pro/>.

Note: Special information and limitations appear within a Note.

Tip: Helpful information appears within a Tip.

JMP Documentation

JMP offers documentation in various formats, from print books and Portable Document Format (PDF) to electronic books (e-books).

- Open the PDF versions from the **Help > Books** menu.
- All books are also combined into one PDF file, called *JMP Documentation Library*, for convenient searching. Open the *JMP Documentation Library* PDF file from the **Help > Books** menu.
- You can also purchase printed documentation and e-books on the SAS website:
<https://www.sas.com/store/search.ep?keyWords=JMP>

JMP Documentation Library

The following table describes the purpose and content of each book in the JMP library.

Document Title	Document Purpose	Document Content
<i>Discovering JMP</i>	If you are not familiar with JMP, start here.	Introduces you to JMP and gets you started creating and analyzing data. Also learn how to share your results.
<i>Using JMP</i>	Learn about JMP data tables and how to perform basic operations.	Covers general JMP concepts and features that span across all of JMP, including importing data, modifying columns properties, sorting data, and connecting to SAS.

Document Title	Document Purpose	Document Content
<i>Basic Analysis</i>	Perform basic analysis using this document.	<p>Describes these Analyze menu platforms:</p> <ul style="list-style-type: none"> • Distribution • Fit Y by X • Tabulate • Text Explorer <p>Covers how to perform bivariate, one-way ANOVA, and contingency analyses through Analyze > Fit Y by X. How to approximate sampling distributions using bootstrapping and how to perform parametric resampling with the Simulate platform are also included.</p>
<i>Essential Graphing</i>	Find the ideal graph for your data.	<p>Describes these Graph menu platforms:</p> <ul style="list-style-type: none"> • Graph Builder • Scatterplot 3D • Contour Plot • Bubble Plot • Parallel Plot • Cell Plot • Scatterplot Matrix • Ternary Plot • Treemap • Chart • Overlay Plot <p>The book also covers how to create background and custom maps.</p>
<i>Profilers</i>	Learn how to use interactive profiling tools, which enable you to view cross-sections of any response surface.	Covers all profilers listed in the Graph menu. Analyzing noise factors is included along with running simulations using random inputs.

Document Title	Document Purpose	Document Content
<i>Design of Experiments Guide</i>	Learn how to design experiments and determine appropriate sample sizes.	Covers all topics in the DOE menu and the Specialized DOE Models menu item in the Analyze > Specialized Modeling menu.
<i>Fitting Linear Models</i>	Learn about Fit Model platform and many of its personalities.	<div>Describes these personalities, all available within the Analyze menu Fit Model platform:</div> <ul style="list-style-type: none">• Standard Least Squares• Stepwise• Generalized Regression• Mixed Model• MANOVA• Loglinear Variance• Nominal Logistic• Ordinal Logistic• Generalized Linear Model

Document Title	Document Purpose	Document Content
<i>Predictive and Specialized Modeling</i>	Learn about additional modeling techniques.	<p>Describes these Analyze > Predictive Modeling menu platforms:</p> <ul style="list-style-type: none"> • Modeling Utilities • Neural • Partition • Bootstrap Forest • Boosted Tree • K Nearest Neighbors • Naive Bayes • Model Comparison • Formula Depot <p>Describes these Analyze > Specialized Modeling menu platforms:</p> <ul style="list-style-type: none"> • Fit Curve • Nonlinear • Functional Data Explorer • Gaussian Process • Time Series • Matched Pairs <p>Describes these Analyze > Screening menu platforms:</p> <ul style="list-style-type: none"> • Response Screening • Process Screening • Predictor Screening • Association Analysis • Process History Explorer <p>The platforms in the Analyze > Specialized Modeling > Specialized DOE Models menu are described in <i>Design of Experiments Guide</i>.</p>


Document Title	Document Purpose	Document Content
<i>Multivariate Methods</i>	Read about techniques for analyzing several variables simultaneously.	<p>Describes these Analyze > Multivariate Methods menu platforms:</p> <ul style="list-style-type: none"> • Multivariate • Principal Components • Discriminant • Partial Least Squares • Multiple Correspondence Analysis • Factor Analysis • Multidimensional Scaling • Item Analysis <p>Describes these Analyze > Clustering menu platforms:</p> <ul style="list-style-type: none"> • Hierarchical Cluster • K Means Cluster • Normal Mixtures • Latent Class Analysis • Cluster Variables
<i>Quality and Process Methods</i>	Read about tools for evaluating and improving processes.	<p>Describes these Analyze > Quality and Process menu platforms:</p> <ul style="list-style-type: none"> • Control Chart Builder and individual control charts • Measurement Systems Analysis • Variability / Attribute Gauge Charts • Process Capability • Pareto Plot • Diagram • Manage Spec Limits

Document Title	Document Purpose	Document Content
<i>Reliability and Survival Methods</i>	Learn to evaluate and improve reliability in a product or system and analyze survival data for people and products.	Describes these Analyze > Reliability and Survival menu platforms: <ul style="list-style-type: none"> • Life Distribution • Fit Life by X • Cumulative Damage • Recurrence Analysis • Degradation and Destructive Degradation • Reliability Forecast • Reliability Growth • Reliability Block Diagram • Repairable Systems Simulation • Survival • Fit Parametric Survival • Fit Proportional Hazards
<i>Consumer Research</i>	Learn about methods for studying consumer preferences and using that insight to create better products and services.	Describes these Analyze > Consumer Research menu platforms: <ul style="list-style-type: none"> • Categorical • Choice • MaxDiff • Uplift • Multiple Factor Analysis
<i>Scripting Guide</i>	Learn about taking advantage of the powerful JMP Scripting Language (JSL).	Covers a variety of topics, such as writing and debugging scripts, manipulating data tables, constructing display boxes, and creating JMP applications.
<i>JSL Syntax Reference</i>	Read about many JSL functions on functions and their arguments, and messages that you send to objects and display boxes.	Includes syntax, examples, and notes for JSL commands.

Note: The **Books** menu also contains two reference cards that can be printed: The *Menu Card* describes JMP menus, and the *Quick Reference* describes JMP keyboard shortcuts.

JMP Help

JMP Help is an abbreviated version of the documentation library that provides targeted information. You can open JMP Help in several ways:

- On Windows, press the F1 key to open the Help system window.
- Get help on a specific part of a data table or report window. Select the Help tool  from the **Tools** menu and then click anywhere in a data table or report window to see the Help for that area.
- Within a JMP window, click the **Help** button.
- Search and view JMP Help on Windows using the **Help > Help Contents**, **Search Help**, and **Help Index** options. On Mac, select **Help > JMP Help**.
- Search the Help at <https://jmp.com/support/help/> (English only).

Additional Resources for Learning JMP

In addition to JMP documentation and JMP Help, you can also learn about JMP using the following resources:

- [“Tutorials”](#)
- [“Sample Data Tables”](#)
- [“Learn about Statistical and JSL Terms”](#)
- [“Learn JMP Tips and Tricks”](#)
- [“Tooltips”](#)
- [“JMP User Community”](#)
- [“JMP Books by Users”](#)
- [“The JMP Starter Window”](#)

Tutorials

You can access JMP tutorials by selecting **Help > Tutorials**. The first item on the **Tutorials** menu is **Tutorials Directory**. This opens a new window with all the tutorials grouped by category.

If you are not familiar with JMP, then start with the **Beginners Tutorial**. It steps you through the JMP interface and explains the basics of using JMP.

The rest of the tutorials help you with specific aspects of JMP, such as designing an experiment and comparing a sample mean to a constant.

Sample Data Tables

All of the examples in the JMP documentation suite use sample data. Select **Help > Sample Data Library** to open the sample data directory.

To view an alphabetized list of sample data tables or view sample data within categories, select **Help > Sample Data**.

Sample data tables are installed in the following directory:

On Windows: C:\Program Files\SAS\JMP\14\Samples\Data

On Macintosh: \Library\Application Support\JMP\14\Samples\Data

In JMP Pro, sample data is installed in the JMPPRO (rather than JMP) directory. In JMP Shrinkwrap, sample data is installed in the JMPSW directory.

To view examples using sample data, select **Help > Sample Data** and navigate to the Teaching Resources section. To learn more about the teaching resources, visit <https://jmp.com/tools>.

Learn about Statistical and JSL Terms

The **Help** menu contains the following indexes:

Statistics Index Provides definitions of statistical terms.

Scripting Index Lets you search for information about JSL functions, objects, and display boxes. You can also edit and run sample scripts from the Scripting Index.

Learn JMP Tips and Tricks

When you first start JMP, you see the Tip of the Day window. This window provides tips for using JMP.

To turn off the Tip of the Day, clear the **Show tips at startup** check box. To view it again, select **Help > Tip of the Day**. Or, you can turn it off using the Preferences window.

Tooltips

JMP provides descriptive tooltips when you place your cursor over items, such as the following:

- Menu or toolbar options

- Labels in graphs
- Text results in the report window (move your cursor in a circle to reveal)
- Files or windows in the Home Window
- Code in the Script Editor

Tip: On Windows, you can hide tooltips in the JMP Preferences. Select **File > Preferences > General** and then deselect **Show menu tips**. This option is not available on Macintosh.

JMP User Community

The JMP User Community provides a range of options to help you learn more about JMP and connect with other JMP users. The learning library of one-page guides, tutorials, and demos is a good place to start. And you can continue your education by registering for a variety of JMP training courses.

Other resources include a discussion forum, sample data and script file exchange, webcasts, and social networking groups.

To access JMP resources on the website, select **Help > JMP User Community** or visit <https://community.jmp.com/>.

JMP Books by Users

Additional books about using JMP that are written by JMP users are available on the JMP website:

https://www.jmp.com/en_us/software/books.html

The JMP Starter Window

The JMP Starter window is a good place to begin if you are not familiar with JMP or data analysis. Options are categorized and described, and you launch them by clicking a button. The JMP Starter window covers many of the options found in the Analyze, Graph, Tables, and File menus. The window also lists JMP Pro features and platforms.

- To open the JMP Starter window, select **View (Window on the Macintosh) > JMP Starter**.
- To display the JMP Starter automatically when you open JMP on Windows, select **File > Preferences > General**, and then select **JMP Starter** from the Initial JMP Window list. On Macintosh, select **JMP > Preferences > Initial JMP Starter Window**.

Technical Support

JMP technical support is provided by statisticians and engineers educated in SAS and JMP, many of whom have graduate degrees in statistics or other technical disciplines.

Many technical support options are provided at <https://www.jmp.com/support>, including the technical support phone number.

Chapter 2

Introducing JMP Basic Concepts

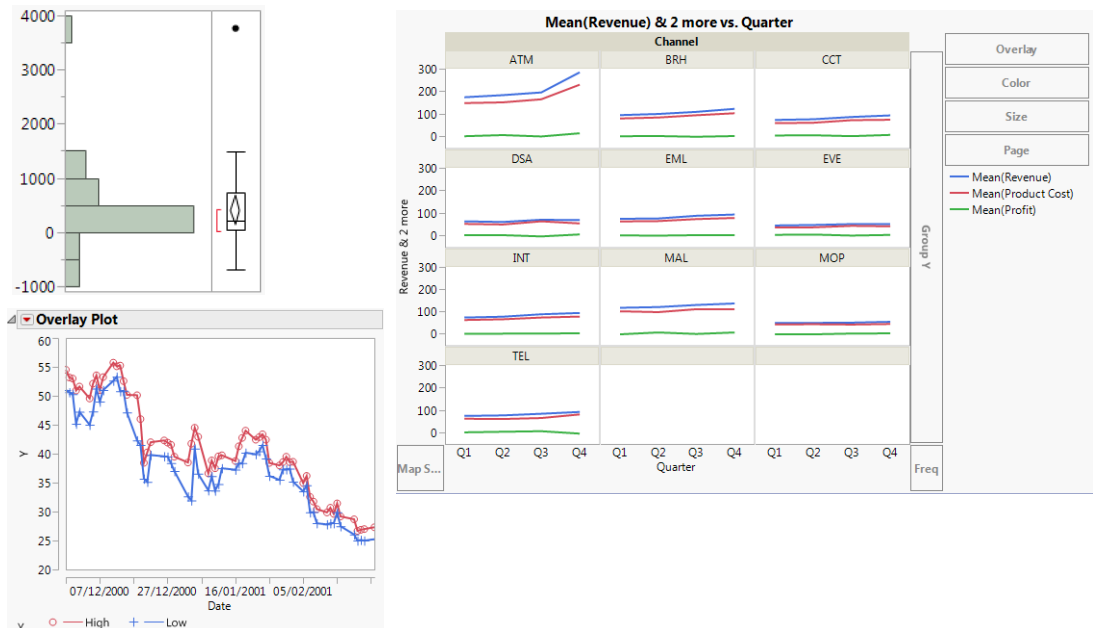
JMP (pronounced *jump*) is a powerful and interactive data visualization and statistical analysis tool. Use JMP to learn more about your data by performing analyses and interacting with the data using data tables, graphs, charts, and reports.

JMP enables researchers to perform a wide range of statistical analyses and modeling. JMP is equally useful to the business analyst who wants to quickly uncover trends and patterns in data. With JMP, you do not have to be an expert in statistics to get information from your data.

For example, you can use JMP to do the following:

- Create interactive graphs and charts to explore your data and discover relationships.
- Discover patterns of variation across many variables at once.
- Explore and summarize large amounts of data.
- Develop powerful statistical models to predict the future.

Figure 2.1 Examples of JMP Reports





Contents

Concepts That You Should Know	53
How Do I Get Started?	53
Starting JMP	54
Using Sample Data	56
Understanding Data Tables	56
Understanding the JMP Workflow	58
Step 1: Launching a Platform and Viewing Results	59
Step 2: Removing the Box Plot	61
Step 3: Requesting Additional Output	61
Step 4: Interacting with Platform Results	62
How is JMP Different from Excel?	63

Concepts That You Should Know

Before you begin using JMP, you should be familiar with these concepts:

- Enter, view, edit, and manipulate data using JMP *data tables*.
- Select a *platform* from the **Analyze**, **Graph**, or **DOE** menus. Platforms contain interactive windows that you use to analyze data and work with graphs.
- Platforms use these windows:
 - *Launch windows* where you set up and run your analysis.
 - *Report windows* showing the output of your analysis.
- Report windows normally contain the following items:
 - A graph of some type (such as a scatterplot or a chart).
 - Specific *reports* that you can show or hide using the *disclosure icon* .
 - Platform *options* that are located within *red triangle menus* .

How Do I Get Started?

The general workflow in JMP is simple:

1. Get your data into JMP.
2. Select a platform and complete its launch window.
3. Explore your results and discover where your data takes you.

This workflow is described in more detail in [“Understanding the JMP Workflow”](#) on page 58.

Typically, you start your work in JMP by using graphs to visualize individual variables and relationships among your variables. Graphs make it easy to see this information, and to see the deeper questions to ask. Then you use analysis platforms to dig deeper into your problems and find solutions.

- The [“Work with Your Data”](#) chapter on page 65 shows you how to get data into JMP.
- The [“Visualize Your Data”](#) chapter on page 95 shows you how to use some of the useful graphs JMP provides to look more closely at your data.
- The [“Analyze Your Data”](#) chapter on page 131 shows you how to use some of the analysis platforms.
- The [“The Big Picture”](#) chapter on page 167 shows you how to analyze distributions, patterns, and similar values in several platforms.

Each chapter teaches through examples. The following sections in this chapter describe data tables and general concepts for working in JMP.

Starting JMP

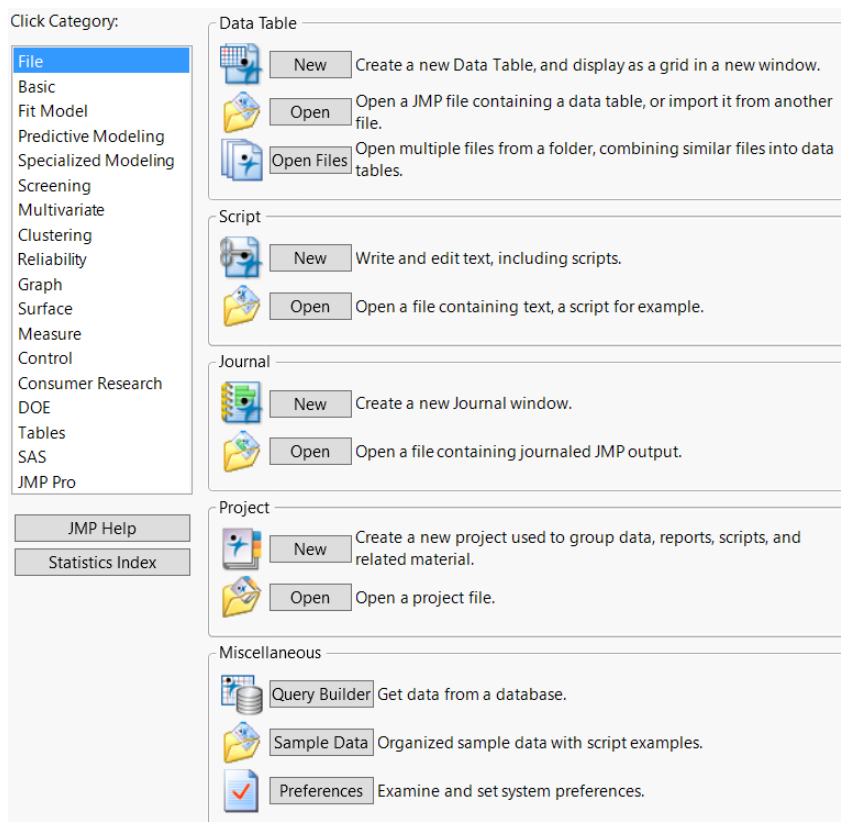
Start JMP in two ways:

- Double-click the JMP icon, normally found on your desktop. This starts JMP but does not open any existing JMP files.
- Double-click an existing JMP file. This starts JMP and opens the file.

The initial view of JMP includes the Tip of the Day window and the Home Window on Windows; on Macintosh, the Tip of the Day and JMP Starter, and Home Window initially appear.

The JMP Starter window classifies actions and platforms using categories.

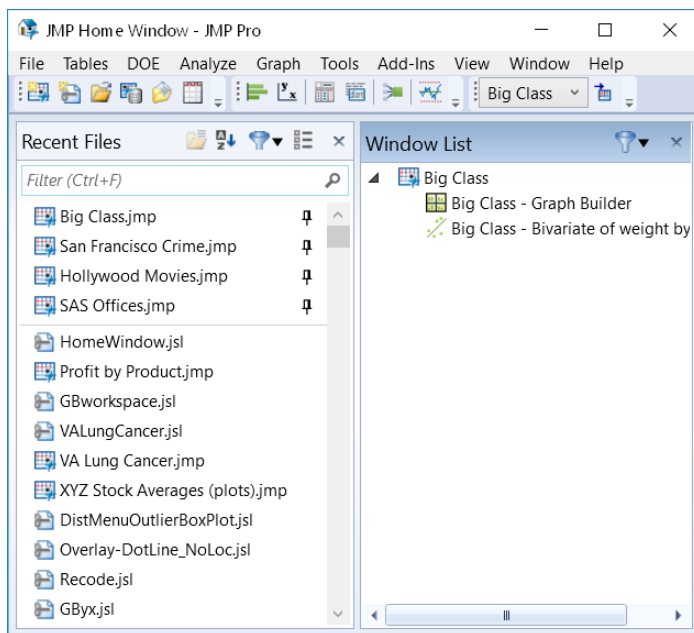
Figure 2.2 The JMP Starter



On the left is a list of categories. Click a category to see the features and the commands related to that category. The JMP Starter also lists JMP Pro features and platforms.

The Home Window helps you organize and access files in JMP.

Figure 2.3 The Home Window on Windows



To open the Home Window on Windows, select **View > Home Window**. On Macintosh, select **Window > JMP Home**. The Home Window includes links to the following:

- the data tables and report windows that are currently open
- files that you have opened recently

For more details about the Home Window, see the Get Started chapter in the *Using JMP* book.

Almost all JMP windows contain a menu bar and a toolbar. You can find most JMP features in three ways:

- using the menu bar
- using the toolbar buttons
- using the buttons on the JMP Starter window

About the Menu Bar and Toolbars

The menus and toolbars are hidden in many windows. To see them, place your mouse pointer over the blue bar under the window's title bar. The menus in the JMP Starter window, the Home Window, and all data tables are always visible.

Using Sample Data

The examples in this book and the other JMP books use sample data tables. The default location on Windows for the sample data is:

C:/Program Files/SAS/JMP/14/Samples/Data

C:/Program Files/SAS/JMPPro/14/Samples/Data

C:/Program Files/SAS/JMPSW/14/Samples/Data

The Sample Data Index groups the data tables by category. Click a disclosure icon to see a list of data tables for that category, and then click a link to open a data table.

Macintosh sample data is installed in /Library/Application Support/JMP/14/Samples/Data.

Opening a JMP Sample Data Table

1. From the **Help** menu, select **Sample Data**.
2. Open the **Data Tables used in Discovering JMP** list by clicking on the disclosure icon next to it.
3. Click the name of the data table to use it in the examples in this book.

Sample Import Data

Use files from other applications to learn how to import data into JMP.

The default location on Windows for the sample import data is:

C:/Program Files/SAS/JMP/14/Samples/Import Data

C:/Program Files/SAS/JMPPro/14/Samples/Import Data

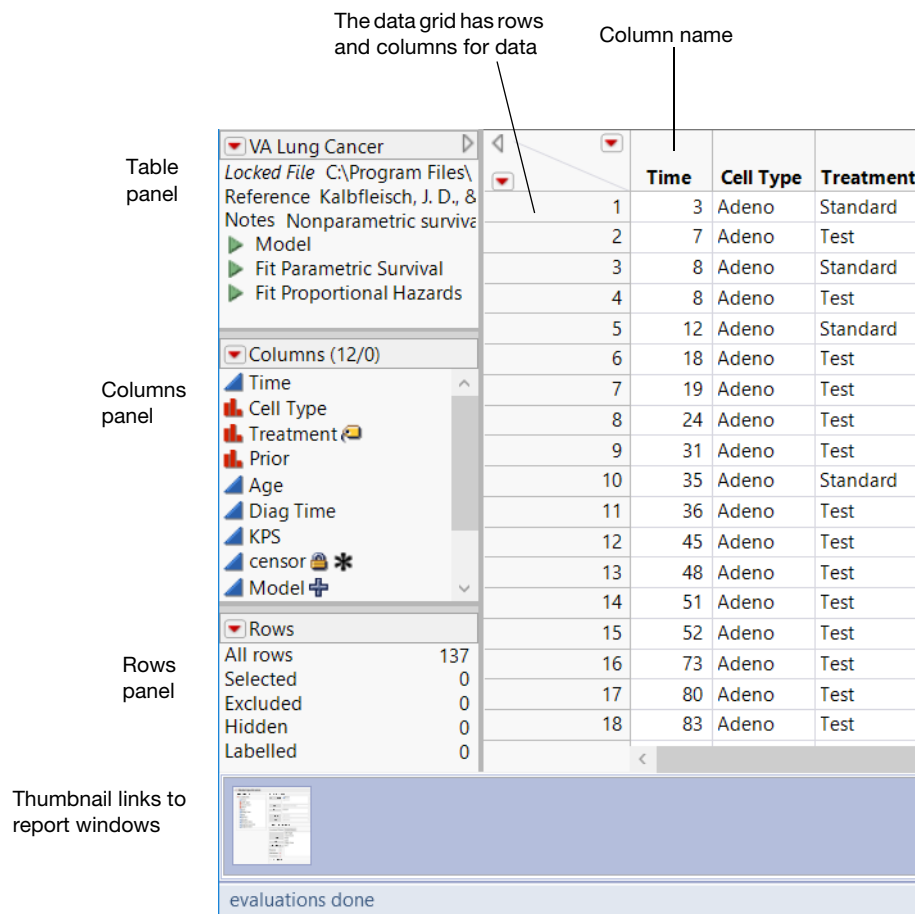
C:/Program Files/SAS/JMPSW/14/Samples/Import Data

Understanding Data Tables

A data table is a collection of data organized in rows and columns. It is similar to a Microsoft[®] Excel[®] spreadsheet, but with some important differences that are discussed in [“How is JMP Different from Excel?”](#) on page 63. A data table might also contain other information like notes, variables, and scripts. These supplementary items are discussed in later chapters.

Open the VA Lung Cancer data table to see the data table described here.

Figure 2.4 A Data Table



A data table contains the following parts:

Data grid The data grid contains the data arranged in rows and columns. Generally, each row in the data grid is an observation, and the columns (also called variables) give information about the observations. In Figure 2.4, each row corresponds to a test subject, and there are twelve columns of information. Although all twelve columns cannot be shown in the data grid, the Columns panel lists them all. The information given about each test subject includes the time, cell type, treatment, and more. Each column has a header, or name. That name is not part of the table's total count of rows.

Table panel The table panel can contain table variables or table scripts. In Figure 2.4, there is one saved script called **Model** that can automatically re-create an analysis. This table also

has a variable named Notes that contains information about the data. Table variables and table scripts are discussed in a later chapter.

Columns panel The columns panel shows the total number of columns, whether any columns are selected, and a list of all the columns by name. The numbers in parentheses (12/0) show that there are twelve columns, and that no columns are selected. An icon to the left of each column name shows that column's modeling type. Modeling types are described in [“Understand Modeling Types”](#) on page 136 in the [“Analyze Your Data”](#) chapter. Icons to the right show any attributes assigned to the column. See [“View or Change Column Information”](#) on page 79 in the [“Work with Your Data”](#) chapter for more information about these icons.

Rows panel The rows panel shows the number of rows in the data table, and how many rows are selected, excluded, hidden, or labeled. In Figure 2.4, there are 137 rows in the data table.

Thumbnail links to report windows This area shows thumbnails of all reports based on the data table. Place your mouse pointer over a thumbnail to see a larger preview of the report window. Double-click a thumbnail to bring the report window to the front.

Interacting with the data grid, which includes adding rows and columns, entering data, and editing data, is discussed in the [“Work with Your Data”](#) chapter on page 65. If you open multiple data tables, each one appears in a separate window.

Understanding the JMP Workflow

Once your data is in a data table, you can create graphs or plots, and perform analyses. All features are located in platforms, which are found primarily on the **Analyze** or **Graph** menus. They are called platforms because they do not just produce simple static results. Platform results appear in report windows, are highly interactive, and are linked to the data table and to each other.

The platforms under the **Analyze** and **Graph** menus provide a variety of analytical features and data exploration tools.

The general steps to produce a graph or analysis are as follows:

1. Open a data table.
2. Select a platform from the Graph or Analyze menu.
3. Complete the platform launch window to set up your analysis.
4. Click **OK** to create the report window that contains your graphs and statistical analyses.
5. Customize your report by using report options.
6. Save, export, and share your results with others.

Later chapters discuss these concepts in greater detail.

The following example shows you how to perform a simple analysis and customize it in four steps. This example uses the Companies.jmp file sample data table to show a basic analysis of the variable Profits (\$M).

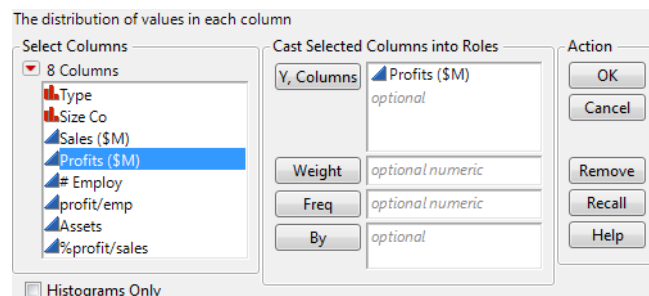
Step 1: Launching a Platform and Viewing Results

1. Select **Help > Sample Data Library** and open Companies.jmp.
2. Select **Analyze > Distribution** to open the Distribution launch window.
3. Select Profits (\$M) in the Select Columns box and click the **Y, Columns** button.

The variable Profits (\$M) appears in the **Y, Columns** role. See Figure 2.5 for the completed window.

Another way to assign variables is to click and drag columns from the Select Columns box to any of the roles boxes.

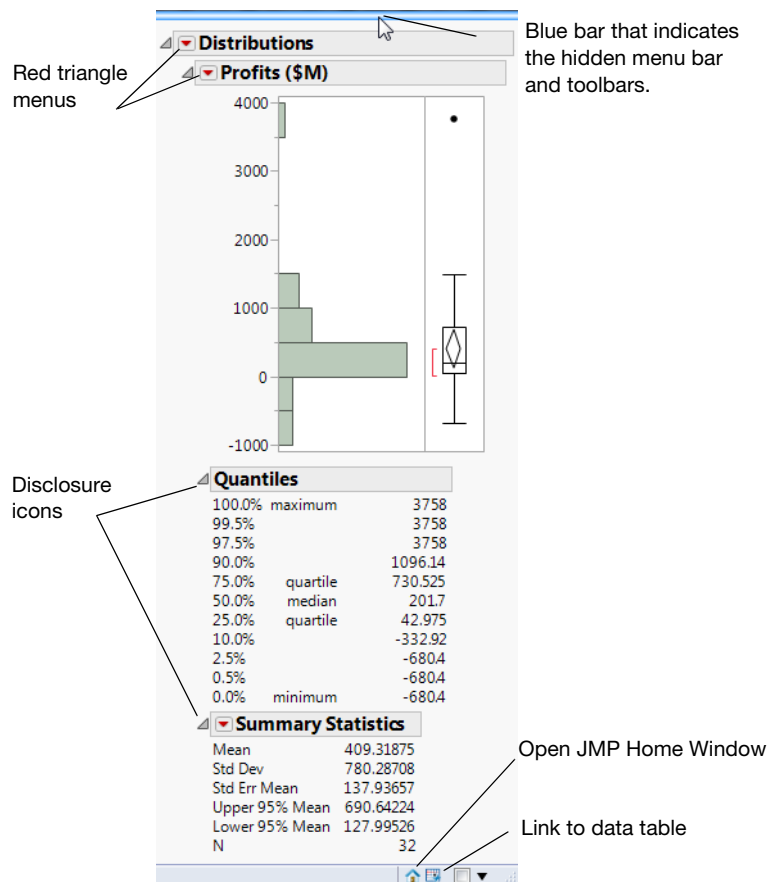
Figure 2.5 Assign Profits (\$M)



4. Click **OK**.

The Distribution report window appears.

Figure 2.6 Distribution Report Window on Windows



The report window contains basic plots or graphs and preliminary analysis reports. The results appear in an outline format, and you can show or hide any report by clicking on the disclosure icon.

Red triangle menus contain options and commands to request additional graphs and analyses at any time.

- On Windows, place your mouse pointer over the blue bar at the top of the window to see the menu bar and the toolbars.
- On Windows, click the data table button in the lower right corner to view the data table that was used to create this report. On Macintosh, click the **Show Data Table** button in the upper right corner of the report window.
- On Windows, click the **JMP Home Window** button in the lower right corner to view the Home Window. On Macintosh, select **Window > JMP Home**.

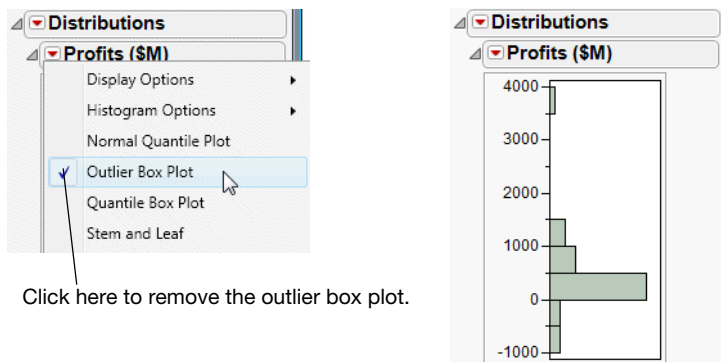
Step 2: Removing the Box Plot

Continue using the Distribution report that you created earlier.

1. Click the red triangle next to Profits (\$M) to see a menu of report options.
2. Deselect **Outlier Box Plot** to turn the option off.

The outlier box plot is removed from the report window.

Figure 2.7 Removing the Outlier Box Plot



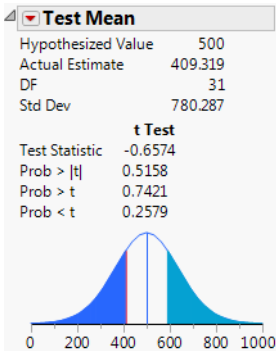
Step 3: Requesting Additional Output

Continue to use the same report window.

1. From the red triangle menu next to Profits (\$M), select **Test Mean**.
The Test Mean window appears.
2. Enter 500 in the **Specify Hypothesized Mean** box.
3. Click **OK**.

The test for the mean is added to the report window.

Figure 2.8 Test for the Mean



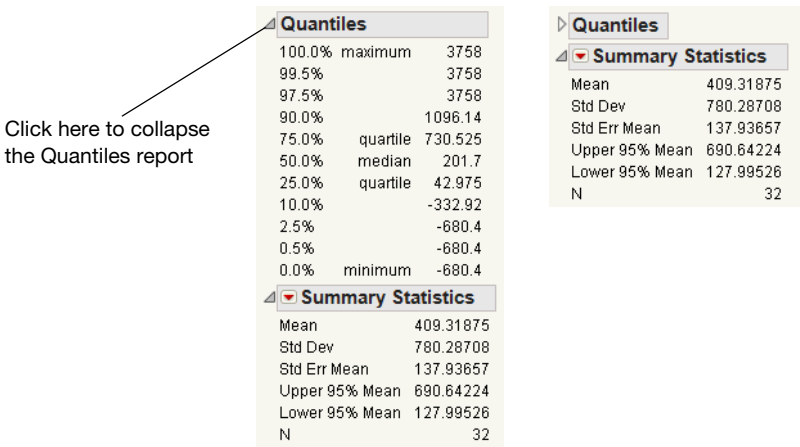
Step 4: Interacting with Platform Results

All platforms produce results that are interactive, for example, the following results:

- Reports can be shown or hidden.
- Additional graphs and statistical details can be added or removed to suit your purposes.
- Platform results are connected to the data table and to each other.

For example, to close the **Quantiles** report, click the disclosure icon next to **Quantiles**.

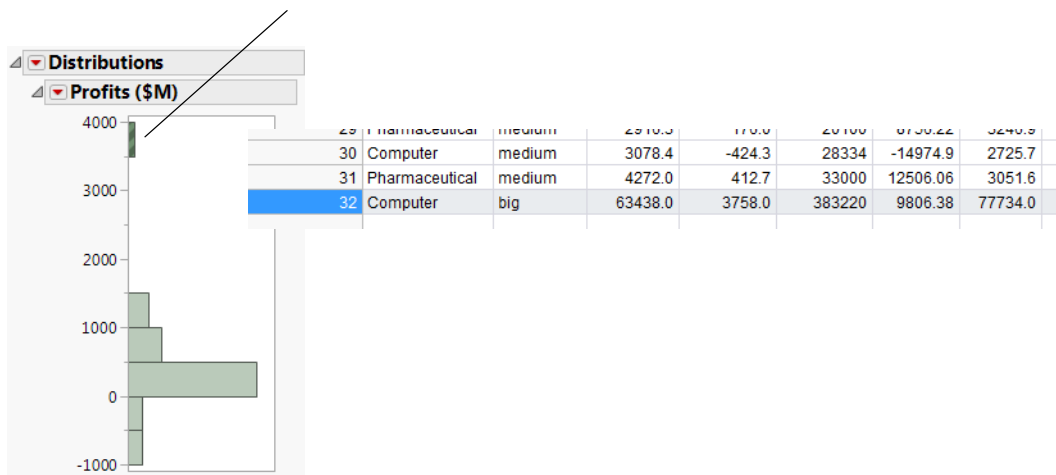
Figure 2.9 Close the Quantiles Report



Platform results are connected to the data table. The histogram in Figure 2.10 shows that a group of companies makes a much higher profit than the others. To quickly identify that group, click on the histogram bar for them. The corresponding rows in the data table are selected.

Figure 2.10 Connection between Platform Results and Data Table

Click the bar to select the corresponding rows



In this case, the group includes only one company, and that one row is selected.

How is JMP Different from Excel?

There are a number of important differences between JMP and Excel or other spreadsheet applications.

Table 2.1 JMP and Excel Differences

Formulas	
Excel	Formulas are applied to individual cells.
JMP	Formulas are applied only to entire columns. “Calculate Values with Formulas” on page 81 in the “Work with Your Data” chapter describes how to use formulas.
Column Names	
Excel	Column names are part of the grid. Numbered rows and labeled columns extend past the data. Numeric and character data reside in the same column.

Table 2.1 JMP and Excel Differences *(Continued)*

JMP	Column names are not part of the grid. There are no rows and columns beyond the existing data. The grid is only as big as the data. A column is either numeric or character. If a column contains both character and numeric data, the entire column’s data type is character, and the numbers are treated as character data. “ Understand Modeling Types ” on page 136 in the “Analyze Your Data” chapter describes how data type influences platform results.
Tables and Worksheets	
Excel	A single spreadsheet contains several tables, or worksheets.
JMP	JMP does not have the concept of worksheets. Each data table is a separate .jmp file and appears in a separate window.
Data Grid	
Excel	Data can be located anywhere in the data grid.
JMP	Data always begins in row 1 and column 1.
Analysis and Graph Reports	
Excel	All data, analyses, and graphs are placed inside the data grid.
JMP	Results appear in a separate window.

Chapter 3

Work with Your Data

Prepare Your Data for Graphing and Analyzing

Before graphing or analyzing your data, the data has to be in a data table and in the proper format. This chapter shows some basic data management tasks, including the following:

- Creating new data tables
- Opening existing data tables
- Importing data from other applications into JMP
- Managing your data

Figure 3.1 Example of a Data Table

Companies								
Locked File C:\Program Files\S...								
Notes Selected Data on the Fort								
Columns (8/0)								
Type								
Size Co								
Sales (\$M)								
Profits (\$M)								
# Employ								
profit/emp								
Assets								
%profit/sales								
Rows								
All rows	32							
Selected	0							
Excluded	0							
Hidden	0							
Labelled	0							

	Type	Size Co	Sales (\$M)	Profits (\$M)	# Employ	profit/emp	Assets
1	Computer	small	855.1	31.0	7523	4120.70	615.2
2	Pharmaceutical	big	5453.5	859.8	40929	21007.11	4851.6
3	Computer	small	2153.7	153.0	8200	18658.54	2233.7
4	Pharmaceutical	big	6747.0	1102.2	50816	21690.02	5681.5
5	Computer	small	5284.0	454.0	12068	37620.15	2743.9
6	Pharmaceutical	big	9422.0	747.0	54100	13807.76	8497.0
7	Computer	small	2876.1	333.3	9500	35084.21	2090.4
8	Computer	small	709.3	41.4	5000	8280.00	468.1
9	Computer	small	2952.1	-680.4	18000	-37800.0	1860.7
10	Computer	small	784.7	89.0	4708	18903.99	955.8
11	Computer	small	1324.3	-119.7	13740	-8711.79	1040.2
12	Pharmaceutical	medium	4175.6	939.5	28200	33315.60	5848.0
13	Computer	big	11899.0	829.0	95000	8726.32	10075.0
14	Computer	small	873.6	79.5	8200	9695.12	808.0
15	Pharmaceutical	big	9844.0	1082.0	83100	13020.46	7919.0
16	Pharmaceutical	small	969.2	227.4	3418	66530.13	784.0
17	Pharmaceutical	medium	6698.4	1495.4	34400	43470.93	6756.7
18	Computer	big	5956.0	412.0	56000	7357.14	4500.0
19	Pharmaceutical	big	5903.7	681.1	42100	16178.15	8324.8
20	Computer	medium	2959.3	252.8	31404	8049.93	5611.1
21	Pharmaceutical	small	1198.3	86.5	8527	10144.25	1791.7

Contents

Get Your Data into JMP	67
Copy and Paste Data.....	67
Import Data	67
Enter Data.....	70
Transfer Data from Excel	72
Work with Data Tables.....	73
Edit Data.....	74
Select, Deselect, and Find Values	75
View or Change Column Information	79
Calculate Values with Formulas	81
Filter Data.....	83
Manage Data	84
View Summary Statistics	85
Create Subsets	89
Join Data Tables.....	91
Sort Tables	93

Get Your Data into JMP

- To copy and paste data from another application, see “[Copy and Paste Data](#)” on page 67.
- To import data from another application, see “[Import Data](#)” on page 67.
- To enter data directly into a data table, see “[Enter Data](#)” on page 70
- To open a data table, double-click on the file, or use the **File > Open** command.

You can also import data into JMP from a database. For more information, see the Import Your Data chapter in the *Using JMP* book.

This chapter uses sample data tables and sample import data that is installed with JMP. To find these files, see “[Using Sample Data](#)” on page 56 in the “Introducing JMP” chapter.

Copy and Paste Data

You can move data into JMP by copying and pasting from another application, such as Excel or a text file.

1. Open the VA Lung Cancer.xls file in Excel. This file is located in the Sample Import Data folder.
2. Select all of the rows and columns, including the column names. There are 12 columns and 138 rows.
3. Copy the selected data.
4. In JMP, select **File > New > Data Table** to create an empty table.
5. Select **Edit > Paste with Column Names** to paste the data and column headings.

If the data that you are pasting into JMP does *not* have column names, then you can use **Edit > Paste**.

Import Data

You can move data into JMP by importing data from another application, such as Excel, SAS, or text files. The basic steps to import data are as follows:

1. Select **File > Open**.
2. Navigate to your file’s location.
3. If your file is not listed in the Open Data File window, select the correct file type from the **Files of type** menu.
4. Click **Open**.

Example of Importing a Microsoft Excel File

1. Select **File > Open**.
2. Navigate to the Samples/Import Data folder.
3. Select **Team Results.xls**.

Note the rows and columns on which the data begin. The spreadsheet also contains two worksheets. In this example, you import the Ungrouped Team Results worksheet.

4. Click **Open**.

The spreadsheet opens in the Excel Import Wizard, where a preview of the data appears along with import options.

Text from the first row of the spreadsheet are column headings. However, you want text in row 3 of the spreadsheet to be converted to column headings.

5. Next to **Column headers start on row**, type 3, and press **Enter**. The column headings are updated in the data preview. The value for the first row of data is updated to 4.
6. Save the settings only for this worksheet:
 - Deselect **Use for all worksheets** in the lower left corner of the window.
 - Select **Ungrouped Team Results** in the upper right corner of the window.
7. Click **Import** to convert the spreadsheet as you specified.

When you import Excel files, JMP predicts whether columns headings exist, and if the column names are on row one. The copy and paste method is recommended for the following situations:

- If the column names are located in a row other than row one
- If the file does not include column names and the data does not start in row one
- If the file contains column names and the data does not start in row two

See [“Copy and Paste Data”](#) on page 67 and the Import Your Data chapter in the *Using JMP* book for more information about importing Excel files.

Example of Importing a Text File

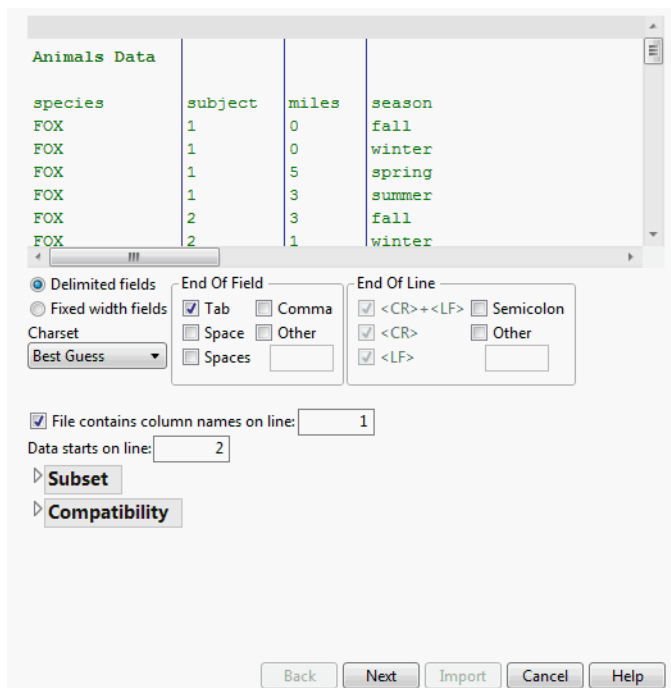
One way to import a text file is to let JMP assume the data's format and place the data in a data table. This method uses settings that you can specify in Preferences. See the JMP Preferences chapter in the *Using JMP* book for information about setting text import preferences.

Another way to import a text file is to use a Text Preview window to see what your data table will look like after importing, and make adjustments. The following example shows you how to use Text Import Preview window.

1. Select **File > Open**.
2. Navigate to the Samples/Import Data folder.

3. Select `Animals_line3.txt`.
4. At the bottom of the Open window, select **Data with Preview**.
5. Click **Open**.

Figure 3.2 Initial Preview Window



This text file has a title on the first line, column names on the third line, and the data starts on line four. If you opened this directly in JMP, the `Animals Data` line would be the first column name, and all the column names and data afterward would be out of sync. The Preview window lets you adjust the settings before you open the file, and see how your adjustments affect the final data table.

6. Enter 3 in the **File contains column names on line** field.
7. Enter 4 in the **Data starts on line** field.
8. Click **Next**.

In the second window, you can exclude columns from the import and change the data modeling of the columns. For this example, use the default settings.

9. Click **Import**.

The new data table has columns named `species`, `subject`, `miles`, and `season`. The `species` and `season` columns are character data. The `subject` and `miles` columns are continuous numeric data.

Tip: You can import several text files at once to create a data table. See the Import Your Data chapter in the *Using JMP* book for details.

Enter Data

You can enter data directly in a data table. The following example shows you how to enter data that was collected over several months into a data table.

Scenario

Table 3.1 shows the data from a study that investigated a new blood pressure medication. Each individual’s blood pressure was measured over a six-month period. Two doses (300mg and 450mg) of the medication were used, along with a control and placebo group. The data shows the average blood pressure for each group.

Table 3.1 Blood Pressure Data

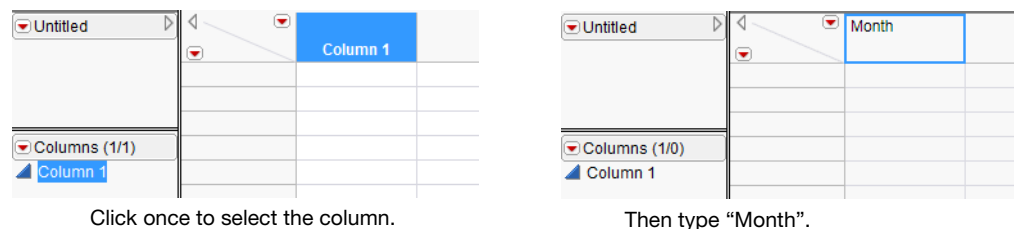
Month	Control	Placebo	300mg	450mg
March	165	163	166	168
April	162	159	165	163
May	164	158	161	153
June	162	161	158	151
July	166	158	160	148
August	163	158	157	150

Enter Data in a New Data Table

1. Select **File > New > Data Table** to create an empty data table.
A new data table has one column and no rows.
2. Select the column name and change the name to **Month**. See Figure 3.3.

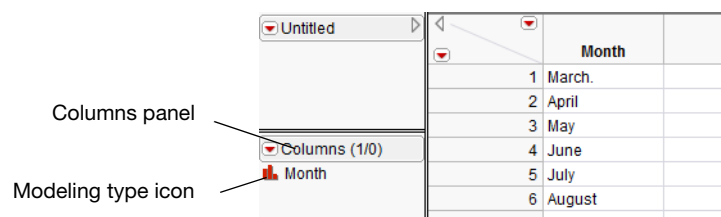
Note: To rename a column, you can also double-click the column name or select the column and press Enter.

Figure 3.3 Entering a Column Name



3. Select **Rows > Add Rows**.
The Add Rows window appears.
4. Since you want to add six rows, type 6.
5. Click **OK**. Six empty rows are added to the data table.
6. Enter the Month information by clicking in a cell and typing.

Figure 3.4 Month Column Completed



In the columns panel, look at the modeling type icon to the left of the column name. It has changed to reflect that Month is now nominal (previously it was continuous). Compare the modeling type shown for Column 1 in Figure 3.3 and for Month in Figure 3.4. This difference is important and is discussed in [“View or Change Column Information”](#) on page 79.

7. Double-click in the space on the right side of the Month column to add the Control column.
8. Change the name to Control.
9. Enter the Control data as shown in Table 3.1. Your data table now consists of six rows and two columns.
10. Continue adding columns and entering data as shown in Table 3.1 to create the final data table with six rows and five columns.

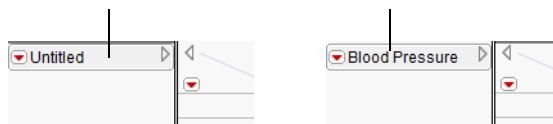
Change the Data Table Name

1. Double-click on the data table name (Untitled) in the Table Panel.
2. Type the new name (Blood Pressure).

Figure 3.5 Changing the Data Table Name

Double-click here.

Type the new name.



Transfer Data from Excel

You can use the JMP Add In for Excel to transfer a spreadsheet from Excel to JMP:

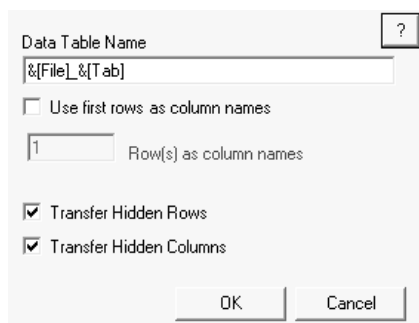
- a data table
- Graph Builder
- Distribution platform
- Fit Y by X platform
- Fit Model platform
- Time Series platform
- Control Chart platform

Set JMP Add In Preferences in Excel

To configure JMP Add In Preferences:

1. In Excel, select **JMP > Preferences**.
The JMP Preferences window appears.

Figure 3.6 JMP Add In Preferences



2. Accept the default **Data Table Name** (File name_Worksheet name) or type a name.
3. Select to **Use the first rows as column names** if the first row in the worksheet contains column headers.

4. If you selected to use the first rows as column headers, type the number of rows used.
5. Select to **Transfer Hidden Rows** if the worksheet contains hidden rows to be included in the JMP data table.
6. Select to **Transfer Hidden Column** if the worksheet contains hidden columns to be included in the JMP data table.
7. Click **OK** to save your preferences.

Transfer to JMP

To transfer an Excel worksheet to JMP:

1. Open the Excel file.
2. Select the worksheet to transfer.
3. Select **JMP** and then select the JMP destination:
 - Data Table
 - Graph Builder
 - Distribution platform
 - Fit Y by X platform
 - Fit Model platform
 - Time Series platform
 - Control Chart platform

The Excel worksheet is opened as a data table in JMP and the selected platform's launch window appears.

Work with Data Tables

This section contains the following information:

- [“Edit Data”](#)
- [“Select, Deselect, and Find Values”](#)
- [“View or Change Column Information”](#)
- [“Calculate Values with Formulas”](#)
- [“Filter Data”](#)

Tip: Consider setting the Autosave Time Out value in the General preferences to automatically save open data tables at the specified number of minutes. This autosave value also applies to journals, scripts, projects, and reports.

Edit Data

You can enter or change data, either a few cells at a time or for an entire column. This section contains the following information:

- [“Change Values”](#)
- [“Recode Values”](#)
- [“Create Patterned Data”](#)

Change Values

To change a value, select a cell and type the change. You can also double-click a cell to edit it.

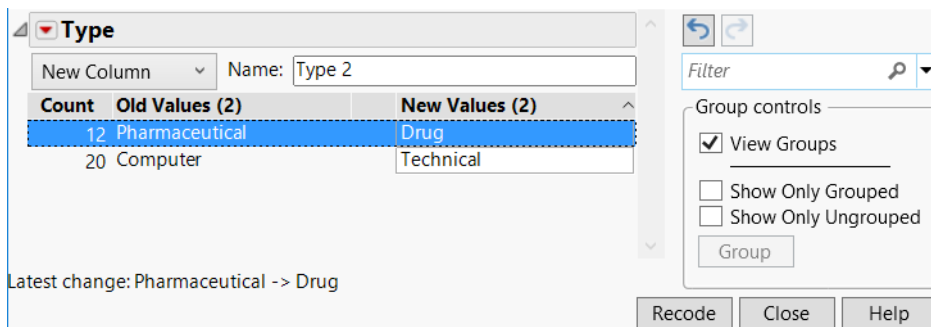
Note: Double-clicking in a cell is not the same as selecting a cell. A single click selects a cell. You can select more than one cell at the same time, and you can perform certain actions on selected cells. Double-clicking only lets you edit a cell. For more information about selecting rows, columns, and cells, see [“Select, Deselect, and Find Values”](#) on page 75.

Recode Values

Use the recoding tool to change all of the values in a column at once. For example, suppose that you are interested in comparing the sales of computer and pharmaceutical companies. Your current company labels are Computer and Pharmaceutical. You want to change them to Technical and Drug. Going through all 32 rows of data and changing all the values would be tedious, inefficient, and error-prone, especially if you had many more rows of data. Recode is a better option.

1. Select **Help > Sample Data Library** and open **Companies.jmp**.
2. Select the **Type** column by clicking once on the column heading.
3. Select **Cols > Recode**.
4. In the **New Value** column of the **Recode** window, type **Technical** in the **Computer** row and **Drug** in the **Pharmaceutical** row.
5. Click **Done** and select the **In Place** option from the list.

Figure 3.7 Recode Window



All cells are updated automatically to the new values.

Create Patterned Data

Use the Fill options to populate a column with patterned data. The Fill options are especially useful if your data table is large, and typing in the values for each row would be cumbersome.

Example of Filling a Column with the Pattern

1. Add a new column.
2. Enter 1 in the first cell, 2 in the second cell, and 3 in the third cell.
3. Select the three cells, and right-click anywhere in the selected cells to see a menu.
4. Select **Fill > Repeat sequence to end of table**.

The rest of the column is filled with the sequence (1, 2, 3, 1, 2, 3, ...).

To continue a pattern instead of repeating it (1, 2, 3, 4, 5, 6, ...), select **Continue sequence to end of table**. This command can also be used to generate patterns like (1, 1, 1, 2, 2, 2, 3, 3, 3, ...).

The Fill options can recognize simple arithmetic and geometric sequences. For character data, the Fill options only repeat the values.

Select, Deselect, and Find Values

You can select rows, columns, or cells within a data table. For example, to create a subset of an existing data table, you must first select the parts of the table that you want to subset. Also, selecting rows can make data points stand out on a graph. Select rows and columns manually by clicking, or select rows that meet certain search criteria. This section contains the following information:

- “[Select and Deselect Rows](#)” on page 76
- “[Select and Deselect Columns](#)” on page 76

- “Select and Deselect Cells” on page 77
- “Search for Values” on page 78

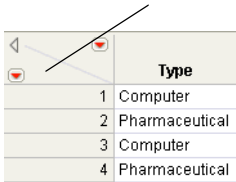
Select and Deselect Rows

Table 3.2 Selecting and Deselecting Rows

Task	Action
Select rows one at a time	Click on the row number.
Select multiple adjacent rows	Click and drag on the row numbers. or Select the beginning row, and then hold down the Shift key and click the last row number.
Select multiple non-adjacent rows	Select the first row, and then hold down the Ctrl key and click the other row numbers.
Deselect rows one at a time	Hold down the Ctrl key and click the row numbers.
Deselect all rows	Click in the lower-triangular space in the top left corner of the table. See Figure 3.8.

Figure 3.8 Deselecting Rows

To deselect all rows at once, click here.



Select and Deselect Columns

Table 3.3 Selecting and Deselecting Columns

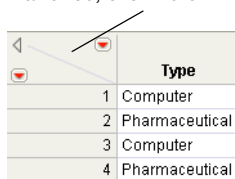
Task	Action
Select columns one at a time	Click the column heading.

Table 3.3 Selecting and Deselecting Columns (*Continued*)

Task	Action
Select multiple adjacent columns	Click and drag across the column headings. or Select the beginning column, and then hold down the Shift key and click the last header.
Select multiple non-adjacent columns	Select the first column, and then hold down the Ctrl key and click the other column headings.
Deselect columns one at a time	Hold down the Ctrl key and click the column heading.
Deselect all columns	Click in the upper-triangular space in the top left corner of the table. See Figure 3.9.

Figure 3.9 Deselecting Columns

To deselect all columns at once, click here.



Select and Deselect Cells

Table 3.4 Selecting and Deselecting Cells

Task	Action
Select cells one at a time	Click each cell individually.
Select multiple adjacent cells	Click and drag across the cells. or Select the beginning cell, and then hold down the Shift key and click the last cell.
Select multiple non-adjacent cells	Select the first cell, and then hold down the Ctrl key and click the other cells.

Table 3.4 Selecting and Deselecting Cells *(Continued)*

Task	Action
Deselect all cells	Click in the upper and lower triangular spaces in the top left corner of the table.

Search for Values

In a data table that has thousands or tens of thousands of rows, it can be difficult to locate a particular cell by scrolling through the table. If you are looking for specific information, use the Search feature to find it. If data match the search criteria, the cell is selected and the data grid scrolls to show it in the window. For example, the *Companies.jmp* data table contains information about a company that has total sales of \$11,899. Use the Search feature to find that cell.

Example of Searching for a Value

1. Select **Edit > Search > Find** to launch the Search window.
2. In the **Find what** box, enter 11899.
3. Click **Find**. JMP finds the first cell that has 11,899 in it, and selects it.

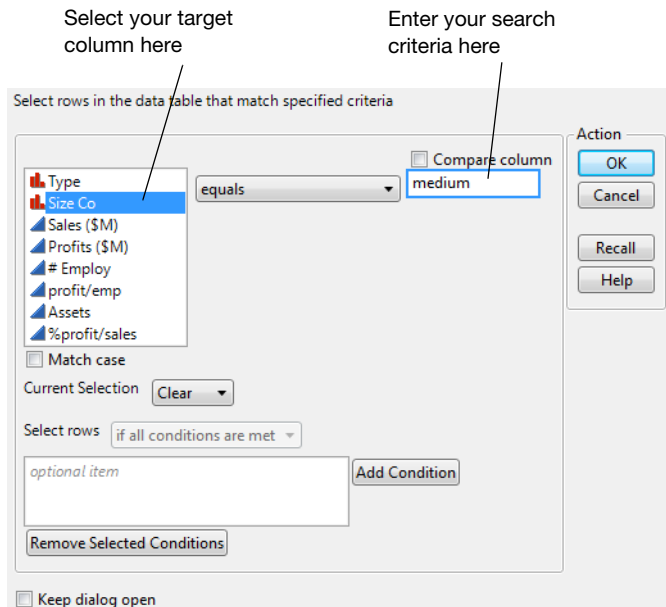
If multiple cells meet the search criteria, click **Find** again to find the next cell that matches the search term.

You can also search for multiple rows at once, with each row matching some criteria.

Example of Select All Rows That Correspond to Medium-Sized Companies

1. Select **Rows > Row Selection > Select Where** to open the **Select rows** window.
2. In the column list box on the left, select *Size Co.*
3. In the text box on the right, enter *medium*.
4. Click **OK**.

Figure 3.10 Select Rows Window



JMP selects all of the rows that have Size Co equal to medium. There are seven.

View or Change Column Information

Information about a column is not limited to the data in the column. Data type, modeling type, format, and formulas can also be set.

To view or change column characteristics, double-click the column heading. Or, right-click the column heading and select **Column Info**. The Column Info window appears.

Figure 3.11 Column Info Window

'%profit/sales' in table 'Companies'

Column Name

☒ Lock

Data Type

Modeling Type

Format Width Dec

☐ Use thousands separator (,)

Column Properties
optional item

Formula

☐ Suppress Eval
☐ Ignore Errors

$$\left(\frac{\text{Profits (\$M)}}{\text{Sales (\$M)}} \right) \cdot 100$$

Column Name Enter or change the column name. No two columns can have the same column name.

Data Type Select one of the following data types:

- **Numeric** specifies the column values as numbers.
- **Character** specifies the column values as non-numeric, such as letters or symbols.
- **Row State** specifies the column values as row states. This is an advanced topic. See The Column Info Window chapter in the *Using JMP* book.

Modeling Type Modeling types define how values are used in analyses. Select one of the following modeling types:

- **Continuous** values are numeric only.
- **Ordinal** values are either numeric or character, and are ordered categories.
- **Nominal** values are either numeric or character, but not ordered.

Format Select a format for numeric values. This option is not available for character data. Here are a few of the most common formats:

- **Best** lets JMP choose the best display format.
- **Fixed Dec** specifies the number of decimal places that appear.
- **Date** specifies the syntax for date values.

- **Time** specifies the syntax for time values.
- **Currency** specifies the type of currency and decimal points that are used for currency values.

Column Properties Set special column properties such as formulas, notes, and value orders. See The Column Info Window chapter in the *Using JMP* book.

Lock Lock a column, so that the values in the column cannot be changed.

Calculate Values with Formulas

Use the Formula Editor to create columns that contain calculated values.

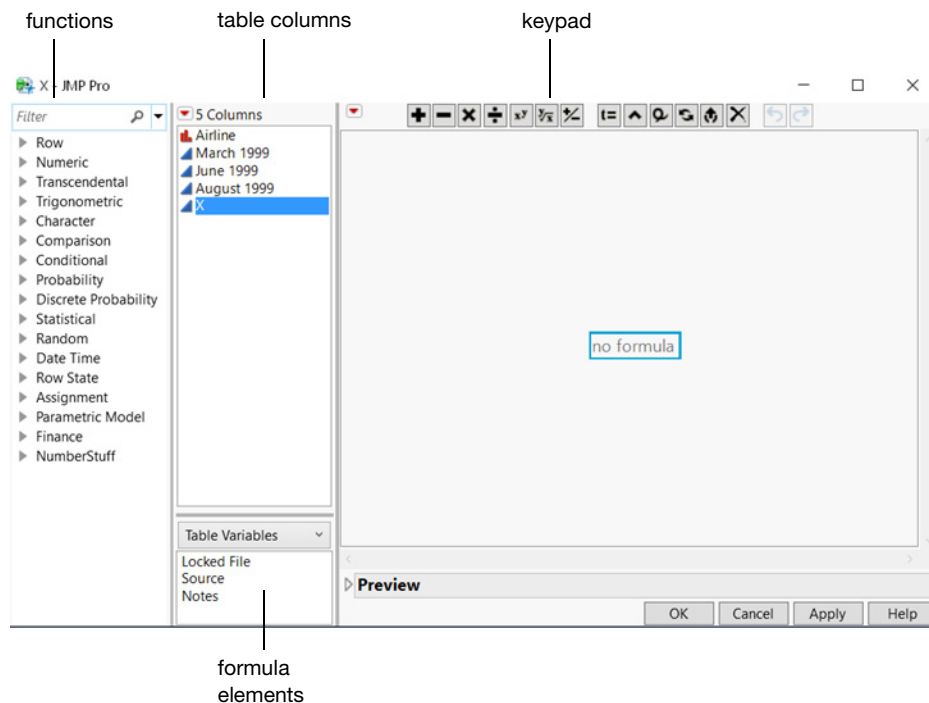
Scenario

The sample data table *On-Time Arrivals.jmp* reflects the percent of on-time arrivals for several airlines. The data was collected for March, June, and August of 1999.

Create the Formula

Suppose that you want to create a new column containing the average on-time percentage for each airline.

1. Add a new column.
2. Right-click the column heading of the new column and select **Formula**. The Formula Editor window appears.

Figure 3.12 Formula Editor

Create the formula for the average on-time percentage of each airline:

3. From the Columns list, select March 1999.
4. Click the $+$ button on the keypad.
5. Select June 1999, followed by another $+$ sign.
6. Select August 1999.

Figure 3.13 Sum of the Months

March 1999 + *June 1999* + *August 1999*

Notice that only August 1999 is selected (has the blue box around it).

7. Click on the box surrounding the entire formula.

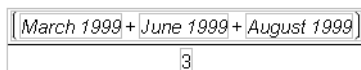
Figure 3.14 Entire Formula Selected

March 1999 + *June 1999* + *August 1999*

8. Click the \div button.

9. Type a 3 in the denominator box, and then click outside of the formula in any of the white space.

Figure 3.15 Completed Formula



10. Click **OK**

The new column contains the averages.

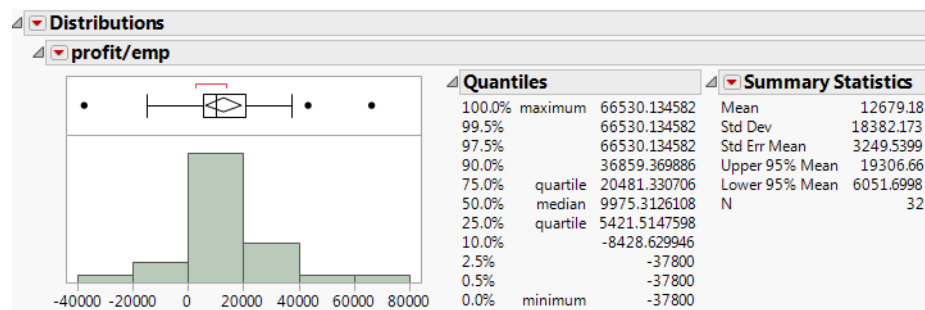
The Formula Editor has many built-in arithmetic and statistical functions. For example, another way to calculate the average on-time arrival percentage is to use the Mean function in the Statistical functions list. For details about all of the Formula Editor functions, see the Formula Editor chapter in the *Using JMP* book.

Filter Data

Use the **Data Filter** to interactively select complex subsets of data, hide these subsets in plots, or exclude them from analyses. For example, look at profit per employee for computer and pharmaceutical companies.

1. Select **Help > Sample Data Library** and open *Companies.jmp*.
2. Select **Analyze > Distribution**.
3. Select profit/emp and click **Y, Columns**.
4. Click **OK**.
5. From the red triangle menu for profit/emp, select **Display Options > Horizontal Layout**.

Figure 3.16 Distribution of profit/emp



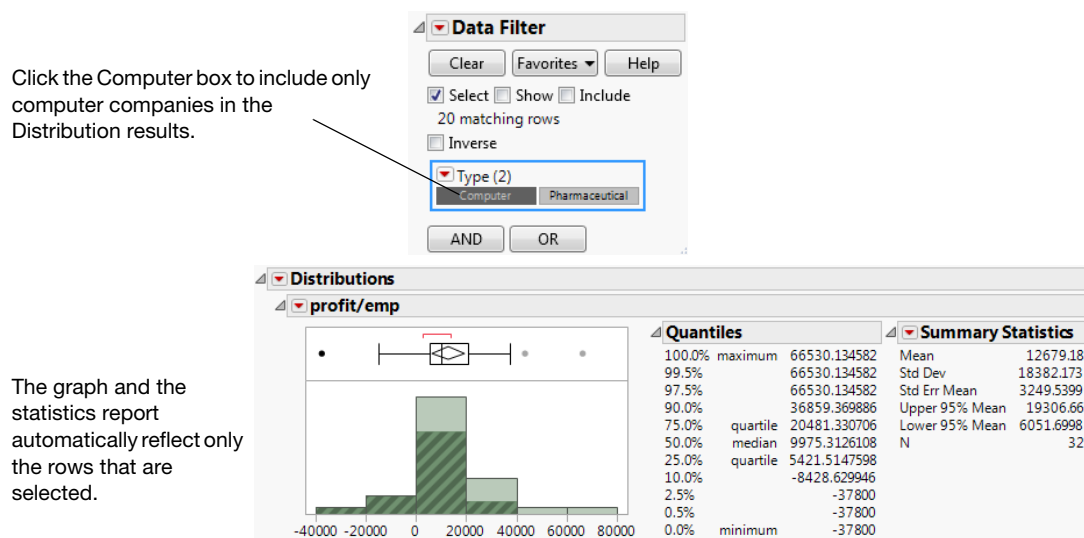
6. Turn on Automatic Recalc by selecting **Redo > Automatic Recalc** from the red triangle menu for **Distributions**.

When this option is on, every change that you make (for example, hiding or excluding points) causes your report window to automatically update itself.

7. In the data table, select **Rows > Data Filter**.
8. Select **Type** and click **Add**.
9. Make sure that **Select** is selected.
10. To filter out the Pharmaceutical companies from the Distribution results, and include only the Computer companies, click the **Computer** box in the Data Filter window.

The distribution results update to only include Computer companies.

Figure 3.17 Filter for Computer Companies



Conversely, to change the Distribution results to include only the Pharmaceutical companies, click the **Pharmaceutical** box on the Data Filter window.

Manage Data

The commands on the **Tables** menu (and **Tabulate** on the **Analyze** menu) summarize and manipulate data tables into the format that you need for graphing and analyzing. This section describes five of these commands:

Summary Creates a table that contains summary statistics that describe your data.

Tabulate Provides a drag and drop workspace to create summary statistics.

Subset Creates a table that contains a subset of your data.

Join Joins the data from two data tables into one new data table.

Sort Sorts your data by one or more columns.

For complete details about these and the other Tables menu commands, see the Reshape Data chapter in the *Using JMP* book.

View Summary Statistics

Summary statistics, such as sums and means, can instantly provide useful information about your data. For example, if you look at the annual profit of each company out of thirty-two companies, it's difficult to compare the profits of small, medium, and large companies. A summary shows that information immediately.

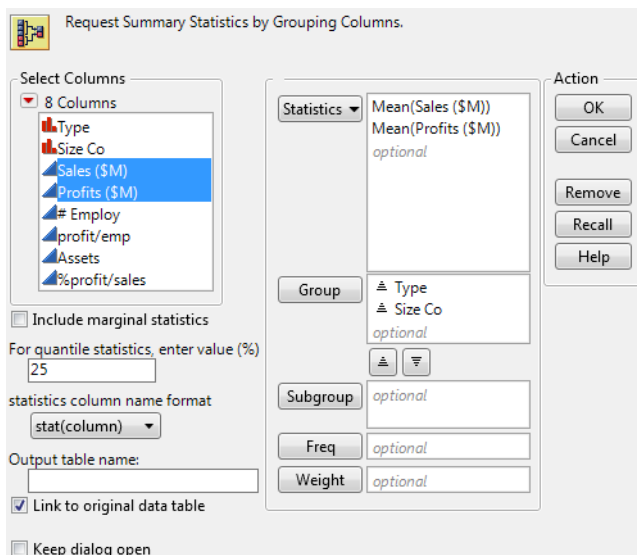
Create summary tables by using either the **Summary** or **Tabulate** commands. The **Summary** command creates a new data table. As with any data table, you can perform analyses and create graphs from the summary table. The **Tabulate** command creates a report window with a table of summary data. You can also create a table from the Tabulate report.

Summary

A summary table contains statistics for each level of a grouping variable. For example, look at the financial data for computer and pharmaceutical companies. Suppose that you want to calculate the mean of sales and the mean of profits, for each combination of company type and size.

1. Select **Help > Sample Data Library** and open *Companies.jmp*.
2. Select **Tables > Summary**.
3. Select **Type and Size Co** and click **Group**.
4. Select **Sales (\$M)** and **Profits (\$M)** and click **Statistics > Mean**.

Figure 3.18 Completed Summary Window



5. Click **OK**.

JMP calculates the mean of Sales (\$M) and the mean of Profit (\$M) for each combination of Type and Size Co.

Figure 3.19 Summary Table

Companies By (Type...		Type	Size Co	N Rows	Mean(Sales (\$M))	Mean(Profits (\$M))
Source	1	Computer	big	4	20597.48	1089.93
	2	Computer	medium	2	3018.85	-85.75
	3	Computer	small	14	1758.06	44.94
	4	Pharmaceutical	big	5	7474.04	894.42
	5	Pharmaceutical	medium	5	4261.06	698.98
	6	Pharmaceutical	small	2	1083.75	156.95
Columns (5/0)						
Rows		Type				
		Size Co				
		N Rows				
		Mean(Sales (\$M))				
		Mean(Profits (\$M))				
All rows		6				
Selected		0				
Excluded		0				
Hidden		0				
Labelled		0				

The summary table contains the following:

- There are columns for each grouping variable (in this example, Type, and Size Co).
- The N Rows column shows the number of rows from the original table that correspond to each combination of grouping variables. For example, the original data table contains 14 rows corresponding to small computer companies.

- There is a column for each summary statistic requested. In this example, there is a column for the mean of Sales (\$M) and a column for the mean of Profits (\$M).

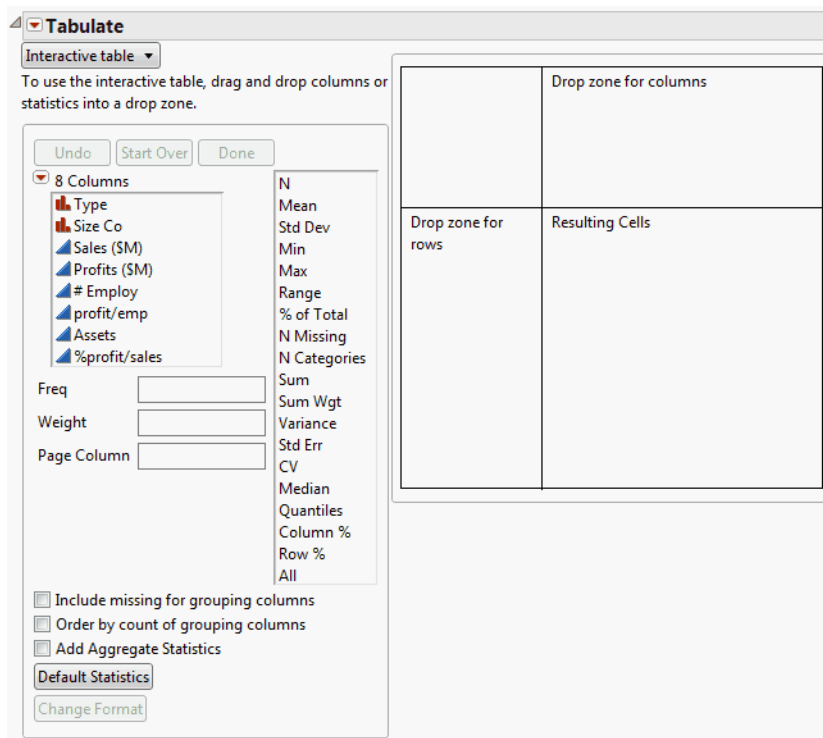
The summary table is linked to the source table. Selecting a row in the summary table also selects the corresponding rows in the source table.

Tabulate

Use the Tabulate command to drag columns into a workspace, creating summary statistics for each combination of grouping variables. This example shows you how to use Tabulate to create the same summary information that you just created using Summary.

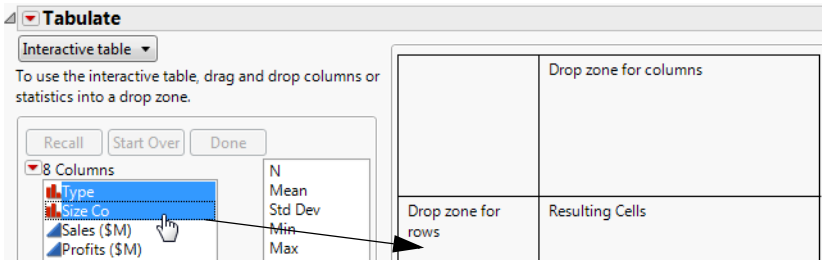
1. Select **Help > Sample Data Library** and open Companies.jmp.
2. Select **Analyze > Tabulate**.

Figure 3.20 Tabulate Workspace



3. Select both Type and Size Co.
4. Drag and drop them into the **Drop zone for rows**.

Figure 3.21 Dragging Columns to the Row Zone



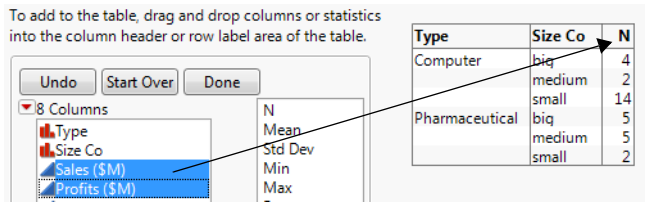
5. Right-click a heading and select **Nest Grouping Columns**.
The initial tabulation shows the number of rows per group.

Figure 3.22 Initial Tabulation

Type	Size Co	N
Computer	big	4
	medium	2
	small	14
Pharmaceutical	big	5
	medium	5
	small	2

6. Select both Sales (\$M) and Profits (\$M), and drag and drop them over the **N** in the table.

Figure 3.23 Adding Sales and Profit



The tabulation now shows the sum of Sales (\$M) and the sum of Profits (\$M) per group.

Figure 3.24 Tabulation of Sums

Type	Size Co	Sales (\$M)	Profits (\$M)
		Sum	Sum
Computer	big	82389.9	4359.7
	medium	6037.7	-171.5
	small	24612.8	629.1
Pharmaceutical	big	37370.2	4472.1
	medium	21305.3	3494.9
	small	2167.5	313.9

- The final step is to change the sums to means. Right-click **Sum** (either of them) and select **Statistics > Mean**.

Figure 3.25 Final Tabulation

		Sales (\$M)	Profits (\$M)
Type	Size Co	Mean	Mean
Computer	big	20597.48	1089.9
	medium	3018.85	-85.75
	small	1758.06	44.94
Pharmaceutical	big	7474.04	894.42
	medium	4261.06	698.98
	small	1083.75	156.95

The means are the same as those obtained using the Summary command. Compare Figure 3.25 to Figure 3.19.

Create Subsets

If you want to look closely at only part of your data table, you can create a subset. For example, suppose that you have already compared the sales and profits of big, medium, and small computer and pharmaceutical companies. Now you want to look at the sales and profits of only the medium-sized companies.

Creating a subset is a two-step process. First select the target data, and then extract the data into a new table.

Subset with the Subset Command

- Select **Help > Sample Data Library** and open Companies.jmp.

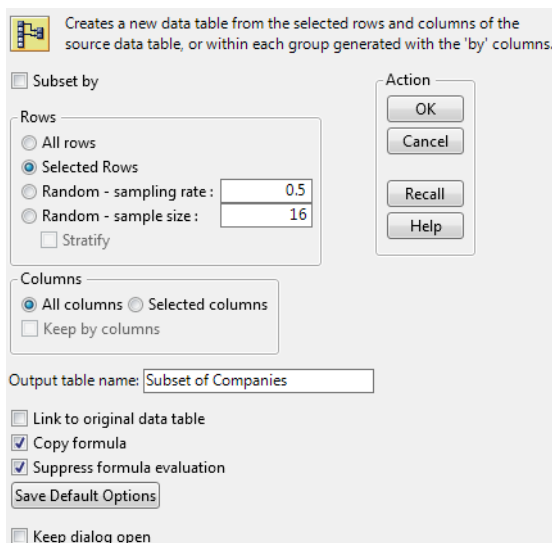
Selecting the Rows and Columns That You Want to Subset

- Select **Rows > Row Selection > Select Where**.
- Select Size Co in the column list box on the left.
- Enter medium in the text enter box.
- Click **OK**.
- Hold down the Ctrl key and select the Type, Sales (\$M), and Profits (\$M) columns.

Creating the Subset Table

- Select **Tables > Subset** to launch the Subset window.

Figure 3.26 Subset Window



8. Select **Selected columns** to subset only the columns that you selected. You can also customize your subset table further by selecting additional options.
9. Click **OK**.

The resulting subset data table has seven rows and three columns. For complete details about the Subset command, see the Reshape Data chapter in the *Using JMP* book.

Subset with the Distribution Platform

Another way to create subsets uses the connection between platform results and data tables.

Example of Creating a Subset Using the Distribution Command

1. Select **Help > Sample Data Library** and open *Companies.jmp*.
2. Select **Analyze > Distribution**.
3. Select **Type** and click **Y, Columns**.
4. Click **OK**.
5. Double-click on the histogram bar that represents **Computer** to create a subset table of the Computer companies.

Caution: This method creates a *linked* subset table. This means if you make any changes to the data in the subset table, the corresponding value changes in the source table.

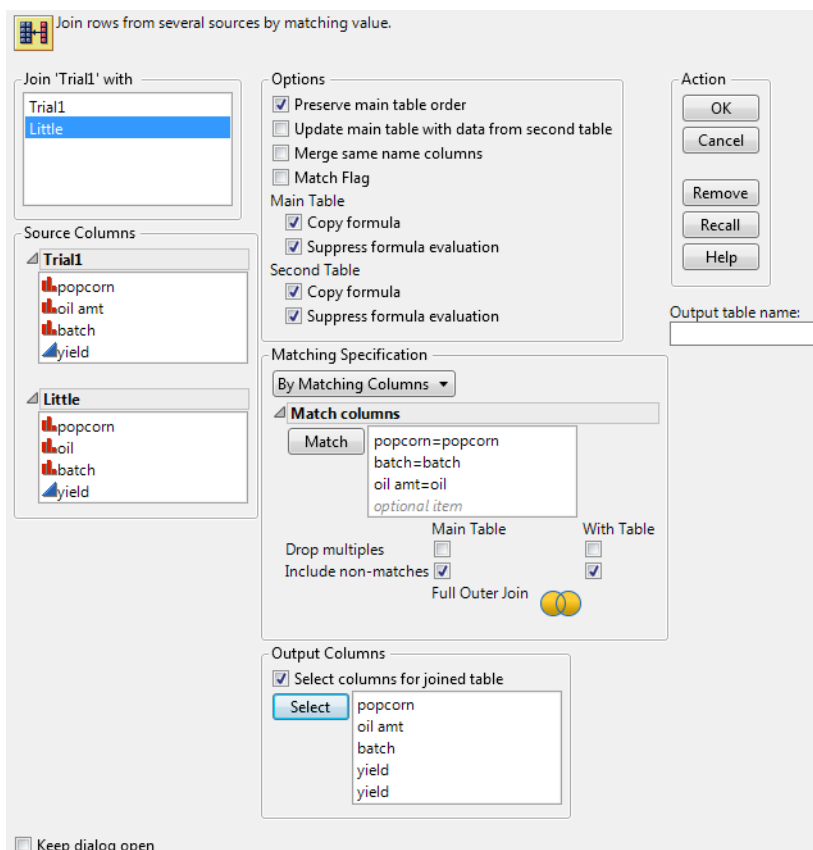
Join Data Tables

Use the Join option to combine information from multiple data tables into a single data table. For example, suppose that you have a data table containing results from an experiment on popcorn yields. In another data table, you have the results of a second experiment on popcorn yields. To compare the two experiments or to analyze the trials using both sets of results, you need to have the data in the same table. Also, the experimental data was not entered into the data tables in the same order. One of the columns has a different name, and the second experiment is incomplete. This means that you cannot copy and paste from one table into another.

Example of Joining Two Data Tables

1. Select **Help > Sample Data Library** and open Trial1.jmp and Little.jmp.
2. Click on Trial1.jmp to make it the active data table.
3. Select **Tables > Join**.
4. In the **Join 'Trial1' With** box, select Little.
5. From the **Matching Specification** menu, select **By Matching Columns**.
6. In the **Source Columns** boxes, select popcorn in both boxes, and then click **Match**.
7. In the same way, match batch and oil amt to oil in both boxes.
Your matching columns do not have to have the same name.
8. Select **Include non-matches** for both tables.
Since one experiment is partial, you want to include all rows, including any with missing data.
9. To avoid duplicate columns, select the **Select columns for joined table** option.
10. From Trial1, select all four columns and click **Select**.
11. From Little, select only yield and click **Select**.

Figure 3.27 Completed Join Window



12. Click OK.

Figure 3.28 Joined Table

▼ Untitled 3				▼	popcorn	oil amt	batch	yield of Trial1	yield of Little
Notes append to TRIA				▼	1	plain	little	large	8.2
Notes 2 Join with Trial					2	gourmet	little	large	8.6
▼ Source					3	plain	lots	large	10.4
					4	gourmet	lots	large	9.2
					5	plain	little	small	9.9
					6	gourmet	little	small	12.1
					7	plain	lots	small	10.6
					8	gourmet	lots	small	18.0

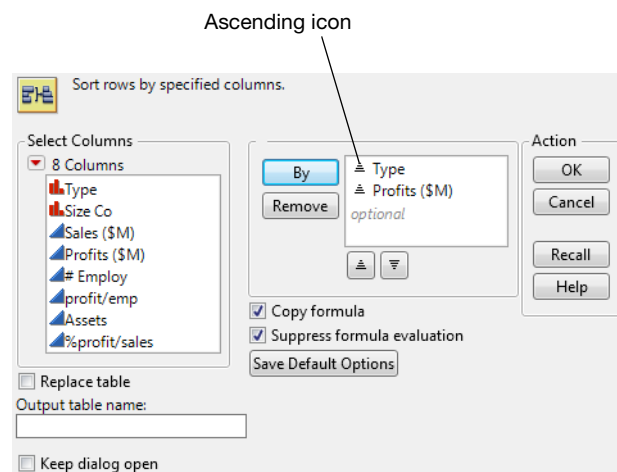
Sort Tables

Use the Sort command to sort a data table by one or more columns in the data table. For example, look at financial data for computer and pharmaceutical companies. Suppose that you want to sort the data table by Type, then by Profits (\$M). Also, you want Profits (\$M) to be in descending order within each Type.

1. Select **Help > Sample Data Library** and open Companies.jmp.
2. Select **Tables > Sort**.
3. Select Type and click **By** to assign Type as a sorting variable.
4. Select Profits (\$M) and click **By**.

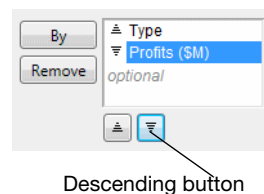
At this point, both variables are set to be sorted in ascending order. See the ascending icon next to the variables in Figure 3.29.

Figure 3.29 Sort Ascending Icon



5. To change Profits (\$M) to sort in descending order, select Profits (\$M) and click the descending button.

Figure 3.30 Change Profits to Descending



The icon next to Profits (\$M) changes to descending.

6. Select the **Replace Table** check box.

When selected, the **Replace Table** option tells JMP to sort the original data table instead of creating a new table with the sorted values. This option is not available if there are any open report windows created from the original data table. Sorting a data table with open report windows might change how some of the data is displayed in the report window, especially in graphs.

7. Click **OK**.

The data table is now sorted by type alphabetically, and by descending profit totals within type.

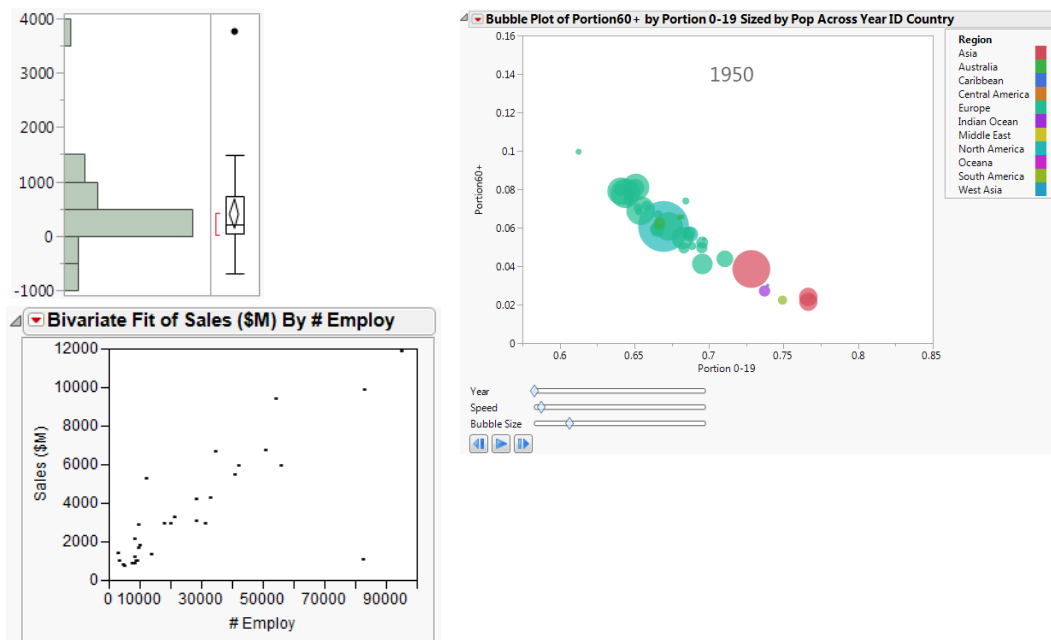
Chapter 4

Visualize Your Data Common Graphs

Visualizing your data is an important first step. The graphs described in this chapter help you discover important details about your data. For example, histograms show you the shape and range of your data, and help you find unusual data points.

This chapter presents several of the most common graphs and plots that enable you to visualize and explore data in JMP. This chapter is an introduction to some of JMP's graphical tools and platforms. Use JMP to visualize the distribution of single variables, or the relationships among multiple variables.

Figure 4.1 Visualizing Data with JMP



Contents

Analyze Single Variables	97
Histograms	97
Bar Charts	99
Compare Multiple Variables	102
Scatterplots	102
Scatterplot Matrix	106
Side-by-Side Box Plots	109
Overlay Plots	112
Variability Chart	115
Graph Builder	118
Bubble Plots	124

Analyze Single Variables

Single-variable graphs, or *univariate* graphs, let you look closely at one variable at a time. When you begin to look at your data, it's important to learn about each variable before looking at how the variables interact with each other. Univariate graphs let you visualize each variable individually.

This section covers two graphs that show the distribution of a single variable:

- “[Histograms](#)” on page 97, for continuous variables
- “[Bar Charts](#)” on page 99, for categorical variables

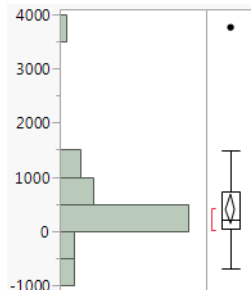
Use the Distribution platform to create both of these graphs. Distribution produces a graphical description and descriptive statistics for each variable.

Histograms

The histogram is one of the most useful graphical tools for understanding the distribution of a continuous variable. Use a histogram to find the following in your data:

- the average value and variation
- extreme values

Figure 4.2 Example of a Histogram



Scenario

This example uses the Companies.jmp data table, which contains data on profits for a group of companies.

A financial analyst wants to explore the following questions:

- Generally, how much profit does each company earn?
- What is the average profit?

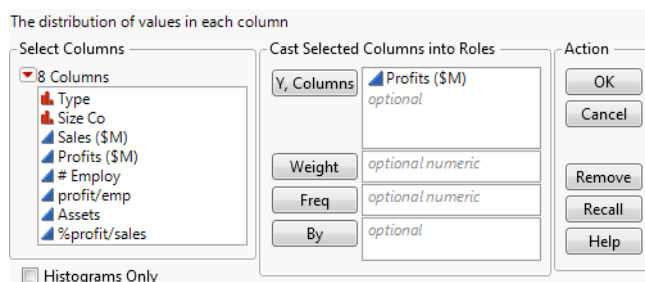
- Are there any companies that earn either extremely high or extremely low profits compared to the other companies?

To answer these questions, use a histogram of Profits (\$M).

Create the Histogram

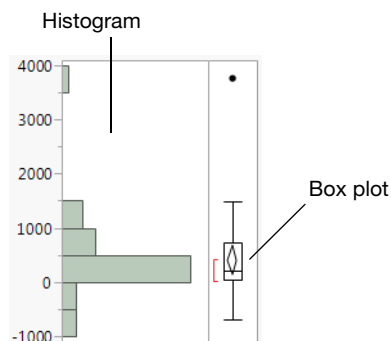
1. Select **Help > Sample Data Library** and open Companies.jmp.
2. Select **Analyze > Distribution**.
3. Select Profits (\$M) and click **Y, Columns**.

Figure 4.3 Distribution Window for Profits (\$M)



4. Click **OK**.

Figure 4.4 Histogram of Profits (\$M)



Interpret the Histogram

The histogram provides these answers:

- Most companies' profits are between \$-1000 and \$1500.
 All the bars except for one are located in this range. Also, more companies' profits range from \$0 to \$500 than any other range. The bar representing that range is much longer than the others.

- The average profit is a little less than \$500.

The middle of the diamond in the box plot indicates the mean value. In this case, the mean is slightly lower than the \$500 mark.

- One company has significantly higher profits than the others, and might be an *outlier*. An outlier is a data point that is separated from the general pattern of the other data points.

This outlier is represented by a single, very short bar at the top of the histogram. The bar is small and represents a small group (in this case, a single company), and it is widely separated from the rest of the histogram bars.

In addition to the histogram, this report includes the following:

- The box plot, which is another graphical summary of the data. For detailed information about the box plot, see the Graph Builder chapter in the *Essential Graphing* book.
- **Quantiles** and **Summary Statistics** reports. These reports are discussed in “[Analyze Distributions](#)” on page 139 in the “Analyze Your Data” chapter.

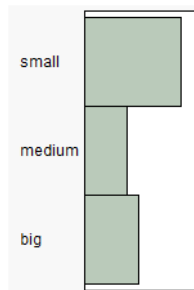
Interact with the Histogram

Data tables and reports are all connected in JMP. Click on a histogram bar to select the corresponding rows in the data table.

Bar Charts

Use a bar chart to visualize the distribution of a categorical variable. A bar chart looks similar to a histogram, since they both have bars that correspond to the levels of a variable. A bar chart shows a bar for every level of the variable, whereas the histogram shows a range of values for the variable.

Figure 4.5 Example of a Bar Chart



Scenario

This example uses the Companies.jmp data table, which contains data on the size and type of a group of companies.

A financial analyst wants to explore the following questions:

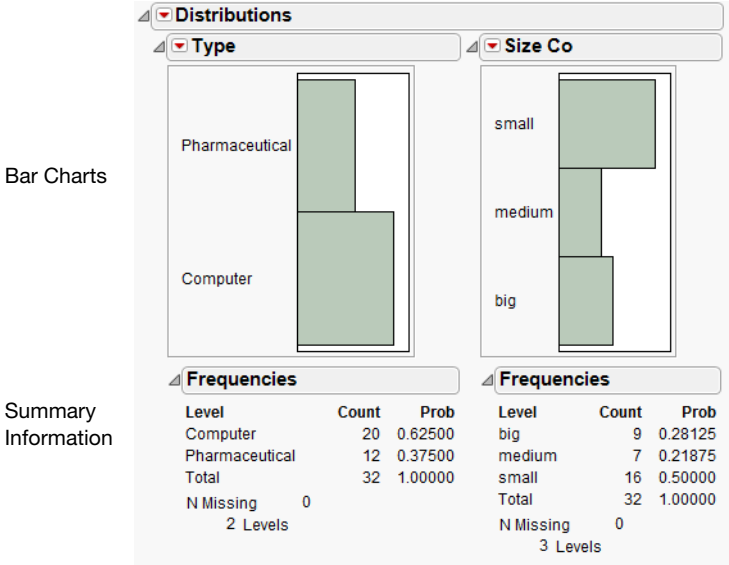
- What is the most common type of company?
- What is the most common size for a company?

To answer these questions, use bar charts of Type and Size Co.

Create the Bar Chart

1. Select **Help > Sample Data Library** and open Companies.jmp.
2. Select **Analyze > Distribution**.
3. Select Type and Size Co and click **Y, Columns**.
4. Click **OK**.

Figure 4.6 Bar Charts of Type and Size Co



Interpret the Bar Charts

The bar charts provide these answers:

- There are more computer companies than pharmaceutical companies.

The bar that represents computer companies is larger than the bar that represents pharmaceutical companies.

- The most common company size is small.

The bar that represents small companies is larger than the bars that represent medium and big companies.

The additional summary output gives detailed frequencies. This report is discussed in [“Distributions of Categorical Variables”](#) on page 142 in the “Analyze Your Data” chapter.

Interact with the Bar Charts

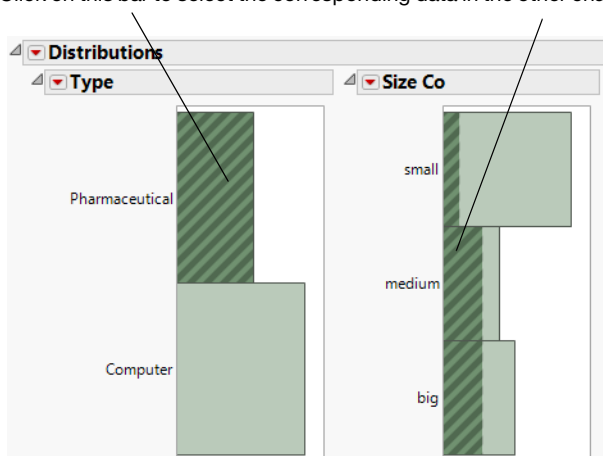
As is the case with histograms, click on individual bars to highlight rows of the data table. If more than one graph is created, clicking on a bar in one bar chart highlights the corresponding bar or bars in the other bar chart.

For example, suppose that you want to see the distribution of company size for the pharmaceutical companies. Click the Pharmaceutical bar in the Type bar chart, and the pharmaceutical companies are highlighted on the Size Co bar chart. Figure 4.7 shows that although most companies in this data table are small, most of the pharmaceutical companies are medium or big.

Also, the corresponding rows in the data table are selected.

Figure 4.7 Clicking Bars

Click on this bar to select the corresponding data in the other chart.



Compare Multiple Variables

Use multiple-variable graphs to visualize the relationships and patterns between two or more variables. This section covers the following graphs:

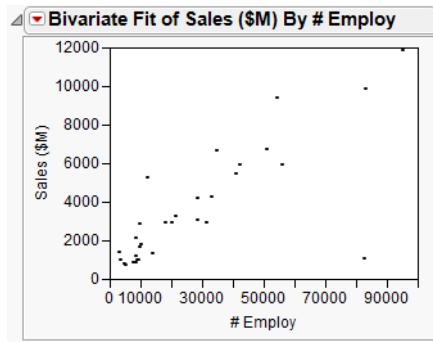
Table 4.1 Multiple-Variable Graphs

“Scatterplots” on page 102	Use scatterplots to compare two continuous variables.
“Scatterplot Matrix” on page 106	Use scatterplot matrices to compare several pairs of continuous variables.
“Side-by-Side Box Plots” on page 109	Use side-by-side box plots to compare one continuous and one categorical variable.
“Overlay Plots” on page 112	Use overlay plots to compare one or more variables on the Y-axis to another variable on the X-axis. Overlay plots are especially useful if the X variable is a time variable, because you can compare how two or more variables change across time.
“Variability Chart” on page 115	Use variability charts to compare one continuous Y variable to one or more categorical X variables. Variability charts show differences in means and variability across several categorical X variables.
“Graph Builder” on page 118	Use Graph Builder to create and change graphs interactively.
“Bubble Plots” on page 124	Bubble plots are specialized scatterplots that use color and bubble sizes to represent up to five variables at once. If one of your variables is a time variable, you can animate the plot to see your other variables change through time.

Scatterplots

The scatterplot is the simplest of all the multiple-variable graphs. Use scatterplots to determine the relationship between two continuous variables and to discover whether two continuous variables are *correlated*. Correlation indicates how closely two variables are related. When you have two variables that are highly correlated, one might influence the other. Or, both might be influenced by other variables in a similar way.

Figure 4.8 Example of a Scatterplot



Scenario

This example uses the *Companies.jmp* data table, which contains sales figures and the number of employees of a group of companies.

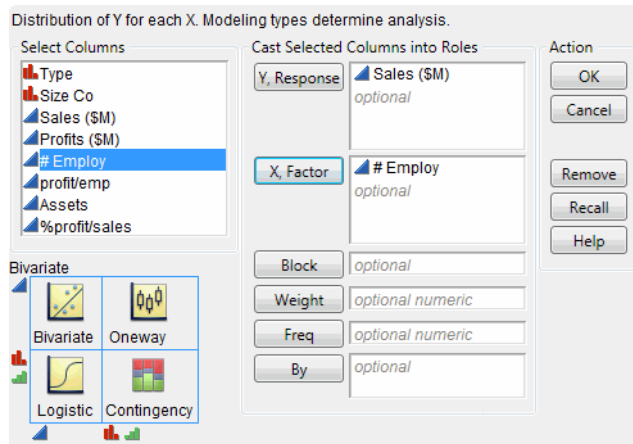
A financial analyst wants to explore the following questions:

- What is the relationship between sales and the number of employees?
- Does the amount of sales increase with the number of employees?
- Can you predict average sales from the number of employees?

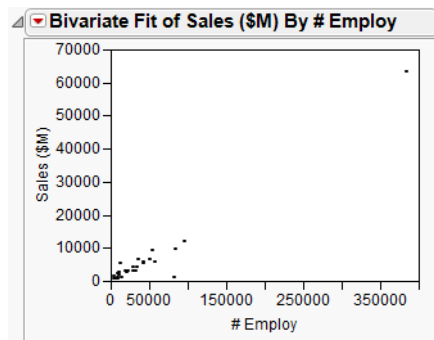
To answer these questions, use a scatterplot of Sales (\$M) versus # Employ.

Create the Scatterplot

1. Select **Help > Sample Data Library** and open *Companies.jmp*.
2. Select **Analyze > Fit Y by X**.
3. Select Sales (\$M) and **Y, Response**.
4. Select # Employ and **X, Factor**.

Figure 4.9 Fit Y by X Window


5. Click OK.

Figure 4.10 Scatterplot of Sales (\$M) versus # Employ


Interpret the Scatterplot

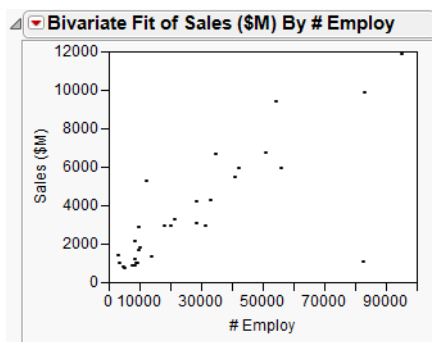
One company has a large number of employees and high sales, represented by the single point at the top right of the plot. The distance between this data point and all the rest makes it difficult to visualize the relationship between the rest of the companies. Remove the point from the plot and re-create the plot by following these steps:

1. Click on the point to select it.
2. Select **Rows > Hide and Exclude**. The data point is hidden and no longer included in calculations.

Note: The difference between hiding and excluding is important. Hiding a point removes it from any graphs but statistical calculations continue to use the point. Excluding a point removes it from any statistical calculations but does not remove it from graphs. When you both hide and exclude a point, you remove it from all calculations and from all graphs.

3. To re-create the plot without the outlier, select **Redo > Redo Analysis** from the red triangle menu for Bivariate. You can close the original report window.

Figure 4.11 Scatterplot with the Outlier Removed



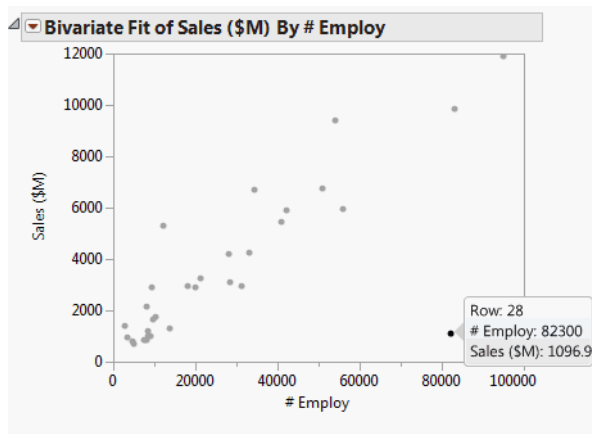
The updated scatterplot provides these answers:

- There is a relationship between the sales and the number of employees.
The data points have a discernible pattern. They are not scattered randomly throughout the graph. You could draw a diagonal line that would be near most of the data points.
- Sales do increase with the number of employees, and the relationship is linear.
If you drew that diagonal line, it would slope from bottom left to top right. This slope shows that as the number of employees increases (left to right on the bottom axis), sales also increases (bottom to top on the left axis). A straight line would be near most of the data points, indicating a linear relationship. If you would have to curve your line to be near the data points, there would still be a relationship (because of the pattern of the points). However, that relationship would not be linear.
- You can predict average sales from the number of employees.
The scatterplot shows that sales generally increase as the number of employees does. You could predict the sales for a company if you knew only the number of employees of that company. Your prediction would be on that imaginary line. It would not be exact, but it would approximate the real sales.

Interact with the Scatterplot

As with other JMP graphics, the scatterplot is interactive. Place your mouse pointer over the point in the bottom right corner with the mouse to reveal the row number and the x and y values.

Figure 4.12 Place Your Mouse Pointer Over a Point



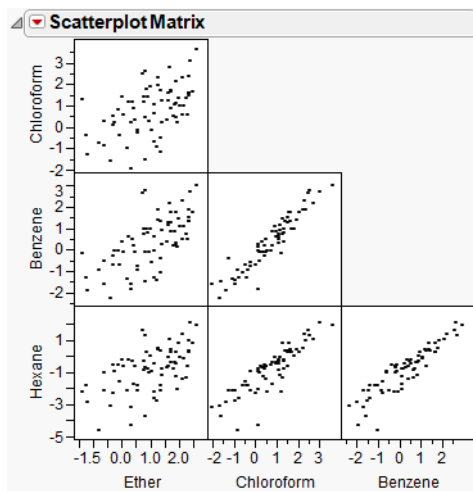
Click on a point to highlight the corresponding row in the data table. Select multiple points by doing one of the following:

- Click and drag with the mouse around the points. This selects points in a rectangular area.
- Select the lasso tool, and then click and drag around multiple points. The lasso tool selects an irregularly shaped area.

Scatterplot Matrix

A scatterplot matrix is a collection of scatterplots organized into a grid (or matrix). Each scatterplot shows the relationship between a pair of variables.

Figure 4.13 Example of a Scatterplot Matrix



Scenario

This example uses the Solubility.jmp data table, which contains data for solubility measurements for 72 different solutes.

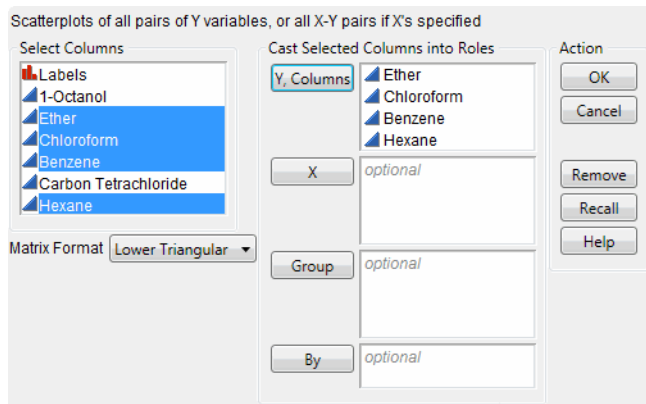
A lab technician wants to explore the following questions:

- Is there a relationship between any pair of chemicals? (There are six possible pairs.)
- Which pair has the strongest relationship?

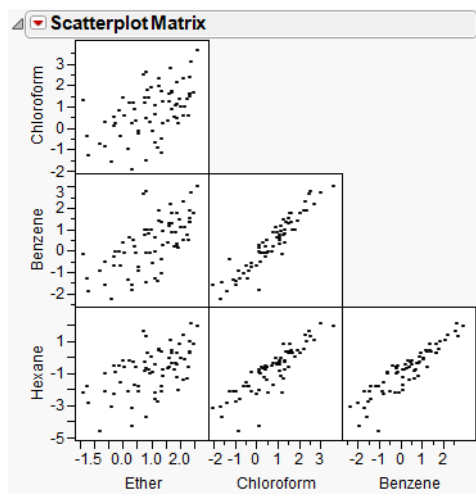
To answer these questions, use a scatterplot matrix of the four solvents.

Create the Scatterplot Matrix

1. Select **Help > Sample Data Library** and open Solubility.jmp.
2. Select **Graph > Scatterplot Matrix**.
3. Select Ether, Chloroform, Benzene, and Hexane, and click **Y, Columns**.

Figure 4.14 Scatterplot Matrix Window


4. Click OK.

Figure 4.15 Scatterplot Matrix


Interpret the Scatterplot Matrix

The scatterplot matrix provides these answers:

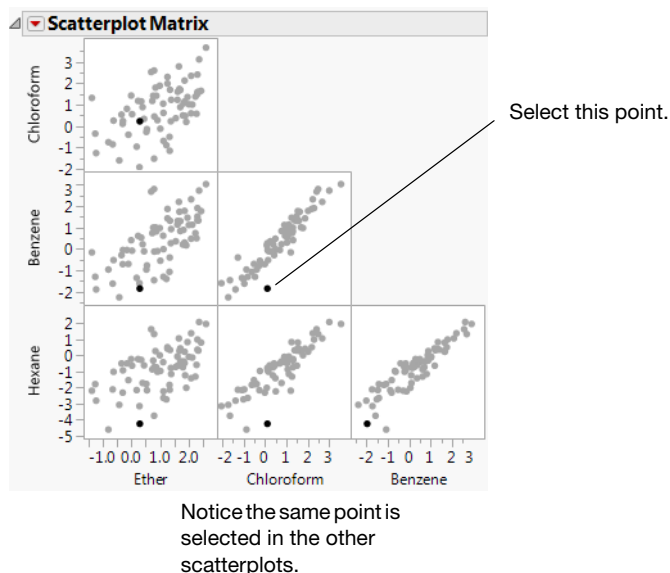
- All six pairs of variables are positively correlated.
As one variable increases, the other variable increases too.
- The strongest relationship appears to be between Benzene and Chloroform.
The data points in the scatterplot for Benzene and Chloroform are the most tightly clustered along an imaginary line.

Interact with the Scatterplot Matrix

If you select a point in one scatterplot, it is selected in all the other scatterplots.

For example, if you select a point in the Benzene versus Chloroform scatterplot, the same point is selected in the other five plots.

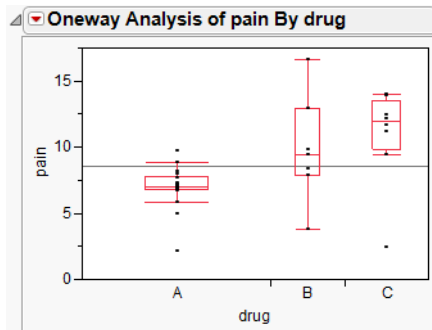
Figure 4.16 Selected Points



Side-by-Side Box Plots

Side-by-side box plots show the following:

- the relationship between one continuous variable and one categorical variable
- differences in the continuous variable across levels of the categorical variable

Figure 4.17 Example of Side-by-Side Box Plots

Scenario

This example uses the *Analgesics.jmp* data table, which contains data on pain measurements taken on patients using three different drugs.

A researcher wants to explore the following questions:

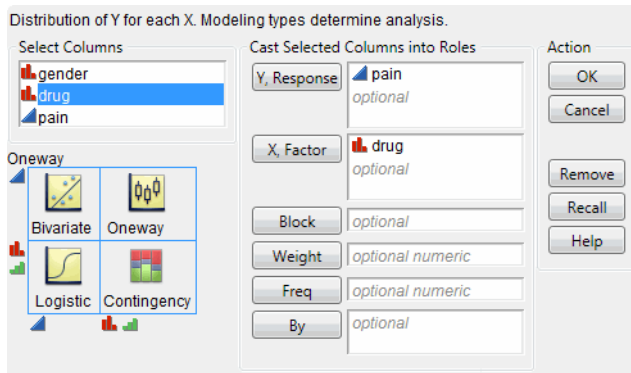
- Are there differences in the average amount of pain control among the drugs?
- Does the *variability* in the pain control given by each drug differ? A drug with high variability would not be as reliable as a drug with low variability.

To answer these questions, use a side-by-side box plot for the pain levels and the drug categories.

Create the Side-by-Side Box Plots

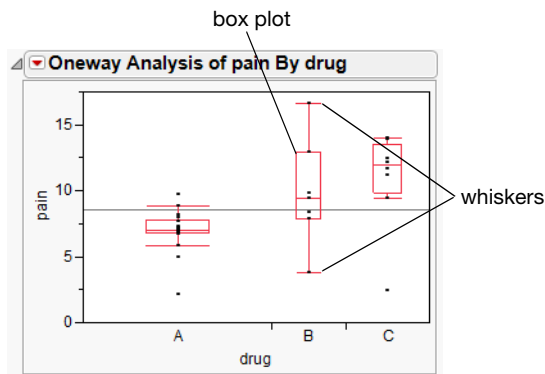
1. Select **Help > Sample Data Library** and open *Analgesics.jmp*.
2. Select **Analyze > Fit Y by X**.
3. Select *pain* and click **Y, Response**.
4. Select *drug* and click **X, Factor**.

Figure 4.18 Fit Y by X Window



- Click **OK**.
- From the red triangle menu, select **Display Options > Box Plots**.

Figure 4.19 Side-by-Side Box Plots



Interpret the Side-by-Side Box Plots

Box plots are designed according to the following principles:

- The line through the box represents the median.
- The middle half of the data is within the box.
- The majority of the data falls between the ends of the whiskers.
- A data point outside the whiskers might be an outlier.

The box plots in Figure 4.19 show these answers:

- There is evidence to believe that patients on drug A feel less pain, since the box plot for drug A is lower on the pain scale than the others.

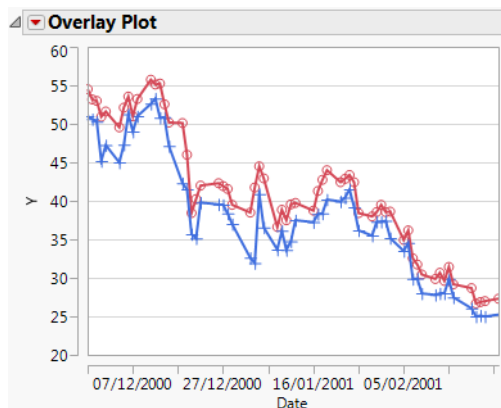
- Drug B appears to have higher variability than Drugs A and C, since the box plot is taller.

There is one point for drug C that is a lot lower than the other points for drug C. Place your mouse pointer over it with your mouse to see that it is row 26 of the data table. That point looks like it is more similar to the data in drug group A or B. The information in row 26 deserves investigation. There might have been a typographical error when the data was recorded.

Overlay Plots

Like scatterplots, overlay plots show the relationship between two or more variables. However, if one of the variables is a time variable, an overlay plot shows trends across time better than scatterplots do.

Figure 4.20 Example of an Overlay Plot



Note: To plot data over time, you can also use Graph Builder, bubble plots, control charts, and variability charts. For complete details about Graph Builder and bubble plots, see the Graph Builder chapter in the *Essential Graphing* book. Refer to the Control Chart Builder chapter and the Variability Gauge Charts chapter in the *Quality and Process Methods* book for information about control charts and variability charts.

Scenario

This example uses the `Stock Prices.jmp` data table, which contains data on the price of a stock over a three-month period.

A potential investor wants to explore the following questions:

- Has the stock's closing price changed over the past three months?

To answer this question, use an overlay plot of the stock's closing price over time.

- How do the stock's high and low prices relate to each other?

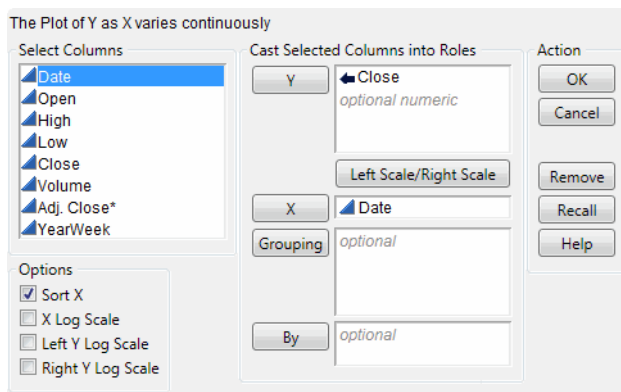
To answer this question, use another overlay plot of the stock's high and low prices over time.

Create the first overlay plot to answer the first question, and then create a second overlay plot to answer the second question.

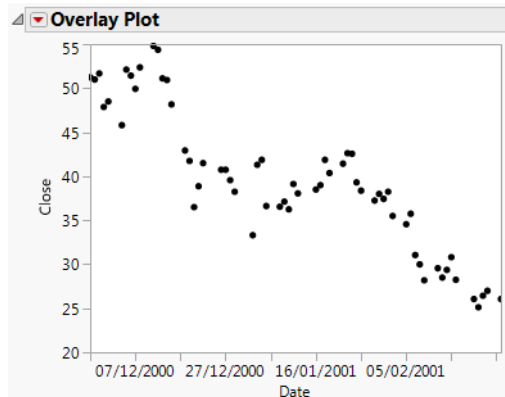
Create the Overlay Plot of the Stock's Price over Time

1. Select **Help > Sample Data Library** and open Stock Prices.jmp.
2. Select **Graph > Legacy > Overlay Plot**.
3. Select Close and click Y.
4. Select Date and click X.

Figure 4.21 Overlay Plot Window



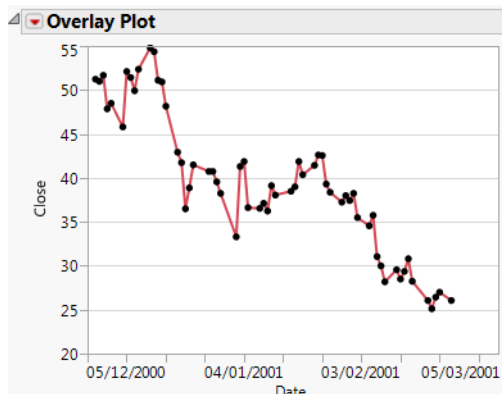
5. Click **OK**.

Figure 4.22 Overlay Plot of the Closing Price over Time

Interpret and Interact with the Overlay Plot

The overlay plot shows that the closing stock price has been decreasing over the last several months. To see the trend more clearly, connect the points and add grid lines.

1. From the red triangle menu, select **Connect Thru Missing**.
2. Double-click the Y axis.
3. Select the **Major Grid Lines** check box.
4. Click **OK**.

Figure 4.23 Connected Points and Grid Lines

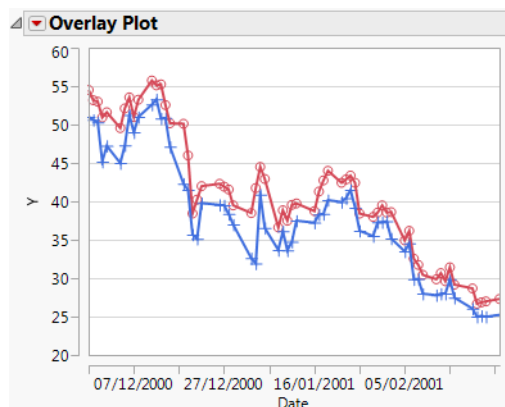
The potential investor can see that although the stock price has gone up and down over the past three months, the overall trend has been downward.

Create the Overlay Plot of the Stock's High and Low Prices

Use an overlay plot to plot more than one Y variable. For example, suppose that you want to see both the high and the low prices on the same plot.

1. Follow the steps in [“Create the Overlay Plot of the Stock's Price over Time”](#) on page 113, this time assigning both High and Low to the Y role.
2. Connect the points and add grid lines as shown in [“Interpret and Interact with the Overlay Plot”](#) on page 114.

Figure 4.24 Two Y Variables



The legend at the bottom of the plot shows the colors and markers used for the High and Low variables in the graph. The overlay plot shows that the High price and Low price track each other very closely.

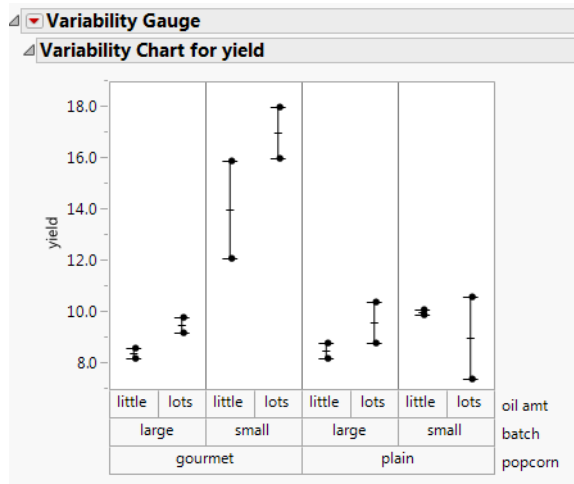
Answer the Questions

Both of the overlay plots answer the two questions asked at the beginning of this example.

- The first plot shows that the price of this stock has not remained the same, but has been decreasing.
- The second plot shows that the high and low prices of this stock are not very different from each other. The stock price does not vary wildly on any given day.

Variability Chart

In the graphs described so far, you specified only a single X variable. Use a variability chart to specify multiple X variables and see differences in means and variability across all of your variables at once.

Figure 4.25 Example of a Variability Chart


Scenario

This example uses the Popcorn.jmp data table with data from a popcorn maker. The yield (the volume of popcorn for a given measure of kernels) was measured for each combination of popcorn style, batch size, and amount of oil used.

The popcorn maker wants to explore the following question:

- Which combination of factors results in the highest popcorn yield?

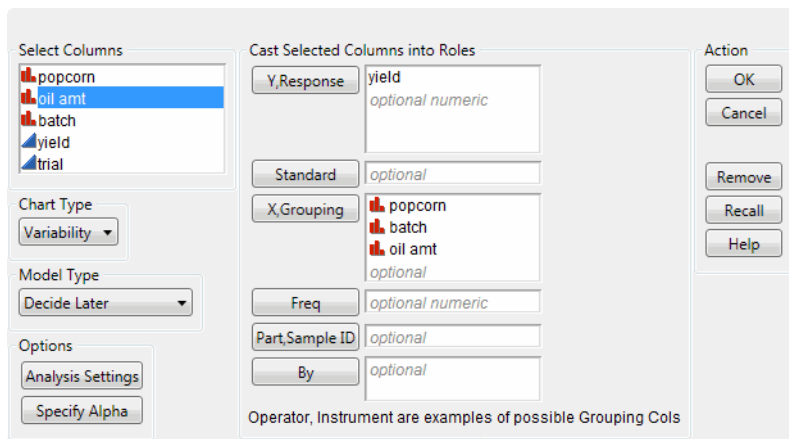
To answer this question, use a variability chart of the yield versus the style, batch size, and oil amount.

Create the Variability Chart

1. Select **Help > Sample Data Library** and open Popcorn.jmp.
2. Select **Analyze > Quality and Process > Variability/Attribute Gauge Chart**.
3. Select yield and click **Y, Response**.
4. Select popcorn and click **X, Grouping**.
5. Select batch and click **X, Grouping**.
6. Select oil amt and click **X, Grouping**.

Note: The order in which you assign the variables to the **X, Grouping** role is important, because the order in this window determines their nesting order in the variability chart.

Figure 4.26 Variability Chart Window

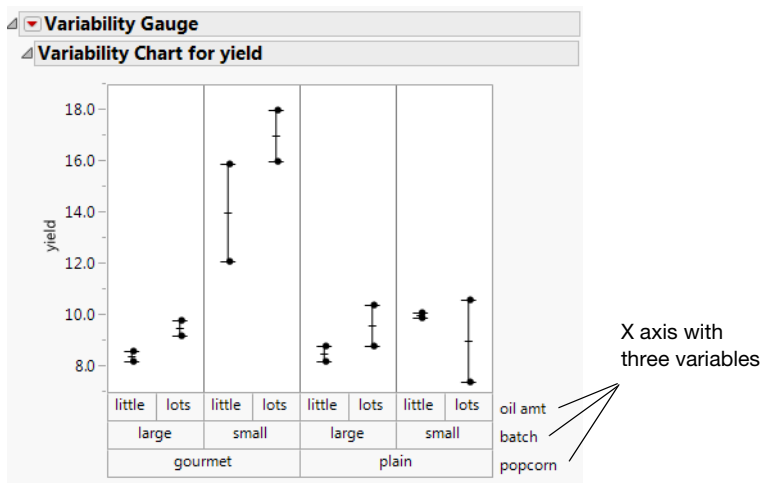


7. Click **OK**.

The top chart is the variability chart, showing the yield broken down by each combination of the three variables. The bottom chart shows the standard deviation for each combination of the three variables. Since the bottom chart does not show the yield, hide it.

8. Deselect **Std Dev Chart** on the red triangle menu.

Figure 4.27 Results Window



Interpret the Variability Chart

The variability chart for yield indicates that small, gourmet batches produce the highest yield.

To be more specific, the popcorn maker might ask this additional question: Is the yield high because those batches are small, or because those batches are gourmet?

The variability chart shows the following:

- The yield from small, plain batches is low.
- The yield from large, gourmet batches is low.

Given this information, the popcorn maker can conclude that only the combination of small and gourmet at the same time results in batches with high yield. It would have been impossible to reach this conclusion with a chart that only allowed a single variable.

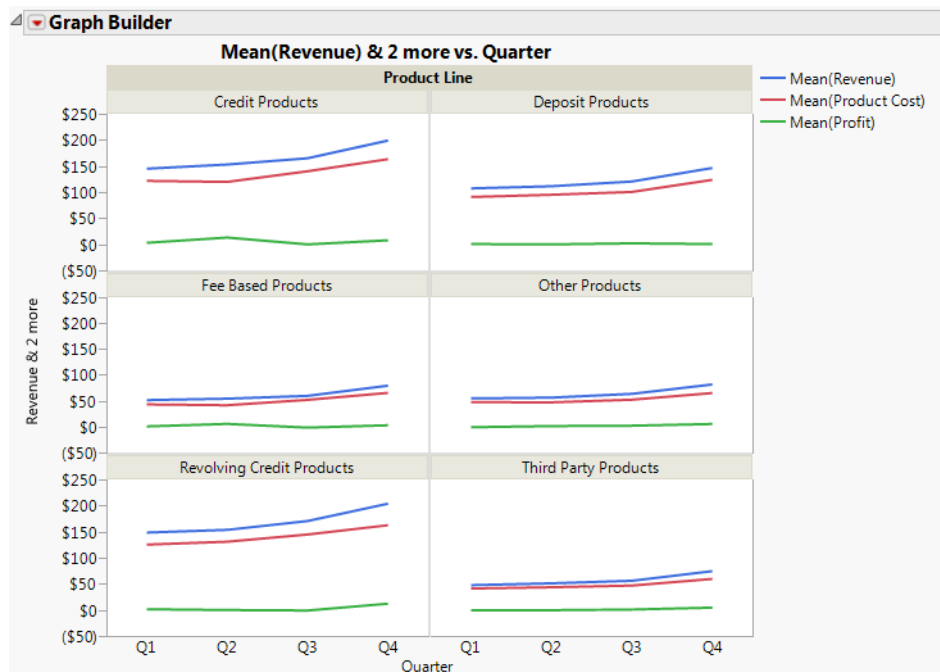
Graph Builder

Use Graph Builder to interactively create and modify graphs. Most graphs in JMP are created by launching a platform and specifying variables. If you want to create a different type of graph, you launch a specific platform from the Graph menu. However, with Graph Builder, you can change the variables and change the type of graph at any time.

Use Graph Builder to accomplish the following tasks:

- Change variables by dragging and dropping them in and out of the graph.
- Create a different type of graph with a few mouse clicks.
- Partition the graph horizontally or vertically.

Figure 4.28 Example of a Graph That Was Created with Graph Builder



Note: Only some of the Graph Builder features are covered here. For complete details, see the Graph Builder chapter in the *Essential Graphing* book.

Scenario

This example uses the Profit by Product.jmp data table, which contains profit data for multiple product lines.

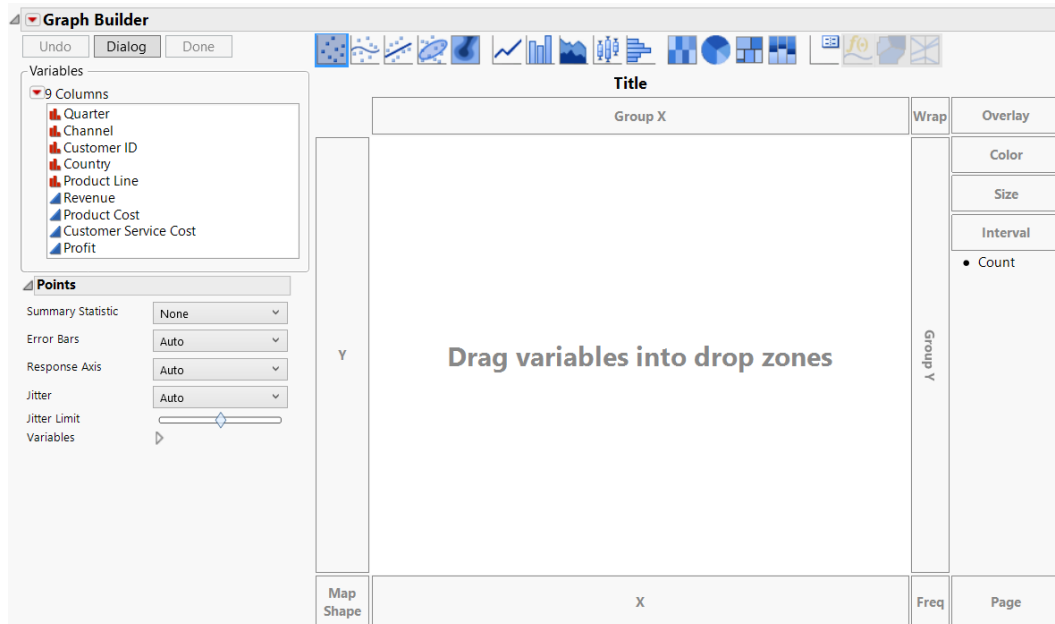
A business analyst wants to explore the following question:

- How is the profitability different between product lines?

To answer this question, use a line plot that displays revenue, product cost, and profit data across different product lines.

Create the Graph

1. Select **Help > Sample Data Library** and open Profit by Product.jmp.
2. Select **Graph > Graph Builder**.

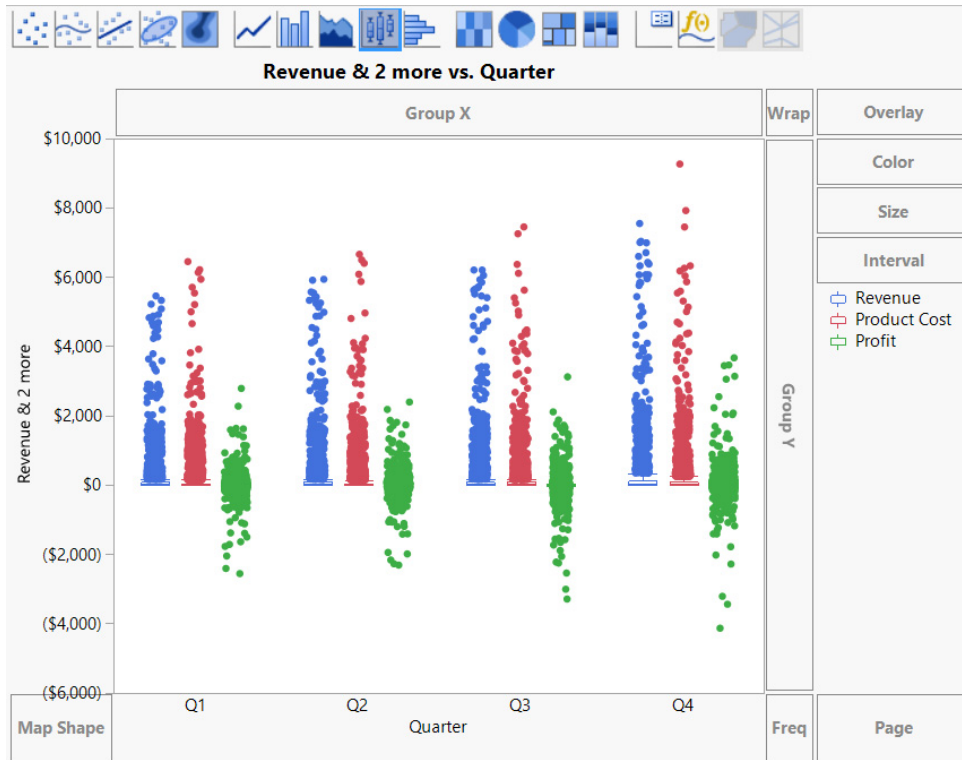
Figure 4.29 Graph Builder Workspace


3. Click Quarter and then drag and drop it onto the X zone to assign Quarter as the X variable.
4. Click Revenue, Product Cost, and Profit, and drag and drop them onto the Y zone to assign all three variables as Y variables.

The X and Y zones are now axes.

Note: You can also click on variables and then click a zone to assign them. However, after a zone becomes an axis, drag and drop additional variables onto the axis rather than clicking on the variables and axis.

Figure 4.30 After Adding Y and X Variables



Based on the variables that you are using, Graph Builder shows side-by-side box plots.


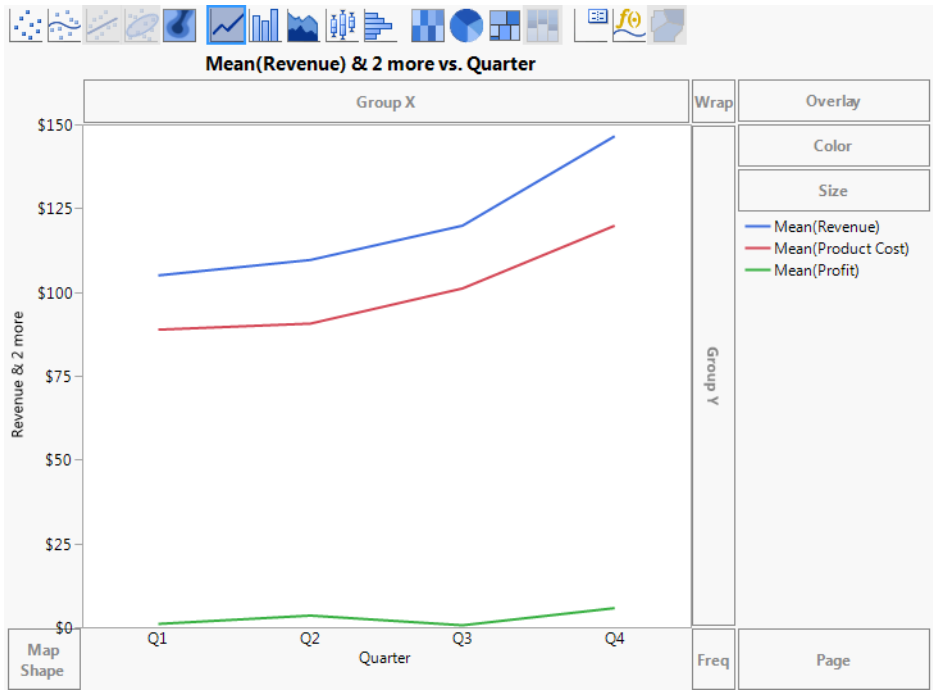
- To change the box plots to a line plot, click the Line  icon.

Figure 4.31 Line Plot



6. To create a separate chart for each product, click Product Line, and drag and drop it into the **Wrap** zone.

A separate line plot is created for each product.

Figure 4.32 Final Line Plots



Interpret the Graph

Figure 4.32 shows revenue, cost, and profit broken down by product line. The business analyst was interested in seeing the difference in profitability between product lines. The line plots in Figure 4.32 can provide some answers, as follows:

- Credit products, deposit products, and revolving credit products produce more revenue than fee-based products, third-party products, and other products.
- However, the profits of all the product lines are similar.

The data table also includes data on sales channels. The business analyst wants to see how revenue, product cost, and profit differ between different sales channels.

1. To remove Product Line from the graph, click the title of the graph (Product Line) and drag and drop it into any empty area within Graph Builder.
2. To add Channel as the wrap variable, click Channel and drag and drop it into the **Wrap** zone.

Figure 4.33 Line Plots Showing Sales Channels

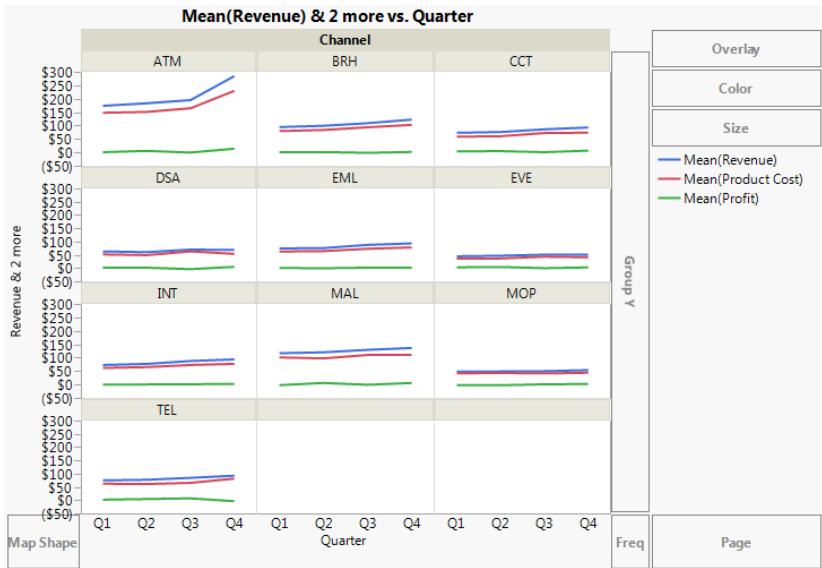
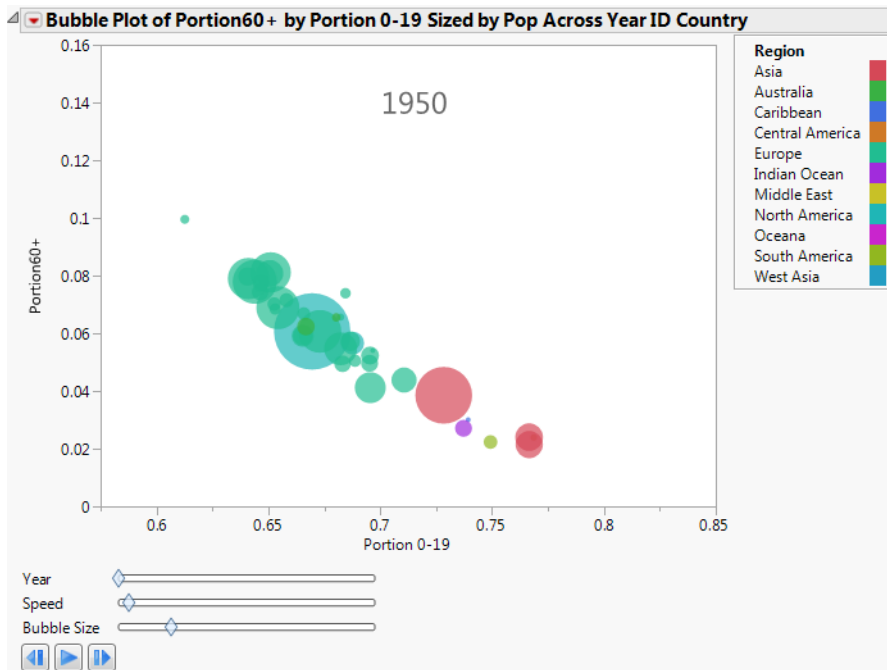


Figure 4.33 provides this answer: revenue and product cost for ATMs are the highest and are growing the most quickly.

Bubble Plots

A bubble plot is a scatterplot that represents its points as bubbles. You can change the size and color of the bubbles, and even animate them over time. With the ability to represent up to five dimensions (*x* position, *y* position, size, color, and time), a bubble plot can produce dramatic visualizations and make data exploration easy.

Figure 4.34 Example of a Bubble Plot



Scenario

This example uses the `PopAgeGroup.jmp` data table, which contains population statistics for 116 countries or territories between the years 1950 to 2004. Total population numbers are broken out by age group, and not every country has data for every year.

A sociologist wants to explore the following question:

- Is the age of the population of the world changing?

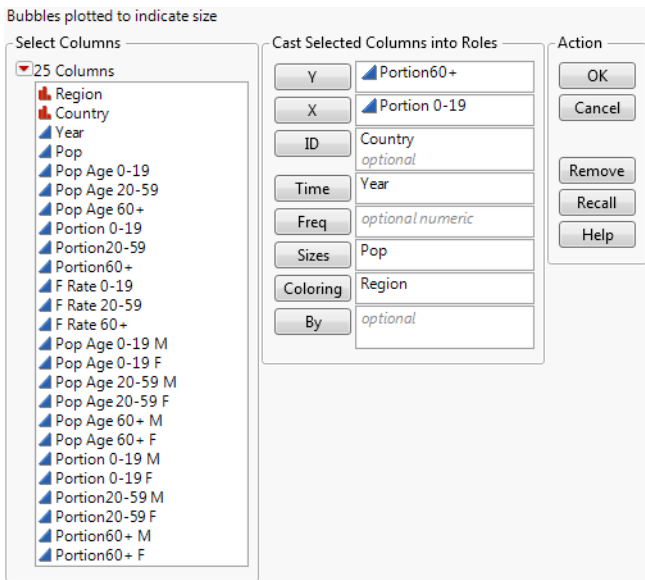
To answer this question, look at the relationship between the oldest (more than 59) and the youngest (younger than 20) portions of the population. Use a bubble plot to determine how this relationship changes over time.

Create the Bubble Plot

1. Select **Help > Sample Data Library** and open `PopAgeGroup.jmp`.
2. Select **Graph > Bubble Plot**.
3. Select `Portion60+` and click **Y**.
This corresponds to the Y variable on the bubble plot.
4. Select `Portion 0-19` and click **X**.

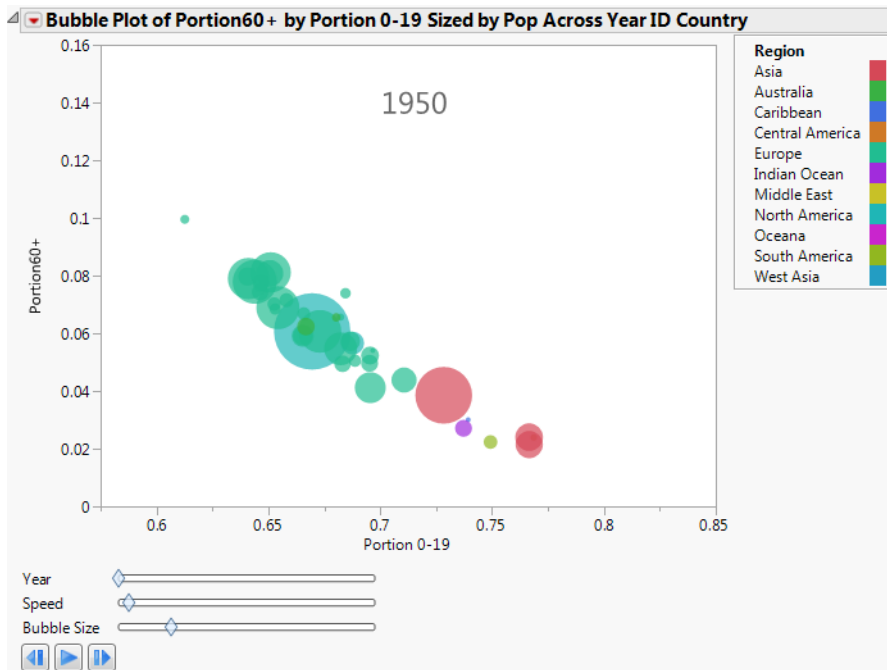
- This corresponds to the *X* variable on the bubble plot.
- 5. Select **Country** and click **ID**.
Each unique level of the ID variable is represented by a bubble on the plot.
 - 6. Select **Year** and click **Time**.
This controls the time indexing when the bubble plot is animated.
 - 7. Select **Pop** and click **Sizes**.
This controls the size of the bubbles.
 - 8. Select **Region** and click **Coloring**.
Each level of the Coloring variable is assigned a unique color. So in this example, all the bubbles for countries located in the same region have the same color. The bubble colors that appear in Figure 4.36 are the JMP default colors.

Figure 4.35 Bubble Plot Launch Window



- 9. Click **OK**.

Figure 4.36 Initial Bubble Plot



Interpret the Bubble Plot


Because the time variable (in this case, year) starts in 1950, the initial bubble plot shows the data for 1950. Animate the bubble plot to cycle through all the years by clicking the play/pause button. Each successive bubble plot shows the data for that year. The data for each year determines the following:


- The X and Y coordinates
- The bubble's sizes
- The bubble's coloring
- Bubble aggregation


Note: For detailed information about how the bubble plot aggregates information across multiple rows, see the Bubble Plots chapter in the *Essential Graphing* book.


The bubble plot for 1950 shows that if a country's proportion of people younger than 20 is high, then the proportion of people more than 59 is low.

Click the play/pause button to animate the bubble plot through the range of years. As time progresses, the Portion 0-19 decreases and the Portion60+ increases.

 plays the animation, turns to a pause button after you click it.

 pauses the animation.

 manually controls the animation back one unit of time.

 manually controls the animation forward one unit of time.

Year is used to change the time index manually.

Speed controls the speed of the animation.

Bubble Size controls the absolute sizes of the bubbles, while maintaining the relative sizes

The sociologist wanted to know how the age of the world's population is changing. The bubble plot indicates that the population of the world is getting older.

Interact with the Bubble Plot

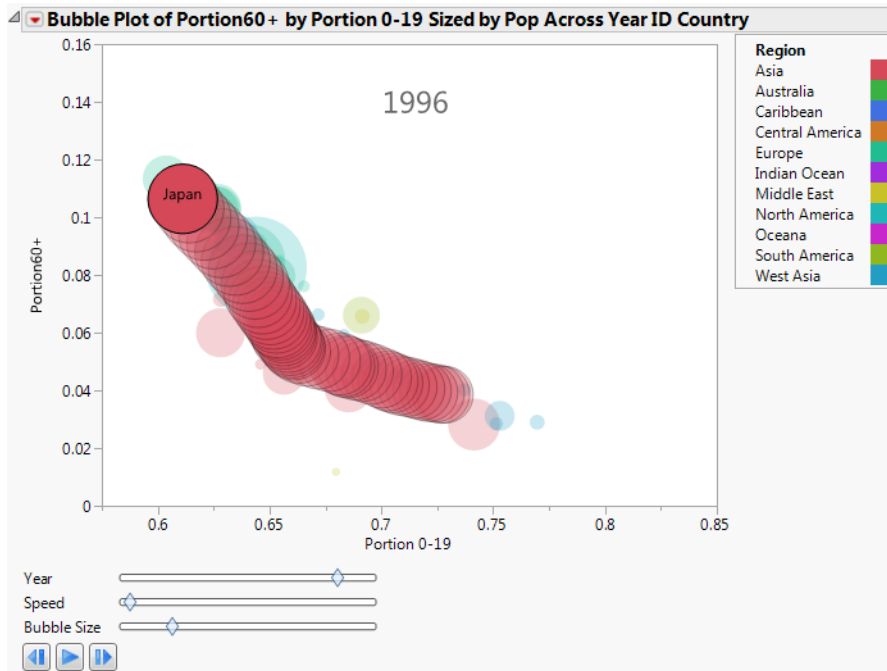
Click to select a bubble to see the trend for that bubble over time. For example, in the 1950 plot, the large bubble in the middle is Japan.

To See the Pattern of Population Changes in Japan through the Years

1. Click in the middle of the Japan bubble to select it.
2. From the red triangle menu, select **Trail Bubbles > Selected**.
3. Click the play button.

As the animation progresses through time, the Japan bubble leaves a trail of bubbles that illustrates its history.

Figure 4.37 Japan's History of Population Shifts



Focusing on the Japan bubble, you can see the following over time:

- The proportion of the population 19 years old or less decreased.
- The proportion of the population 60 years old or more increased.

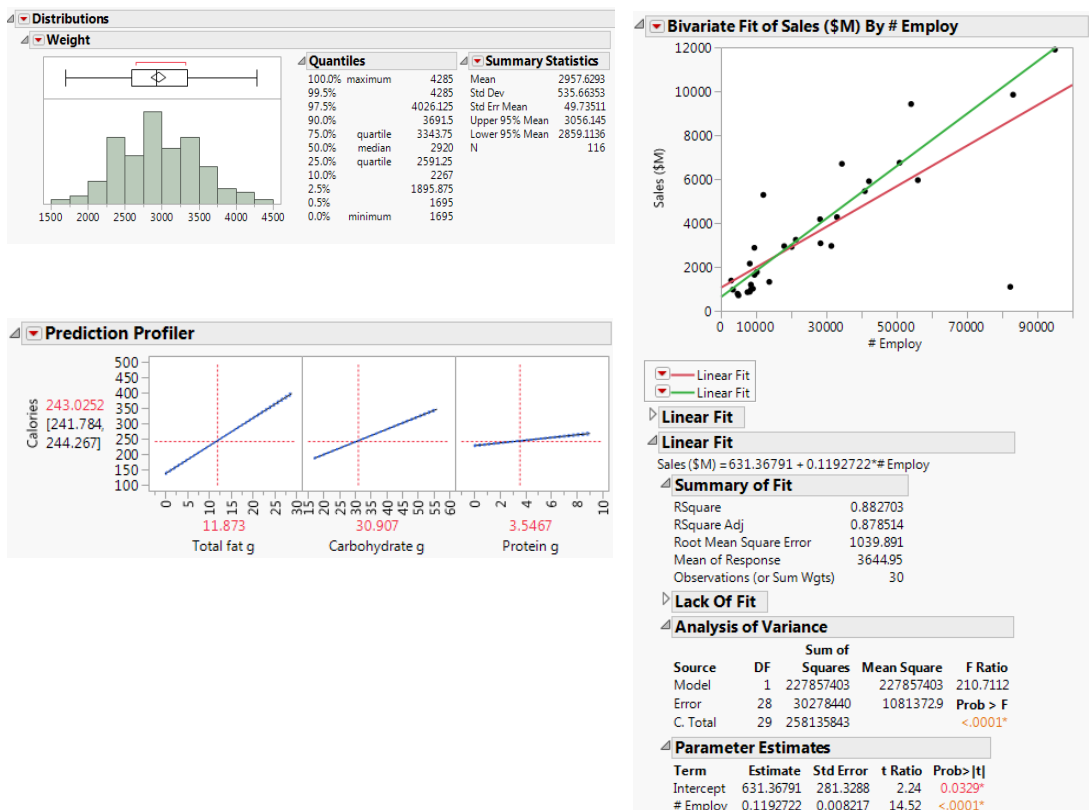
Chapter 5

Analyze Your Data Distributions, Relationships, and Models

Analyzing your data helps you make informed decisions. Data analysis often involves these actions:

- Examining distributions
- Discovering relationships
- Hypothesis testing
- Building models

Figure 5.1 Analysis Examples



Contents

About This Chapter	133
The Importance of Graphing Your Data	133
Understand Modeling Types	136
Example of Viewing Modeling Type Results	136
Change the Modeling Type	137
Analyze Distributions	139
Distributions of Continuous Variables	140
Distributions of Categorical Variables	142
Analyze Relationships	145
Use Regression with One Predictor	146
Compare Averages for One Variable	151
Compare Proportions	154
Compare Averages for Multiple Variables	156
Use Regression with Multiple Predictors	160

About This Chapter

Before you analyze your data, review the following information:

- [“The Importance of Graphing Your Data”](#) on page 133
- [“Understand Modeling Types”](#) on page 136

The rest of this chapter shows you how to use some basic analytical methods in JMP:

- [“Analyze Distributions”](#) on page 139
- [“Analyze Relationships”](#) on page 145

For a description of advanced modeling and analysis techniques, refer to the following JMP books:

- *Fitting Linear Models*
- *Multivariate Methods*
- *Predictive and Specialized Modeling*
- *Consumer Research*
- *Reliability and Survival Methods*
- *Quality and Process Methods*

The Importance of Graphing Your Data

Graphing, or visualizing, your data is important to any data analysis, and should always occur before the use of statistical tests or model building. To illustrate why data visualization should be an early step in your data analysis process, consider the following example:

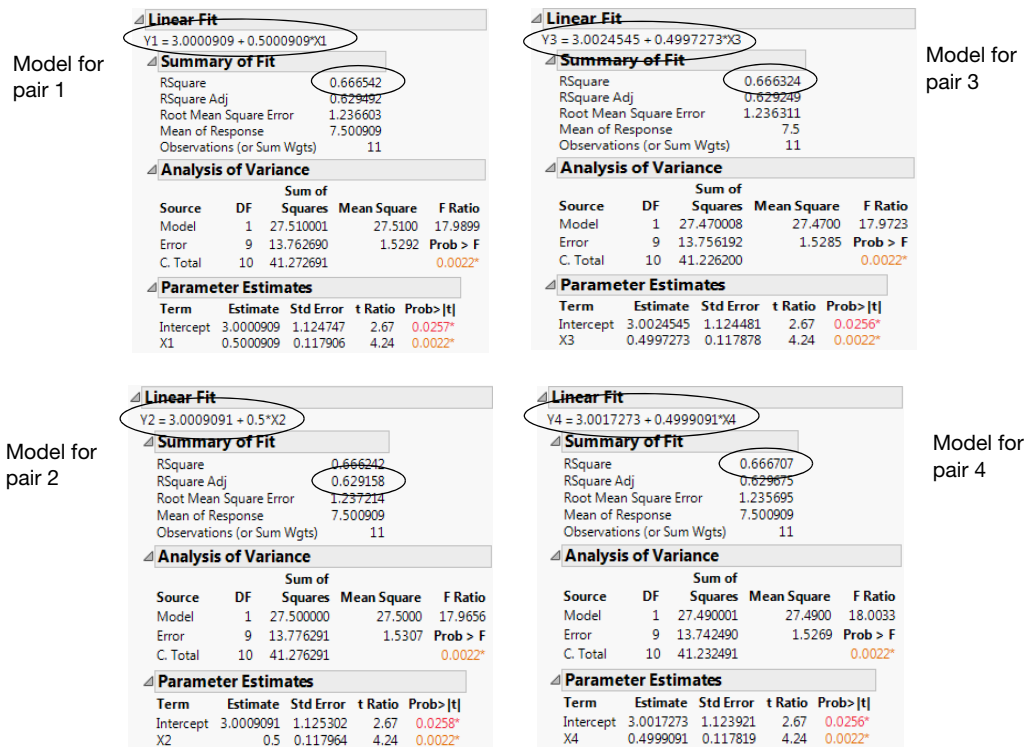
1. Select **Help > Sample Data Library** and open *Anscombe.jmp* (F. J. Anscombe (1973), *American Statistician*, 27, 17-21).

This data consists of four pairs of X and Y variables.

2. In the Table panel, click the green triangle next to the **The Quartet** script.

The script creates a simple linear regression on each pair of variables using **Fit Y by X**. The **Show Points** option is turned off, so that none of the data can be seen on the scatterplots. Figure 5.2 shows the model fit and other summary information for each regression.

Figure 5.2 Four Models

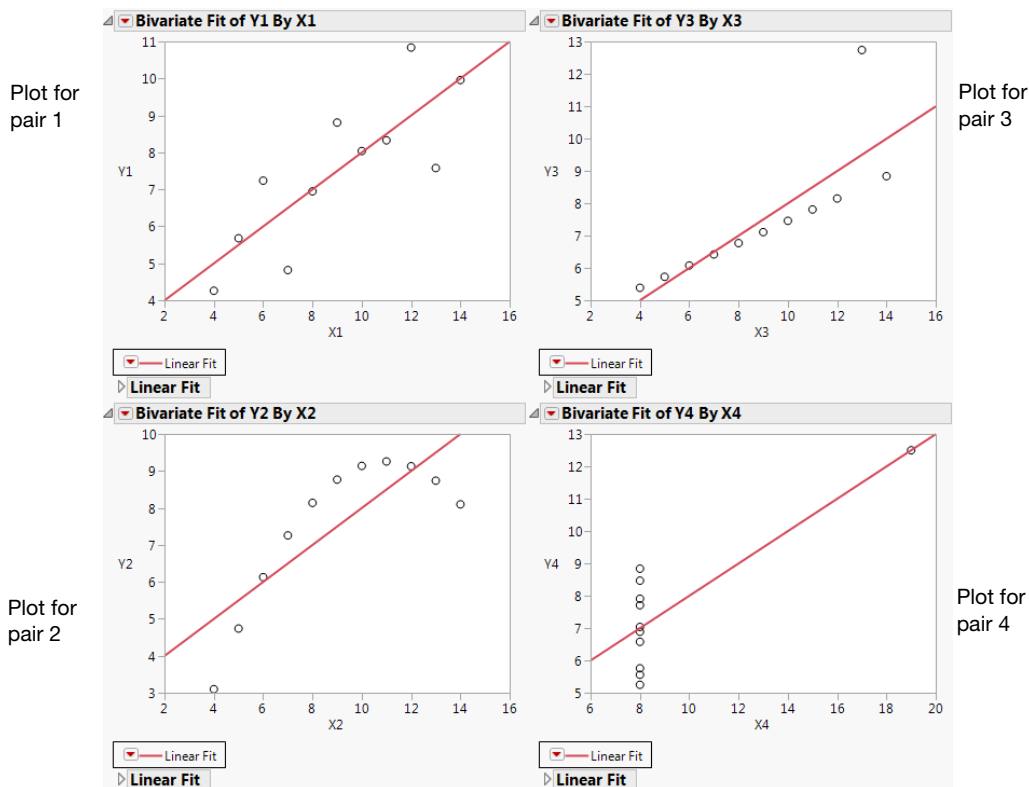


Notice that all four models and the RSquare values are nearly identical. The fitted model in each case is essentially $Y = 3 + 0.5X$, and the RSquare value in each case is essentially 0.66. If your data analysis took into account only the above summary information, you would likely conclude that the relationship between X and Y is the same in each case. However, at this point, you have not visualized your data. Your conclusion might be wrong.

To Visualize the Data, Add the Points to All Four Scatterplots

1. Hold down the Ctrl key.
2. From the red triangle menu for any one of the Bivariate Fits, select **Show Points**.

Figure 5.3 Scatterplots with Points Added



The scatterplots show that the relationship between X and Y is not the same for the four pairs, although the lines describing the relationships are the same:

- Plot 1 represents a linear relationship.
- Plot 2 represents a non-linear relationship.
- Plot 3 represents a linear relationship, except for one outlier.
- Plot 4 has all the data at $x = 8$, except for one point.

This example illustrates that conclusions that are based on statistics alone can be inadequate. A visual exploration of the data should be an early part of any data analysis.

Understand Modeling Types

In JMP, data can be of different types. JMP refers to this as the modeling type of the data. Table 5.1 describes the three modeling types in JMP.

Table 5.1 Modeling Types

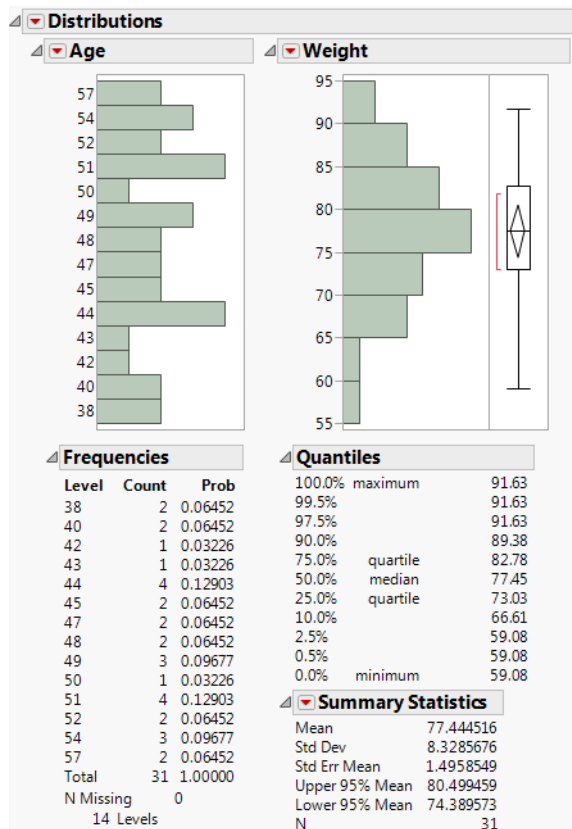
Modeling Type and Description	Examples	Specific Example
Continuous Numeric data only. Used in operations like sums and means.	Height Temperature Time	The time to complete a test might be 2 hours, or 2.13 hours.
Ordinal Numeric or character data. Values belong to ordered categories.	Month (1,2,...,12) Letter grade (A, B,...F) Size (small, medium, large)	The month of the year can be 2 (February) or 3 (March), but not 2.13. February comes before March.
Nominal Numeric or character data. Values belong to categories, but the order is not important.	Gender (M or F) Color Test result (pass or fail)	The gender can be M or F, with no order. Gender categories can also be represented by a number (M=1 and F=2).

Example of Viewing Modeling Type Results

Different modeling types produce different results in JMP. To see an example of the differences, follow these steps:

1. Select **Help > Sample Data Library** and open Linnerud.jmp.
2. Select **Analyze > Distribution**.
3. Select Age and Weight and click **Y, Columns**.
4. Click **OK**.

Figure 5.4 Distribution Results for Age and Weight



Although Age and Weight are both numeric variables, they are not treated the same. Table 5.2 compares the differences between the results for weight and age.

Table 5.2 Results for weight and age

Variable	Modeling Type	Results
Weight	Continuous	Histogram, Quantiles, and Summary Statistics
Age	Ordinal	Bar chart and Frequencies

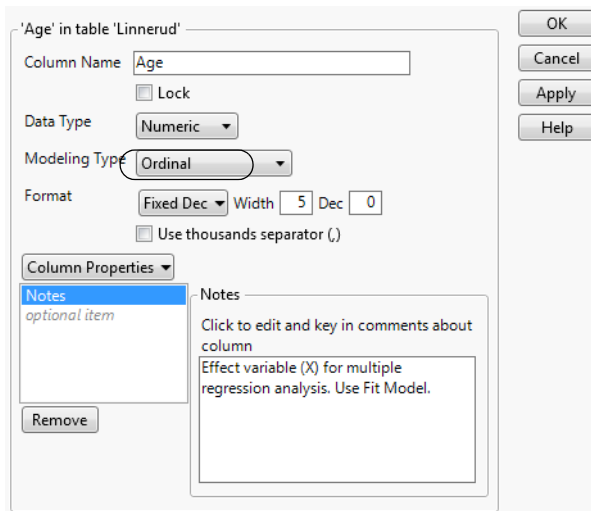
Change the Modeling Type

To treat a variable differently, change the modeling type. For example, in Figure 5.4, the modeling type for Age is ordinal. Remember that for an ordinal variable, JMP calculates

frequency counts. Suppose that you wanted to find the average age instead of frequency counts. Change the modeling type to continuous, which shows the mean age.

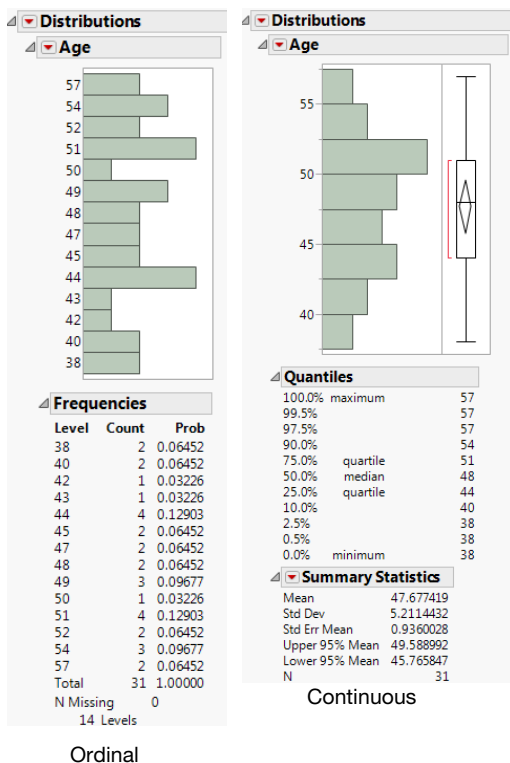
1. Double-click the **Age** column heading. The Column Info window appears.
2. Change the Modeling Type to **Continuous**.

Figure 5.5 Column Info Window



3. Click **OK**.
4. Repeat the steps in the example (see [“Example of Viewing Modeling Type Results”](#) on page 136) to create the distribution. Figure 5.6 shows the distribution results when Age is ordinal and continuous.

Figure 5.6 Different Modeling Types for Age



When age is ordinal, you can see the frequency counts for each age. For example, age 48 appears two times. When age is continuous, you can find the mean age, which is nearly 48 (47.677)

Analyze Distributions

To analyze a single variable, you can examine the distribution of the variable, using the Distribution platform. Report content for each variable varies, depending on whether the variable is categorical (nominal or ordinal) or continuous.

Note: For complete details about the **Distribution** platform, see the Distributions chapter in the *Basic Analysis* book.

Distributions of Continuous Variables

Analyzing a continuous variable might include questions such as the following:

- Does the shape of the data match any known distributions?
- Are there any outliers in the data?
- What is the average of the data?
- Is the average statistically different from a target or historical value?
- How spread out are the data? In other words, what is the standard deviation?
- What are the minimum and maximum values?

You can answer these and other questions with graphs, summary statistics, and simple statistical tests.

Scenario

This example uses the *Car Physical Data.jmp* data table, which contains information about 116 different car models.

A planning specialist has been asked by a railroad company to determine the possible issues involved in transporting cars by train. Using the data, the planning specialist wants to explore the following questions:

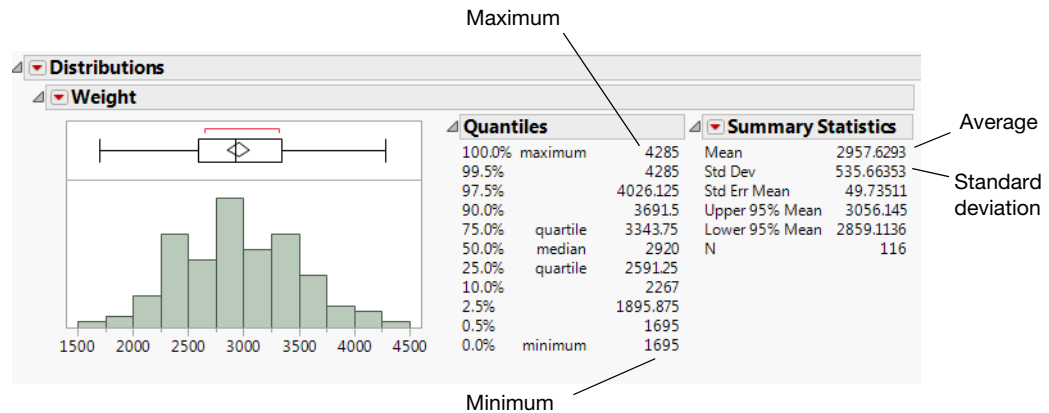
- What is the average car weight?
- How spread out are the cars' weights (standard deviation)?
- What are the minimum and maximum weights of cars?
- Are there any outliers in the data?

Use a histogram of weight to answer these questions.

Create the Histogram

1. Select **Help > Sample Data Library** and open *Car Physical Data.jmp*.
2. Select **Analyze > Distribution**.
3. Select **Weight** and click **Y, Columns**.
4. Click **OK**.
5. To rotate the report window, select **Display Options > Horizontal Layout** from the red triangle menu next to **Weight**.

Figure 5.7 Distribution of Weight



The report window contains three sections:

- A histogram and a box plot to visualize the data.
- A Quantiles report that shows the percentiles of the distribution.
- A Summary Statistics report that shows the mean, standard deviation, and other statistics.

Interpret the Distribution Results

Using the results presented in Figure 5.7, the planning specialist can answer the questions.

What is the average car weight? The Histogram shows a weight of around 3,000 lbs.

How spread out are the weights (standard deviation)? The Summary Statistics show a weight of around 2,958 lbs. The Summary Statistics show a standard deviation of around 536 lbs.

What are the minimum and maximum weights? The Histogram shows a minimum of around 1,500 lbs. and a maximum of around 4,500 lbs. The Quantiles show a minimum of around 1,695 lbs. and a maximum of around 4,285 lbs.

Are there any outliers? No.

The default report window in Figure 5.7 provides a minimal set of graphs and statistics. Additional graphs and statistics are available on the red triangle menu.

Draw Conclusions

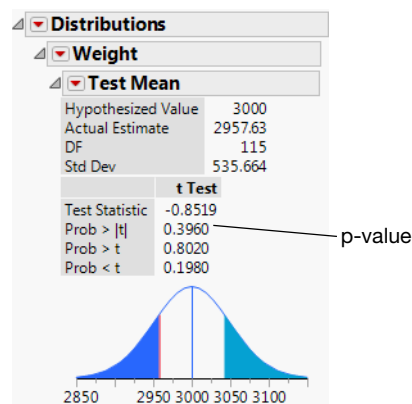
Based on other research, the railroad company has determined that an average weight of 3000 pounds is the most efficient to transport. Now, the planning specialist needs to find out whether the average car weight in the general population of cars that they might transport is

3000 pounds. Use a t-test to draw inferences about the broader population based on this sample of the population.

Test Conclusions

1. From the red triangle menu for **Weight**, select **Test Mean**.
2. In the window that appears, type 3000 in the Specify Hypothesized Mean box.
3. Click **OK**.

Figure 5.8 Test Mean Results



Interpret the t-Test

The primary result of a t-test is the p-value. In this example, the p-value is 0.396 and the analyst is using a significance level of 0.05. Since 0.396 is greater than 0.05, you cannot conclude that the average weight of car models in the broader population is significantly different from 3000 pounds. Had the p-value been lower than the significance level, the planning specialist would have concluded that the average car weight in the broader population *is* significantly different from 3000 pounds.

Distributions of Categorical Variables

Analyzing a categorical (ordinal or nominal) variable might include questions such as the following:

- How many levels does the variable have?
- How many data points does each level have?
- Is the data uniformly distributed?
- What proportions of the total do each level represent?

Scenario

See the scenario in “[Distributions of Continuous Variables](#)” on page 140.

Now that the railroad company has determined that the average weight of the cars is not significantly different from the target weight, there are more questions to address.

The planning specialist wants to answer these questions for the railroad company:

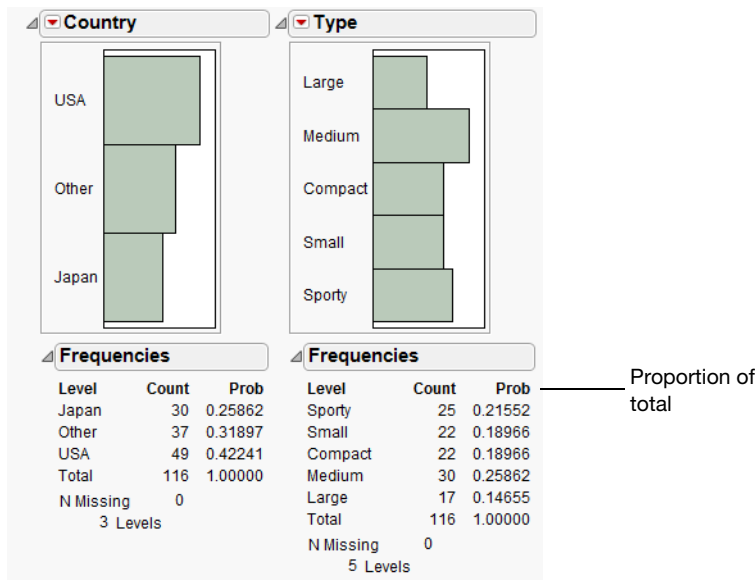
- What are the types of cars?
- What are the countries of origin?

To answer these questions, look at the distribution for Type and Country.

Create the Distribution

1. Select **Help > Sample Data Library** and open Car Physical Data.jmp.
2. Select **Analyze > Distribution**.
3. Select Country and Type and click **Y, Columns**.
4. Click **OK**.

Figure 5.9 Distribution for Country and Type



Interpret the Distribution Results

The report window includes a bar chart and a Frequencies report for Country and Type. The bar chart is a graphical representation of the frequency information provided in the Frequencies report. The Frequencies report contains the following:

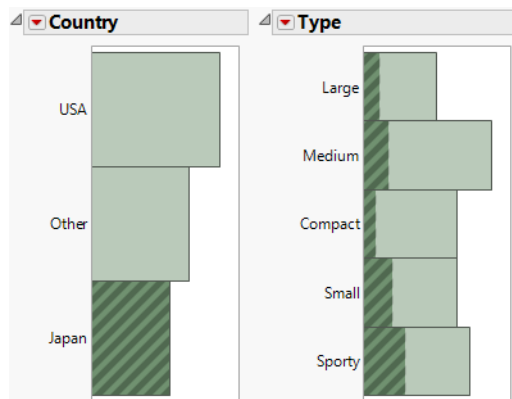
- Categories of data. For example, Japan is a category of Country, and Sporty is a category of Type.
- Total counts for each category.
- Proportion of the total each category represents.

For example, there are 22 compact cars, or about 19% of the 116 observations.

Interact with the Distribution Results

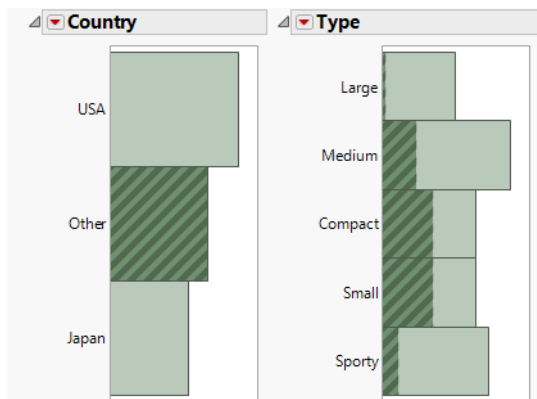
Selecting a bar in one chart also selects the corresponding data in the other chart. For example, select the Japan bar in the Country bar chart to see that a large number of Japanese cars are sporty.

Figure 5.10 Japanese Cars



Select the Other category to see that a majority of these cars are small or compact, and almost none are large.

Figure 5.11 Other Cars



Analyze Relationships

Scatterplots and other such graphs can help you visualize relationships between variables. Once you have visualized relationships, the next step is to analyze those relationships so that you can describe them numerically. That numerical description of the relationship between variables is called a *model*. Even more importantly, a model also predicts the average value of one variable (Y) from the value of another variable (X). The X variable is also called a predictor. Generally, this model is called a *regression* model.

With JMP, the **Fit Y by X** platform and the **Fit Model** platform creates regression models.

Note: Only the basic platforms and options are covered here. For complete details and explanations of all platform options, see *Basic Analysis*, *Essential Graphing*, and the books listed in [“About This Chapter”](#) on page 133.

Table 5.3 shows the four primary types of relationships.

Table 5.3 Relationship Types

X	Y	Section
Continuous	Continuous	<ul style="list-style-type: none"> “Use Regression with One Predictor” on page 146 “Use Regression with Multiple Predictors” on page 160

Table 5.3 Relationship Types (*Continued*)

X	Y	Section
Categorical	Continuous	<ul style="list-style-type: none"> • “Compare Averages for One Variable” on page 151 • “Compare Averages for Multiple Variables” on page 156
Categorical	Categorical	“Compare Proportions” on page 154
Continuous	Categorical	Logistic regression is an advanced topic. See the Logistic Analysis chapter in the <i>Basic Analysis</i> book.

Use Regression with One Predictor

Scenario

This example uses the *Companies.jmp* data table, which contains financial data for 32 companies from the pharmaceutical and computer industries.

Intuitively, it makes sense that companies with more employees can generate more sales revenue than companies with fewer employees. A data analyst wants to predict the overall sales revenue for each company based on the number of employees.

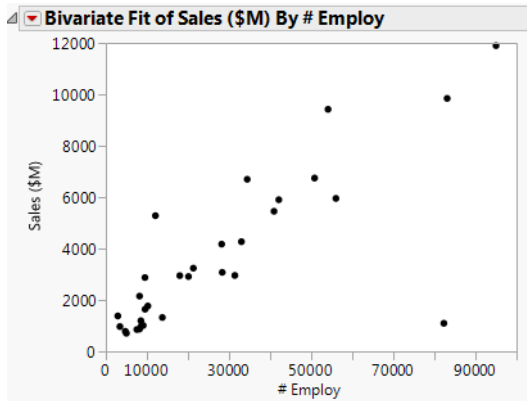
To accomplish this task, do the following:

- [“Discover the Relationship”](#) on page 146
- [“Fit the Regression Model”](#) on page 147
- [“Predict Average Sales”](#) on page 149

Discover the Relationship

First, create a scatterplot to see the relationship between the number of employees and the amount of sales revenue. This scatterplot was created in [“Create the Scatterplot”](#) on page 103 in the “Visualize Your Data” chapter. After hiding and excluding one outlier (a company with significantly more employees and higher sales), the plot in Figure 5.12 shows the result.

Figure 5.12 Scatterplot of Sales (\$M) versus # Employ

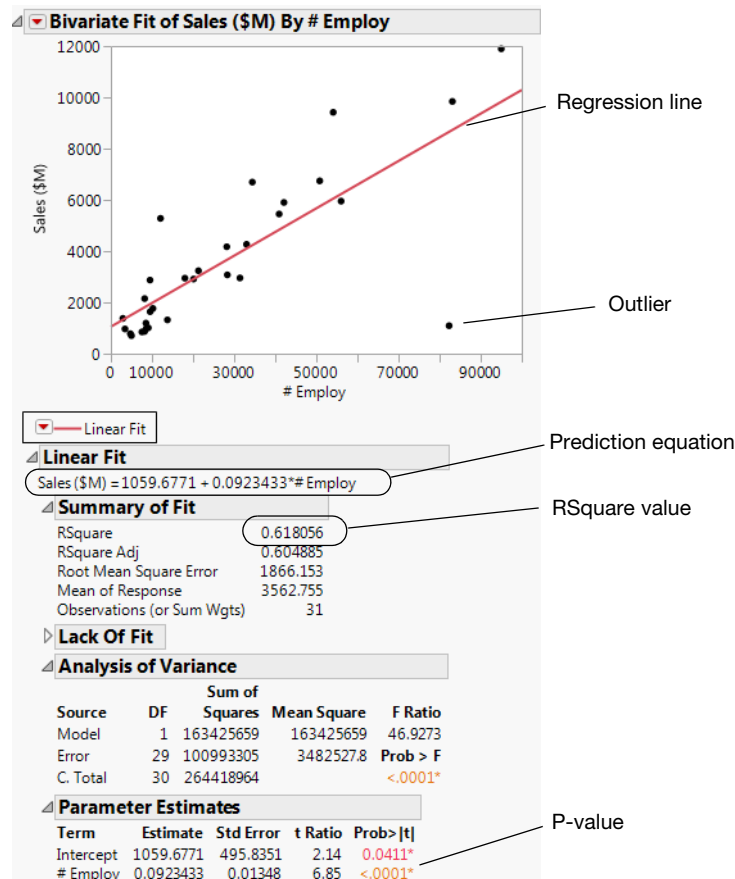


This scatterplot provides a clearer picture of the relationship between sales and the number of employees. As expected, the more employees a company has, the higher sales that it can generate. This visually confirms the data analyst's guess, but it does not predict sales for a given number of employees.

Fit the Regression Model

To predict the sales revenue from the number of employees, fit a regression model. From the red triangle for **Bivariate Fit**, select **Fit Line**. A regression line is added to the scatterplot and reports are added to the report window.

Figure 5.13 Regression Line



Within the reports, look at the following results:

- the p-value of <.0001
- the RSquare value of 0.618

From these results, the data analyst can conclude the following:

- The p-value for the #Employ model term is small. This supports that at the 0.05 significance level the coefficient for #Employ is not zero. Therefore, including the number of employees in the prediction model significantly improves the ability to predict average sales over a model without the number of employees.
- The RSquare value of 0.618 indicates that this model explains about 62% of the variability in sales. The RSquare value is the coefficient of determination and indicates the proportion of the variance in the dependent (response) variable that is explained by your model. RSquare can range from 0 to 1. A model with an RSquare of 0 has no explanatory power. A model with an RSquare of 1 predicts the response perfectly.

Predict Average Sales

Use the regression model to predict the average sales a company might expect if they have a certain number of employees. The prediction equation for the model is included in the report:

$$\text{Average sales} = 1059.68 + 0.092 * \text{employees}$$

For example, in a company with 70,000 employees sales are predicted to be about \$7,500:

$$\$7,499.68 = 1059.68 + 0.092 * 70,000$$

In the lower right area of the current scatterplot, there is an outlier that does not follow the general pattern of the other companies. The data analyst wants to know whether the prediction model changes when this outlier is excluded.

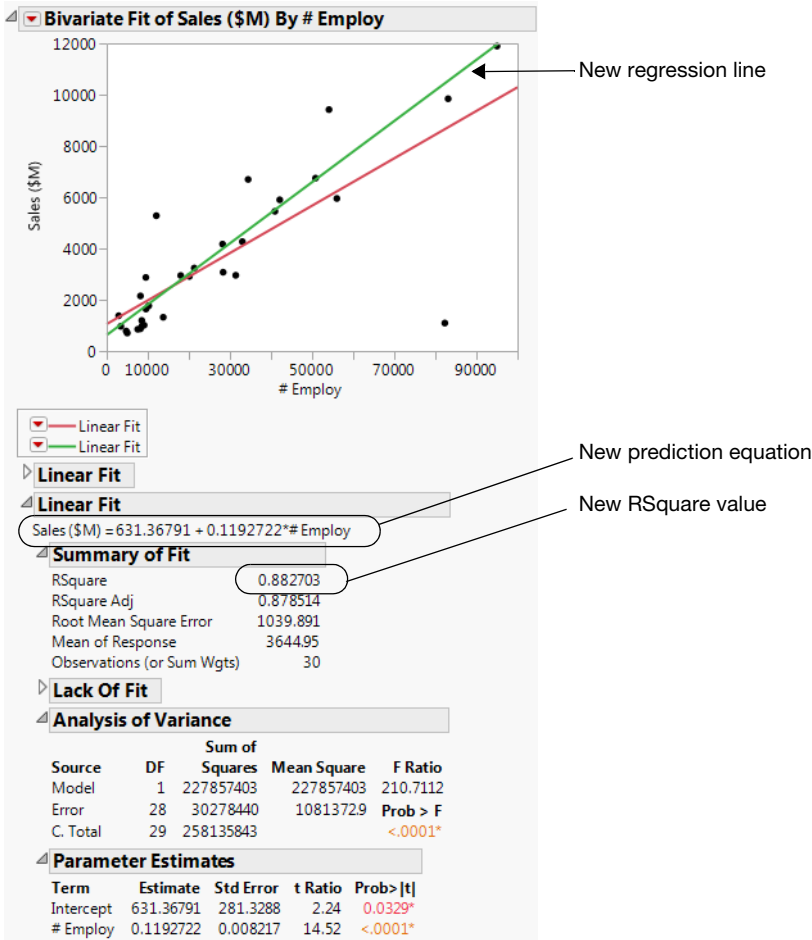
Exclude the Outlier

1. Click on the outlier.
2. Select **Rows > Exclude/Unexclude**.
3. Fit this model by selecting **Fit Line** from the red triangle menu for **Bivariate Fit**.

The following are added to the report window (see Figure 5.14):

- a new regression line
- a new Linear Fit report, which includes:
 - a new prediction equation
 - a new RSquare value

Figure 5.14 Comparing the Models



Interpret the Results

Using the results in Figure 5.14, the data analyst can make the following conclusions:

- The outlier was pulling down the regression line for the larger companies, and pulling the line up for the smaller companies.
- The new model for the data without the outlier is a stronger model than the first model. The new RSquare value of 0.88 is higher and closer to 1 than the initial analysis.

Draw Conclusions

Using the new prediction equation, the predicted average sales for a company with 70,000 employees can be calculated as follows:

$$\$8961.37 = 631.37 + 0.119 * 70,000$$

The prediction from the first model was about \$7500. The second model predicts a sales total of about \$8960 or an increase of \$1460 as compared to the first model.

The second model, after removing the outlier, describes and predicts sales totals based on the number of employees better than the first model. The data analyst now has a good model to use.

Compare Averages for One Variable

If you have a continuous Y variable, and a categorical X variable, you can compare averages across the levels of the X variable.

Scenario

This example uses the Companies.jmp data table, which contains financial data for 32 companies from the pharmaceutical and computer industries.

A financial analyst wants to explore the following question:

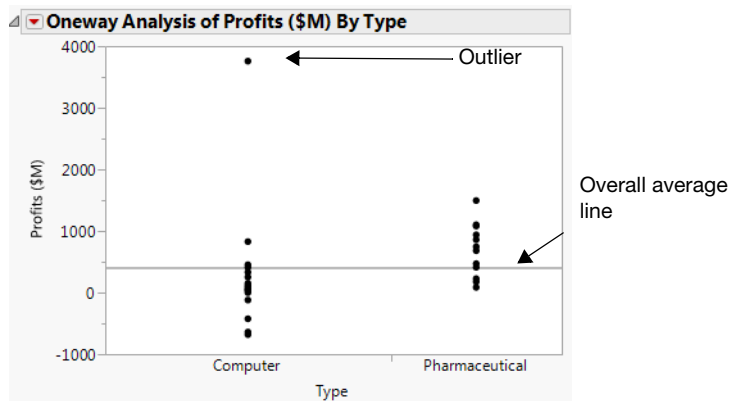
- How do the profits of computer companies compare to the profits of pharmaceutical companies?

To answer this question, fit Profits (\$M) by Type.

Discover the Relationship

1. Select **Help > Sample Data Library** and open Companies.jmp.
2. If you still have the Companies.jmp sample data table open, you might have rows that are excluded or hidden. To return the rows to the default state (all rows included and none hidden), select **Rows > Clear Row States**.
3. Select **Analyze > Fit Y by X**.
4. Select Profits (\$M) and click **Y, Response**.
5. Select Type and click **X, Factor**.
6. Click **OK**.

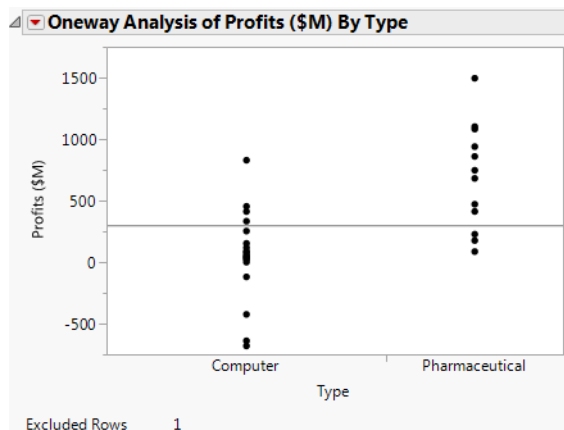
Figure 5.15 Profits by Company Type



There is an outlier in the Computer Type. The outlier is stretching the scale of the plot and making it difficult to compare the profits. Exclude and hide the outlier:

1. Click on the outlier.
2. Select **Rows > Exclude/Unexclude**. The data point is no longer included in calculations.
3. Select **Rows > Hide/Unhide**. The data point is hidden from all graphs.
4. To re-create the plot without the outlier, select **Redo > Redo Analysis** from the red triangle menu for Oneway Analysis. You can close the original Scatterplot window.

Figure 5.16 Updated Plot

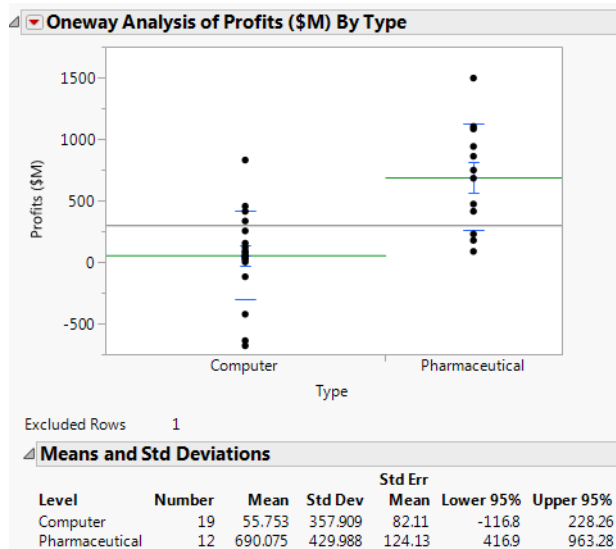


Removing the outlier gives the financial analyst a clearer picture of the data.

5. To continue analyzing the relationship, select these options from the red triangle menu for Oneway Analysis:
 - **Display Options > Mean Lines**. This adds mean lines to the scatterplot.

- **Means and Std Dev.** This displays a report that provides averages and standard deviations.

Figure 5.17 Mean Lines and Report



Interpret the Results

The financial analyst wanted to know how the profits of computer companies compared to the profits of pharmaceutical companies. The updated scatterplot shows that pharmaceutical companies have higher average profits than computer companies. In the report, if you subtract one mean value from the other, the difference in profit is about \$635 million. The plot also shows that some of the computer companies have negative profits and all of the pharmaceutical companies have positive profits.

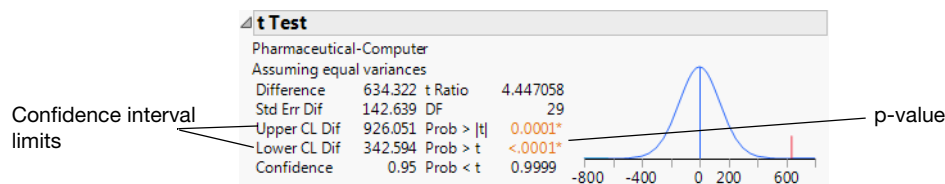
Perform the T-test

The financial analyst has looked at only a sample of companies (the companies in the data table). The financial analyst now wants to examine these questions:

- Does a difference exist in the broader population, or is the difference of \$635 million due to chance?
- If there is a difference, what is it?

To answer these questions, perform a two-sample t-test. A t-test lets you use data from a sample to make inferences about the larger population.

To perform the t-test, select **Means/Anova/Pooled t** from the red triangle for Oneway Analysis.

Figure 5.18 t Test Results


The p-value of 0.0001 is less than the significance level of 0.05, which indicates statistical significance. Therefore, the financial analyst can conclude that the difference in average profits for the sample data is not due to chance alone. This means that in the larger population, the average profits for pharmaceutical companies are different from the average profits for computer companies.

Draw Conclusions

Use the confidence interval limits to determine how much difference exists in the profits of both types of companies. Look at the **Upper CL Dif** and **Lower CL Dif** values in Figure 5.18. The financial analyst concludes that the average profit of pharmaceutical companies is between \$343 million and \$926 million higher than the average profit of computer companies.

Compare Proportions

If you have categorical X and Y variables, you can compare the proportions of the levels within the Y variable to the levels within the X variable.

Scenario

This example continues to use the Companies.jmp data table. In [“Compare Averages for One Variable”](#) on page 151, a financial analyst determined that pharmaceutical companies have higher profits on average than do computer companies.

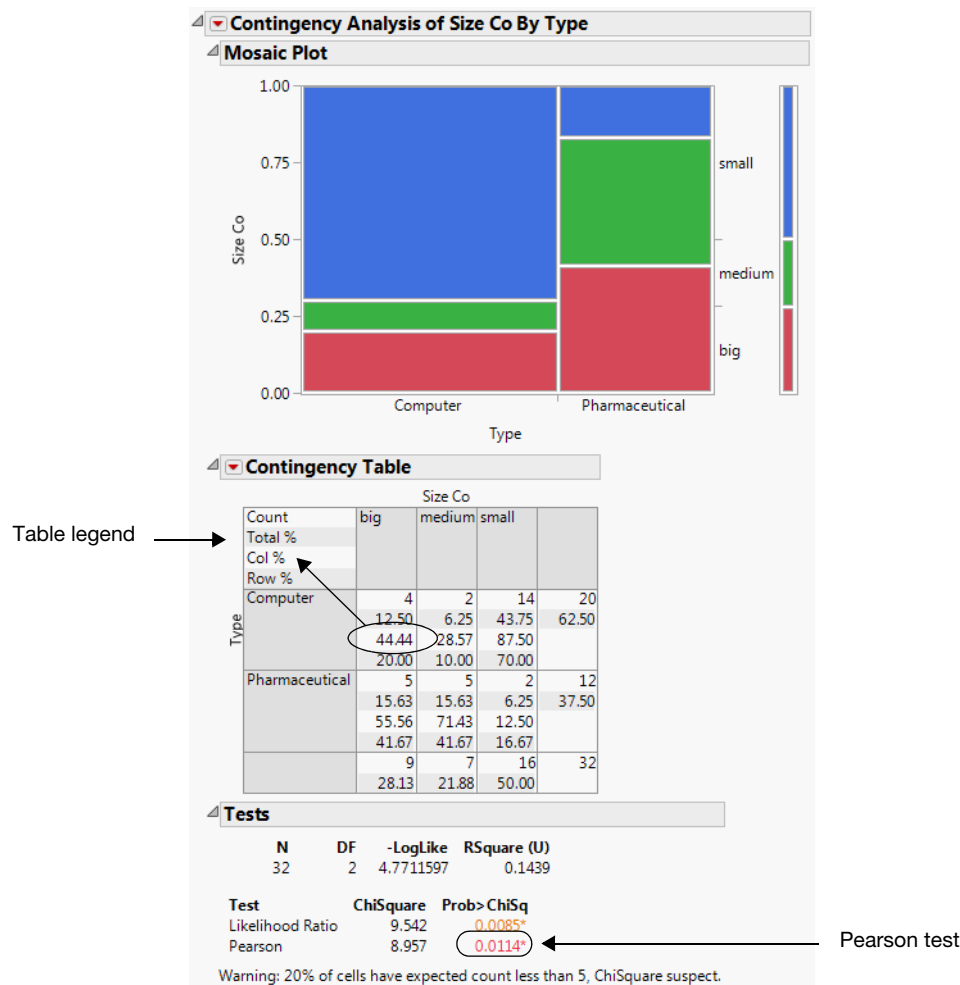
The financial analyst wants to know whether the size of a company affects profits more for one type of company than the other? However, before examining this question, the financial analyst needs to know whether the populations of computer and pharmaceutical companies consist of the same proportions of small, medium, and big companies.

Discover the Relationship

1. Select **Help > Sample Data Library** and open Companies.jmp.
2. If you still have the Companies.jmp data file open from the previous example, you might have rows that are excluded or hidden. To return the rows to the default state (all rows included and none hidden), select **Rows > Clear Row States**.

3. Select **Analyze > Fit Y by X**.
4. Select Size Co and click **Y, Response**.
5. Select Type and click **X, Factor**.
6. Click **OK**.

Figure 5.19 Company Size by Company Type



The Contingency Table contains information that is not applicable for this example. From the red triangle menu for **Contingency Table** deselect **Total %** and **Col %** to remove that information. Figure 5.20 shows the updated table.

Figure 5.20 Updated Contingency Table

Contingency Table				
Type	Size Co			
	big	medium	small	
	Count			Row %
Computer	4	2	14	20
	20.00	10.00	70.00	
Pharmaceutical	5	5	2	12
	41.67	41.67	16.67	
	9	7	16	32

Interpret the Results

The statistics in the Contingency Table are graphically represented in the Mosaic Plot. Together, the Mosaic Plot and the Contingency Table compare the percentages of small, medium, and big companies between the two industries. For example, the Mosaic Plot shows that the computer industry has a higher percentage of small companies compared to the pharmaceutical industry. The Contingency Table shows the exact statistics: 70% of computer companies are small, and about 17% of pharmaceutical companies are small.

Interpret the Test

The financial analyst has looked at only a sample of companies (the companies in the data table). The financial analyst needs to know whether the percentages differ in the broader populations of all computer and pharmaceutical companies.

To answer this question, use the p-value from the Pearson test in the **Tests** report. See Figure 5.19. Since the p-value of 0.011 is less than the significance level of 0.05, the financial analyst concludes the following:

- The differences in the sample data are not due to chance alone.
- The percentages differ in the broader population.

Now the financial analyst knows that the proportions of small, medium, and big companies are different, and can answer the question: Does the size of company affect profits more for one type of company than the other?

Compare Averages for Multiple Variables

The section “[Compare Averages for One Variable](#)” on page 151, compared averages across the levels of a categorical variable. To compare averages across the levels of two or more variables at once, use the *Analysis of Variance* technique (or ANOVA).

Scenario

The financial analyst can answer the question that we started to work through in the Comparing Proportions section, which is: Does the size of the company have a larger effect on the company's profits, based on type (pharmaceutical or computer)?

To answer this question, compare the company profits by these two variables:

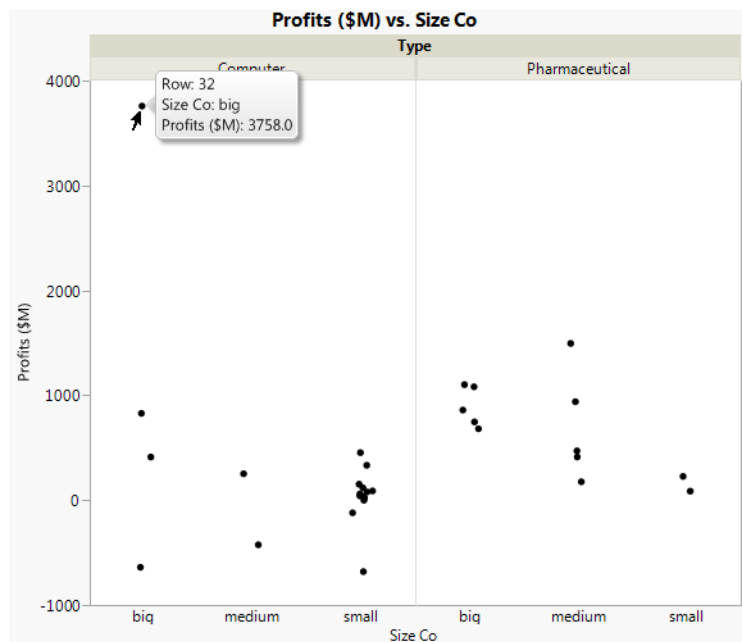
- Type (pharmaceutical or computer)
- Size (small, medium, big)

Discover the Relationship

To visualize the differences in profit for all of the combinations of type and size, use a graph:

1. Select **Help > Sample Data Library** and open Companies.jmp.
2. Select **Graph > Graph Builder**. The Graph Builder window appears.
3. Click Profits (\$M) and drag and drop it into the **Y** zone.
4. Click Size Co and drag and drop it into the **X** zone.
5. Click Type and drag and drop it into the **Group X** zone.

Figure 5.21 Graph of Company Profits



The graph shows that one big computer company has very large profits. That outlier is stretching the scale of the graph, making it difficult to compare the other data points.


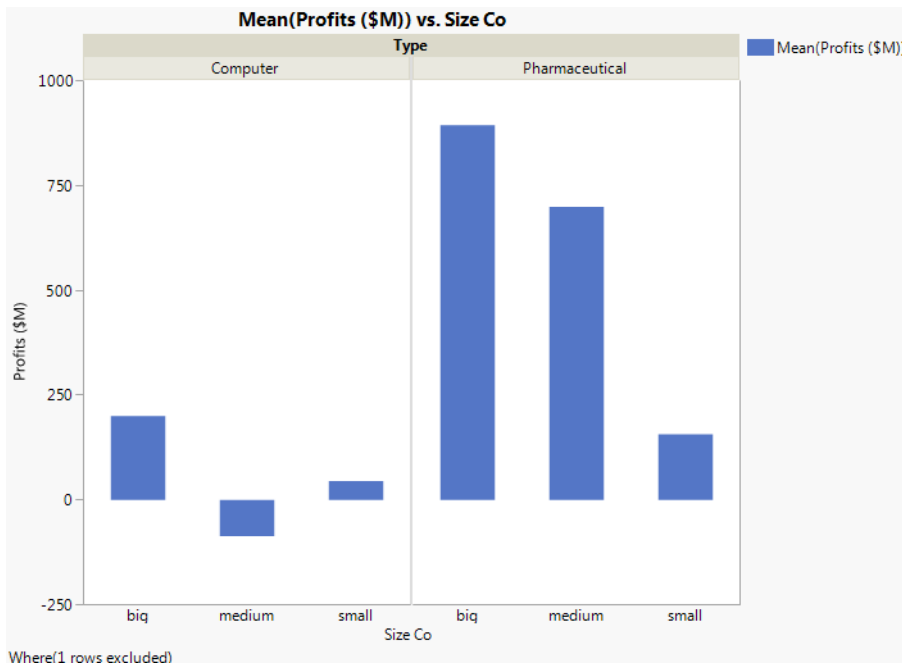
6. Select the outlier, then right-click and select **Rows > Row Exclude**. The point is removed, and the scale of the graph automatically updates.
7. Click on the Bar  icon. Comparing mean profits is easier with bar charts than with points.

Figure 5.22 Graph with Outlier Removed



The updated graph shows that pharmaceutical companies have higher average profits. The graph also shows that profits differ between company sizes for only the pharmaceutical companies. When the effect of one variable (company size) changes for different levels of another variable (company type), this is called an *interaction*.

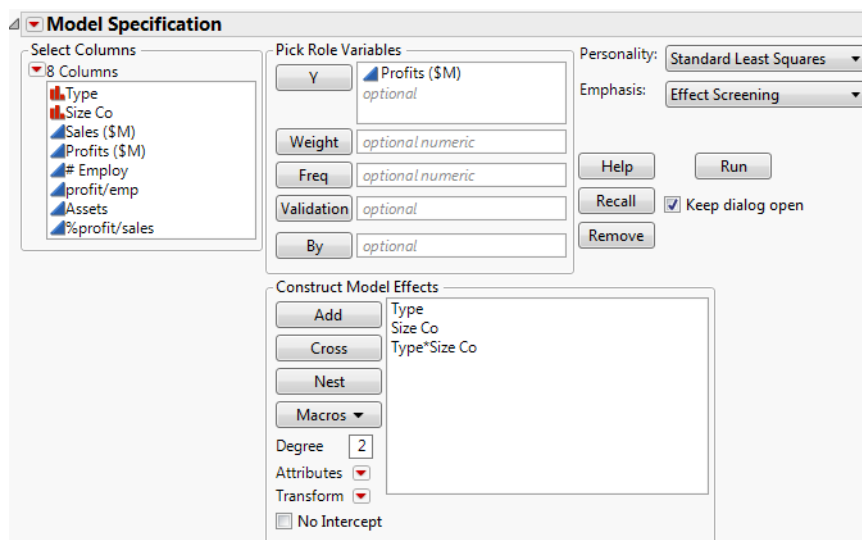
Quantify the Relationship

Because this data is only a sample, the financial analyst needs to determine the following:

- if the differences are limited to this sample and due to chance
- or
- if the same patterns exist in the broader population

1. Return to the Companies.jmp sample data table that has the data point excluded. See [“Discover the Relationship”](#) on page 157.
2. Select **Analyze > Fit Model**.
3. Select Profits (\$M) and click **Y**.
4. Select both Type and Size Co.
5. Click the **Macros** button and select **Full Factorial**.
6. From the Emphasis menu, select **Effect Screening**.
7. Select the **Keep dialog open** option.

Figure 5.23 Completed Fit Model Window



8. Click **Run**. The report window shows the model results.

To decide whether the differences in profits are real, or due to chance, examine the **Effect Tests** report.

Note: For complete details about all of the **Fit Model** results, see the Model Specification chapter in the *Fitting Linear Models* book.

View Effect Tests

The Effect Tests report (see Figure 5.24) shows the results of the statistical tests. There is a test for each of the effects included in the model on the Fit Model window: Type, Size Co, and Type*Size Co.

Figure 5.24 Effect Tests Report

Effect Tests					
Source	Nparm	DF	Sum of Squares	F Ratio	Prob > F
Type	1	1	1401847.4	10.1368	0.0039*
Size Co	2	2	724616.2	2.6198	0.0927
Type*Size Co	2	2	448061.5	1.6200	0.2180

First, look at the test for the interaction in the model: the Type*Size Co effect. Figure 5.22 showed that the pharmaceutical companies appeared to have different profits between company sizes. However, the effect test indicates that there is no interaction between type and size as it relates to profit. The p-value of 0.218 is large (greater than the significance level of 0.05). Therefore, remove that effect from the model, and re-run the model.

1. Return to the Fit Model window.
2. In the Construct Model Effects box, select the **Type*Size Co** effect and click **Remove**.
3. Click **Run**.

Figure 5.25 Updated Effect Tests Report

Effect Tests					
Source	Nparm	DF	Sum of Squares	F Ratio	Prob > F
Type	1	1	1,356,298	9.3768	0.0049*
Size Co	2	2	434,161.3	1.5008	0.2410

The p-value for the Size Co effect is large, indicating that there are no differences based on size in the broader population. The p-value for the Type effect is small, indicating that the differences that you saw in the data between computer and pharmaceutical companies is not due to chance.

Draw Conclusions

The financial analyst wanted to know whether the size of the company has a larger effect on the company's profits, based on type (pharmaceutical or computer). The financial analyst can now answer the question as follows:

- There is a real difference in profits between computer and pharmaceutical companies in the broader population.
- There is no correlation between the company's size and type and its profits.

Use Regression with Multiple Predictors

The section [“Use Regression with One Predictor”](#) on page 146 showed you how to build simple regression models consisting of one predictor variable and one response variable.

Multiple regression predicts the average response variable using two or more predictor variables.

Scenario

This example uses the Candy Bars.jmp data table, which contains nutrition information for candy bars.

A dietitian wants to predict calories using the following information:

- Total fat
- Carbohydrates
- Protein

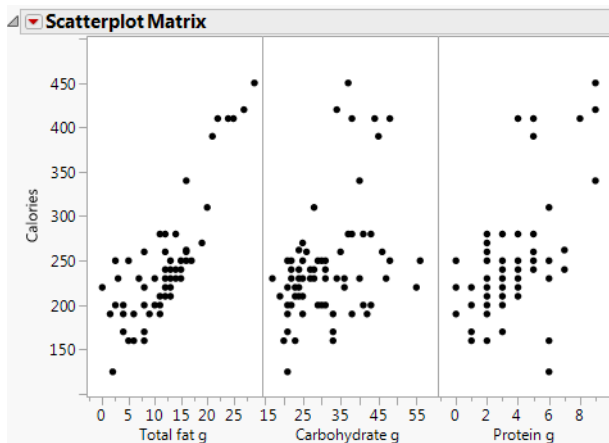
Use *multiple regression* to predict the average response variable using these three predictor variables.

Discover the Relationship

To visualize the relationship between calories and total fat, carbohydrates, and protein, create a scatterplot matrix:

1. Select **Help > Sample Data Library** and open Candy Bars.jmp.
2. Select **Graph > Scatterplot Matrix**.
3. Select Calories and click **Y, Columns**.
4. Select Total fat g, Carbohydrate g, and Protein g, and click **X**.
5. Click **OK**.

Figure 5.26 Scatterplot Matrix Results



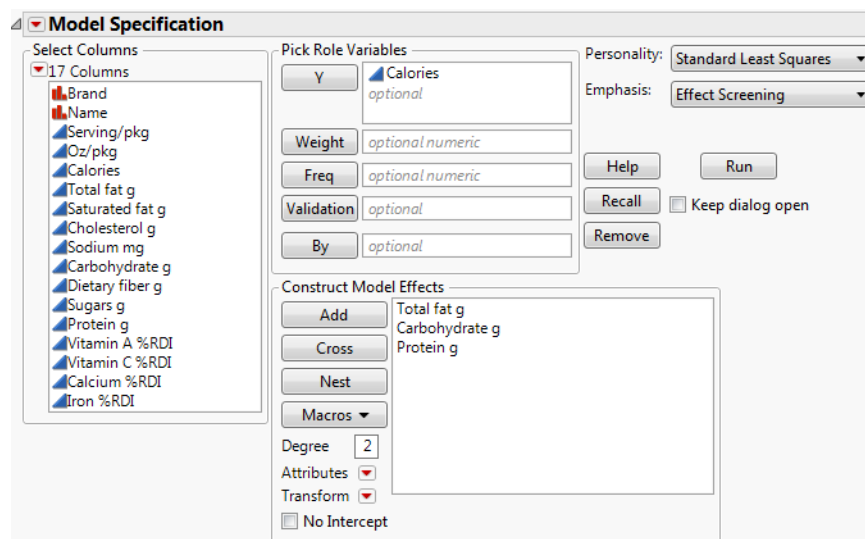
The scatterplot matrix shows that there is a positive correlation between calories and all three variables. The correlation between calories and total fat is the strongest. Now that the dietitian knows that there is a relationship, the dietitian can build a multiple regression model to predict average calories.

Build the Multiple Regression Model

Continue to use the Candy Bars.jmp sample data table.

1. Select **Analyze > Fit Model**.
2. Select Calories and click **Y**.
3. Select Total Fat g, Carbohydrate g, and Protein g and click **Add**.
4. Next to Emphasis, select **Effect Screening**.

Figure 5.27 Fit Model Window



5. Click **Run**.

The report window shows the model results. To interpret the model results, focus on these areas:

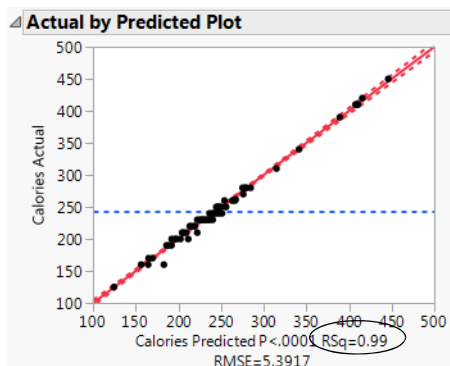
- [“View the Actual by Predicted Plot”](#) on page 163
- [“Interpret the Parameter Estimates”](#) on page 163
- [“Use the Prediction Profiler”](#) on page 164

Note: For complete details about all of the model results, see the Model Specification chapter in the *Fitting Linear Models* book.

View the Actual by Predicted Plot

The Actual by Predicted Plot shows the actual calories versus the predicted calories. As the predicted values come closer to the actual values, the points on the scatterplot fall closer around the red line. See Figure 5.28. Because the points are all very close to the line, you can see that the model predicts calories based on the chosen factors well.

Figure 5.28 Actual by Predicted Plot



Another measure of model accuracy is the RSq value (which appears below the plot in Figure 5.28). The RSq value measures the percentage of variability in calories, as explained by the model. A value closer to 1 means a model is predicting well. In this example, the RSq value is 0.99.

Interpret the Parameter Estimates

The Parameter Estimates report shows the following information:

- The model coefficients
- P-values for each parameter

Figure 5.29 Parameter Estimates Report

Model coefficients			P-values	
Parameter Estimates				
Term	Estimate	Std Error	t Ratio	Prob> t
Intercept	-5.964301	2.899986	-2.06	0.0434*
Total fat g	8.9899516	0.144981	62.01	<.0001*
Carbohydrate g	4.097505	0.071025	57.69	<.0001*
Protein g	4.4013313	0.39785	11.06	<.0001*

In this example, the p-values are all very small (<.0001). This indicates that all three effects (fat, carbohydrate, and protein) contribute significantly when predicting calories.

You can use the model coefficients to predict the value of calories for particular values of fat, carbohydrate, and protein. For example, suppose that you want to predict the average calories for any candy bar that has these characteristics:

- Fat = 11 g
- Carbohydrate = 43 g
- Protein = 2 g

Using these values, you can calculate the predicted average calories as follows:

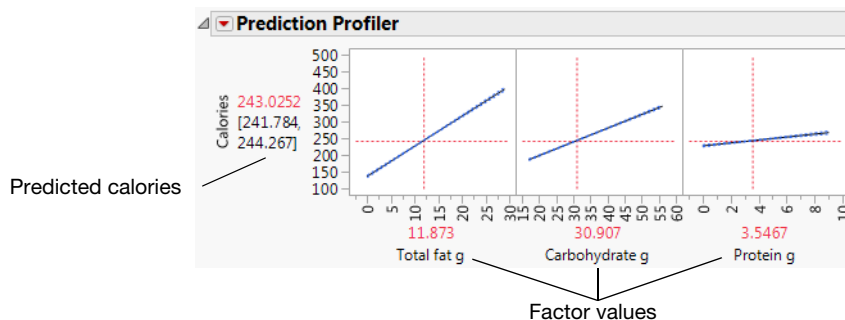
$$277.92 = -5.9643 + 8.99 \times 11 + 4.0975 \times 43 + 4.4013 \times 2$$

The characteristics in this example are the same as the Milky Way candy bar (on row 59 of the data table). The actual calories for the Milky Way are 280, showing that the model predicts well.

Use the Prediction Profiler

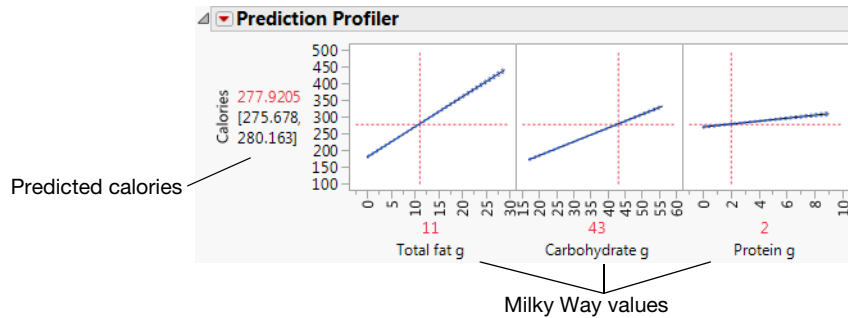
Use the Prediction Profiler to see how changes in the factors affect the predicted values. The profile lines show the magnitude of change in calories as the factor changes. The line for Total fat g is the steepest, meaning that changes in total fat have the largest effect on calories.

Figure 5.30 Prediction Profiler



Click and drag the vertical line for each factor to see how the predicted value changes. You can also click the current factor values and change them. For example, click on the factor values and type the values for the Milky Way candy bar (row 59).

Figure 5.31 Factor Values for the Milky Way



Note: For complete details about the Prediction Profiler, see the Profiler chapter in the *Profilers* book.

Draw Conclusions

The dietitian now has a good model to predict calories of a candy bar based on its total fat, carbohydrates, and protein.

Chapter 6

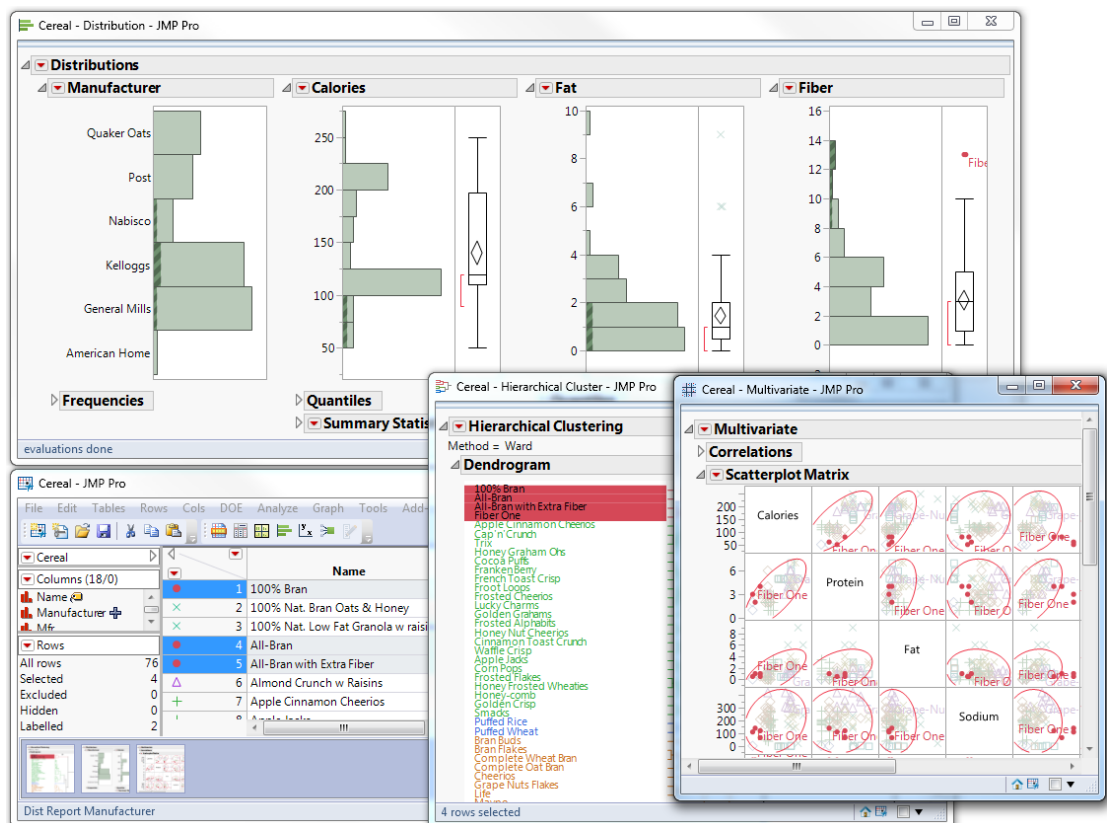
The Big Picture

Exploring Data in Multiple Platforms

JMP provides a host of statistical discovery platforms to help you explore different aspects of your data. You might start with a simple look at individual variables in histograms and then progress to multivariate and cluster analyses to get a deeper look. Each step of the way, you learn more about your data.

This chapter steps through an analysis of the Cereal.jmp sample data table that is installed with JMP. You learn how to explore the data in the Distribution, Multivariate, and Hierarchical Clustering platforms.

Figure 6.1 Linked Analyses in JMP



Contents

Fun Fact: Linked Analyses 169

Explore Data in Multiple Platforms 169

 Analyze Distributions..... 169

 Analyze Patterns and Relationships..... 173

 Analyze Similar Values 177

Fun Fact: Linked Analyses

One of the powerful features in JMP is its linked analyses. The graphs and reports that you create are linked to each other through the data table. As shown in Figure 6.1, data that are selected in the data table are also selected in the three report windows. The linked analyses enable you to select data in one window and see where it occurs in the other windows. As you work through the examples in this chapter, keep the JMP windows open to see these interactions yourself.

Explore Data in Multiple Platforms

Which cereals are part of a healthy diet? The *Cereal.jmp* sample data (real data gathered from boxes of popular cereals) provides statistics on fiber content, calories, and other nutritional information. To identify the most healthful cereals, you step through interpreting histograms and descriptive statistics, correlations and outlier detection, scatterplots, and cluster analysis.

Analyze Distributions

The Distribution platform illustrates the distribution of a single variable (*univariate* analysis) using histograms, additional graphs, and reports. The word *univariate* simply means involving one variable instead of two (bivariate) or many (multivariate). However, you can examine the distribution of several individual variables within a single report. The report content for each variable changes depending on whether the variable is categorical (nominal or ordinal) or continuous.

- For categorical variables, the initial graph is a histogram. The histogram shows a bar for each level of the ordinal or nominal variable. The reports show counts and proportions.
- For continuous variables, the initial graphs show a histogram and an outlier box plot. The histogram shows a bar for grouped values of the continuous variable. The reports show selected quantiles and summary statistics.

Once you know how your data are distributed, you can plan the appropriate type of analysis going forward.

Note: For details about the Distribution platform, see the Distributions chapter in the *Basic Analysis* book.

Scenario

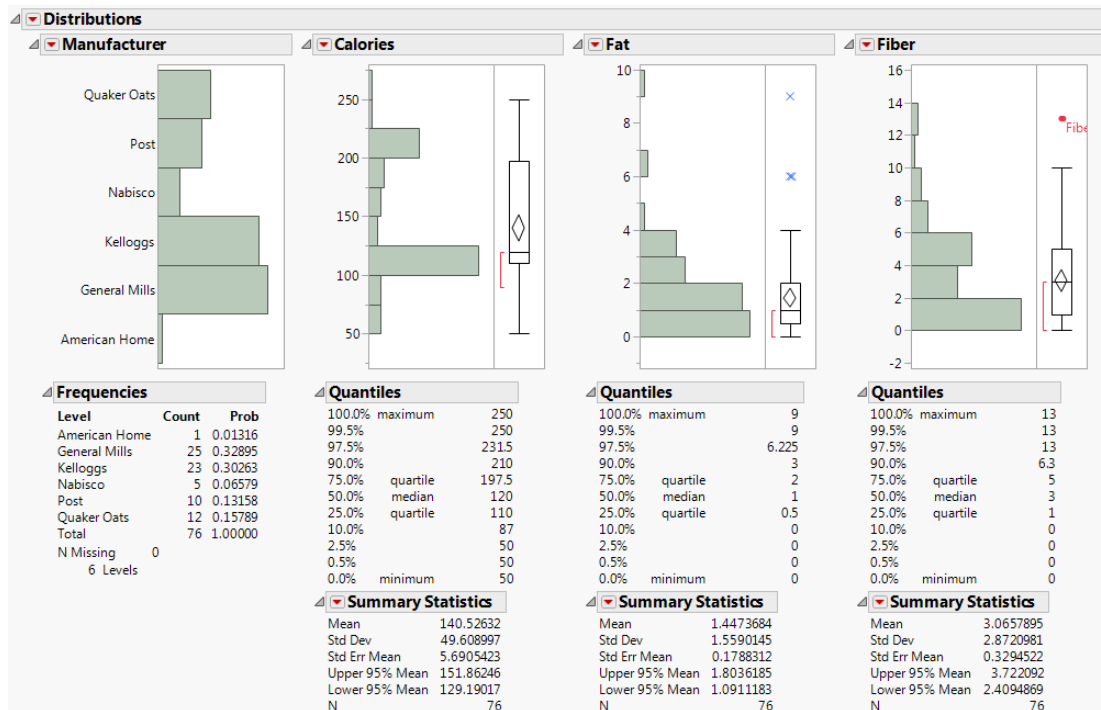
You want to view the nutritional values of cereals so that you can eat a more healthful diet. Analyzing distributions of cereal data reveals answers to the following questions:

- Which cereals are highest in fiber?
- What is the average, minimum, and maximum number of calories?
- What is the median amount of fat?
- Which cereal contains the most fat?
- Are there any outliers in the data?

Create the Distributions

1. Select **Help > Sample Data Library** and open Cereal.jmp.
2. Select **Analyze > Distribution**.
3. Press Ctrl and click Manufacturer, Calories, Fat, and Fiber.
4. Click **Y, Columns** and then click **OK**.

Figure 6.2 Distributions for Manufacturer, Calories, Fat, and Fiber



In the Fiber distributions, notice the following:

- Fiber One and All-Bran with Extra Fiber contain the most fiber as shown in the Fiber box plot. These cereals are outliers in terms of fiber content.

The row that contains Fiber One in Cereal.jmp is labeled. This label shows the name of the cereal next to a data point in graphs. To see the entire label, drag the right-most vertical border to the right. Place your cursor over the unlabeled data point to see “All Bran with Extra Fiber”.

In the Fat distributions, notice the following:

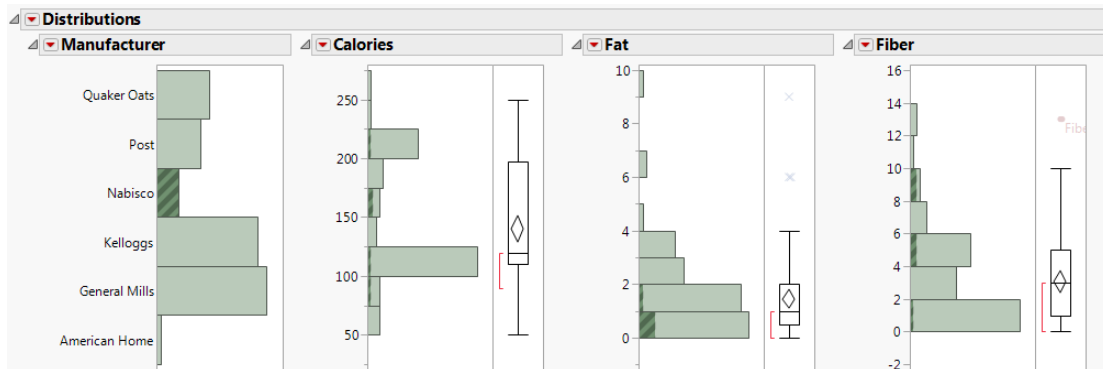
- Place your cursor over the top data point (the x marker) in the Fat box plot to see that 100% Nat. Bran Oats & Honey is the highest in fat.
- In the Fat Quantiles report, the median amount of fat is 1 gram.

In the Calories Quantiles report, notice the following:

- The maximum number of calories is 250.
- The minimum number of calories is 50.

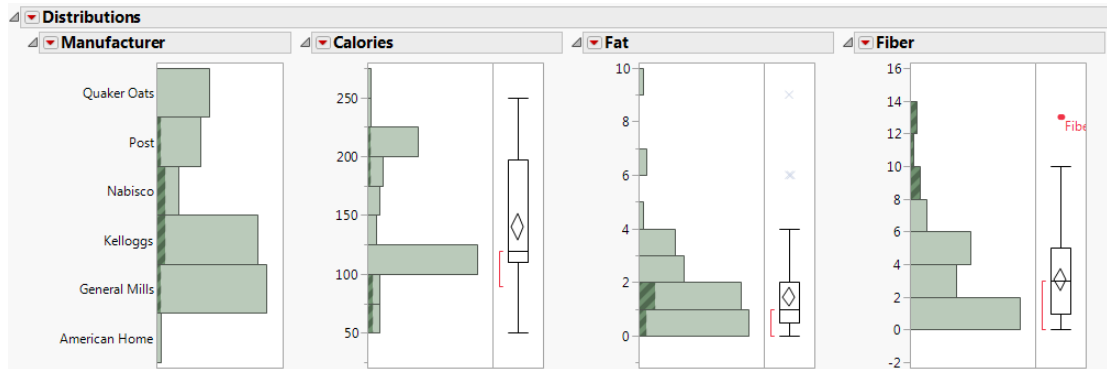
5. In the Manufacturer histogram, click on the bar for Nabisco.

Figure 6.3 Distributions for Nabisco Cereals



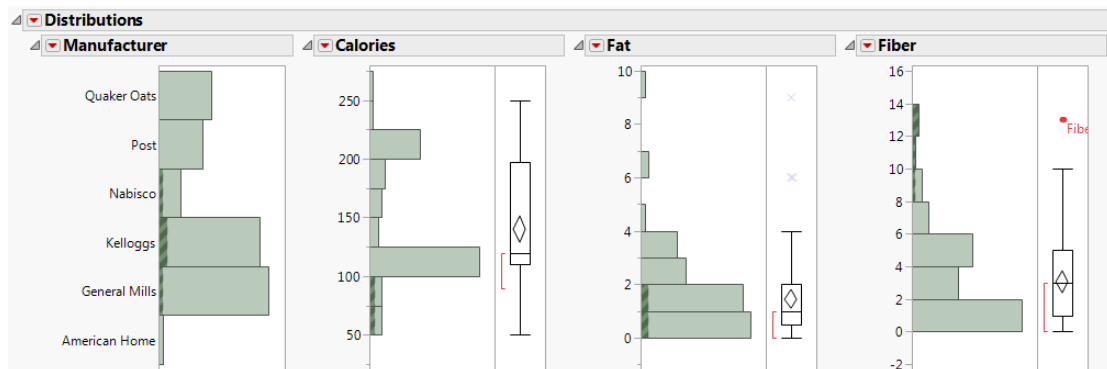
The Calories, Fat, and Fiber distributions for Nabisco cereals are highlighted in the other histograms. You can view the Calories, Fat, and Fiber distributions for the Nabisco cereals relative to the Calories, Fat, and Fiber distributions for the overall data. For example, the Fat distribution of Nabisco cereals seems to be lower than the Fat distribution for the overall data.

6. Click below the last Fiber bar to deselect all bars.
7. Press Shift and, in the Fiber histogram, click all histogram bars with a value above 8.

Figure 6.4 High-Fiber Cereals


The highest-fiber cereals are highlighted in the Calories and Fat histograms. Because the histograms are linked, note that some of the high-fiber cereals are also low in fat.

8. Press Ctrl and Shift and deselect the two Calories histogram bars that are at or near 200. High calorie cereals are eliminated from the histograms.

Figure 6.5 High-Fiber and Low-Calorie Cereals


Tip: Leave the Distributions report open. You will use it later in a cluster analysis. See [“Analyze Similar Values”](#) on page 177.

Interpret the Results

Looking at the results, you can answer the following questions:

Which cereals are highest in fiber? The Fiber box plot shows that All-Bran with Extra Fiber and Fiber One have the highest amount of fiber. These two cereals are outliers.

What is the average, minimum, and maximum number of calories? The Calories histogram shows that the number of calories ranges from 50 to 275. The Calories Quantiles show that the number of calories ranges from 50 to 250, and the median number of calories is 120. The distribution is not uniform.

What is the median amount of fat? The Fat Quantiles report shows that the median amount of fat is 1 gram.

Which cereal contains the most fat? The Fat box plot shows that 100% Nat. Bran Oats & Honey is the highest in fat. This cereal is an outlier.

Draw Conclusions

To increase the amount of fiber in your diet, you decide to try All-Bran with Extra Fiber and Fiber One. These cereals are lower in calories and fat. Most cereals do not greatly increase the amount of fat in your diet, but you plan to avoid the high fat 100% Nat. Bran Oats & Honey. And although most cereals are relatively low in fat, they are not necessarily low in calories.

Analyze Patterns and Relationships

Now that you have identified which cereals to eat or avoid, you want to see how the cereal variables relate to each other. The Multivariate platform enables you to observe patterns and relationships between variables. From the Multivariate report, you can do the following:

- summarize the strength of the linear relationships between each pair of response variables using the Correlations table
- identify dependencies, outliers, and clusters using the Scatterplot Matrix
- use other techniques to examine multiple variables, such as partial, inverse, and pairwise correlations, covariance matrices, and principal components

Note: For details about the Multivariate platform, see the Correlations and Multivariate Techniques chapter in the *Multivariate Methods* book.

Scenario

You want to see the relationships between variables such as fat and calories. Analyzing the cereal data in the Multivariate platform reveals answers to the following questions:

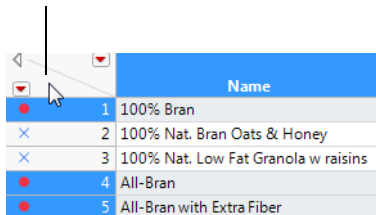
- Which pairs of variables are highly correlated?
- Which pairs of variables are not correlated?

Create the Multivariate Report

1. In the Cereal.jmp data table, click the bottom triangle at the top of the Columns panel to deselect the rows (Figure 6.6).

Figure 6.6 Deselecting Rows

Click here to deselect the rows.



2. Select **Analyze > Multivariate Methods > Multivariate**.
3. Select Calories through Potassium, click **Y, Columns**, and then click **OK**.

The Multivariate report appears. The report contains the Correlations report and Scatterplot Matrix by default. The Correlations report is a matrix of correlation coefficients that summarizes the strength of the linear relationships between each pair of response (Y) variables. The dark numbers indicate a lower degree of correlation.

Figure 6.7 Correlations Report

Correlations											
	Calories	Protein	Fat	Sodium	Fiber	Complex Carbo	Tot Carbo	Sugars	Calories fr Fat	Potassium	
Calories	1.0000	0.7041	0.6460	0.1996	0.1953	0.6688	0.9076	0.5060	0.6709	0.4451	
Protein	0.7041	1.0000	0.4080	0.0050	0.5470	0.6486	0.6937	-0.0010	0.4288	0.6782	
Fat	0.6460	0.4080	1.0000	-0.0768	0.1824	0.1037	0.3860	0.4148	0.9013	0.3420	
Sodium	0.1996	0.0050	-0.0768	1.0000	-0.0448	0.2619	0.3066	0.1767	0.0572	0.0459	
Fiber	0.1953	0.5470	0.1824	-0.0448	1.0000	0.1769	0.3668	-0.1264	0.2553	0.8326	
Complex Carbo	0.6688	0.6486	0.1037	0.2619	0.1769	1.0000	0.7773	-0.1601	0.1558	0.2693	
Tot Carbo	0.9076	0.6937	0.3860	0.3066	0.3668	0.7773	1.0000	0.4263	0.4636	0.5375	
Sugars	0.5060	-0.0010	0.4148	0.1767	-0.1264	-0.1601	0.4263	1.0000	0.4369	0.1166	
Calories fr Fat	0.6709	0.4288	0.9013	0.0572	0.2553	0.1558	0.4636	0.4369	1.0000	0.3694	
Potassium	0.4451	0.6782	0.3420	0.0459	0.8326	0.2693	0.5375	0.1166	0.3694	1.0000	

Note the following:

- In the Calories column, the number of calories is highly correlated with all variables except for sodium and fiber.
- In the Fiber column, fiber and potassium appear to be highly correlated.
- In the Sodium column, sodium is not highly correlated with the other variables.

The density ellipses in the Scatterplot Matrix further illustrates relationships between variables. Figure 6.8 shows a portion of the plot.

[illegible]

Note the following:

- The ellipses are fairly round in the Sodium row. This shape indicates that Sodium is uncorrelated with other variables.
- The blue x markers, which represent Nat. Bran Oats & Honey, Cracklin' Oat Bran, and Banana Nut Crunch, appear outside the ellipses in the Fat row. This placement indicates that the datum is an outlier (because of the amount of fat in the cereal).

You will further explore a scatterplot matrix later.

4. Select **Pairwise Correlations** from the Multivariate red triangle menu to display the Pairwise Correlations report.

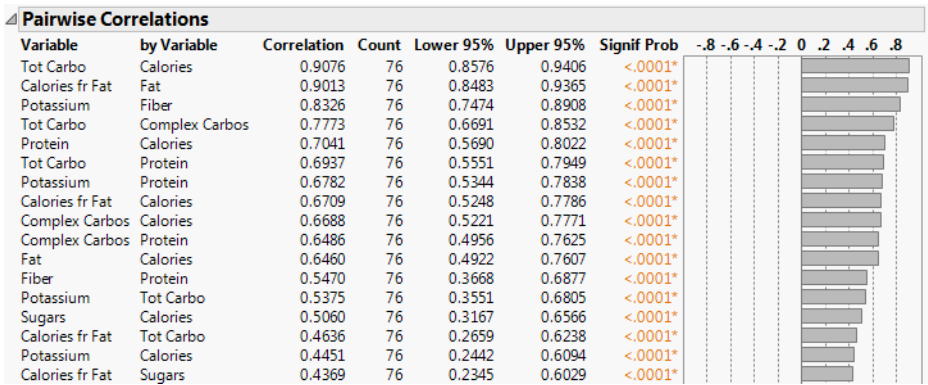
Variable	by Variable	Correlation	Count	Lower 95%	Upper 95%	Signif Prob
Protein	Calories	0.7041	76	0.5690	0.8022	<.0001*
Fat	Calories	0.6460	76	0.4922	0.7607	<.0001*
Fat	Protein	0.4080	76	0.2011	0.5801	0.0003*
Sodium	Calories	0.1996	76	-0.0270	0.4068	0.0838
Sodium	Protein	0.0050	76	-0.2207	0.2302	0.9655
Sodium	Fat	-0.0768	76	-0.2971	0.1513	0.5099
Fiber	Calories	0.1953	76	-0.0315	0.4030	0.0908
Fiber	Protein	0.5470	76	0.3668	0.6877	<.0001*
Fiber	Fat	0.1824	76	-0.0449	0.3918	0.1147
Fiber	Sodium	-0.0448	76	-0.2675	0.1825	0.7010

The Pairwise Correlations report lists the Pearson product-moment correlations for each pair of Y variables. The report also shows significance probabilities and compares the correlations in a bar chart.

- 5. To quickly see which pairs are highly correlated, right-click in the report and select the **Sort by Column, Signif Prob, Ascending** check box, and then click **OK**.

The most highly correlated pairs appear at the top of the report. The small *p*-values for the pairs indicate evidence of correlation. The most significant correlation is between Tot Carbo (total carbohydrates) and Calories.

Figure 6.10 Small *P*-values for Pairs



Interpret the Results

Looking at the results, you can answer the following questions:

Which pairs of variables are highly correlated? The Correlations report and Scatterplot Matrix show that the number of calories is highly correlated with all variables except for sodium and fiber. The Pairwise Correlations report shows that Tot Carbo (total carbohydrates) and Calories is the most correlated pair of variables.

Which pairs of variables are not correlated? The Correlations report and Scatterplot Matrix show that Sodium is not correlated with the other variables.

Draw Conclusions

You confirm the previous decision to avoid the high fat 100% Nat. Bran Oats & Honey. Trying All-Bran with Extra Fiber and Fiber One was also a smart decision. These two high-fiber cereals have the added benefit of contributing a lower number of calories, fat, and sugars and a higher amount of potassium. You also decide to avoid high-carbohydrate cereals because they likely contain a large number of calories.

Analyze Similar Values

Clustering is a multivariate technique that groups observations together that share similar values across a number of variables. Hierarchical clustering combines rows in a hierarchical sequence that is portrayed as a tree. Cereals with certain characteristics, such as high fiber, are grouped in clusters so that you can view similarities among cereals.

Note: For details about hierarchical clustering, see the Hierarchical Cluster chapter in the *Multivariate Methods* book.

Scenario

You want to know which cereals are similar to each other and which ones are dissimilar. Analyzing clusters of cereal data reveals answers to the following questions:

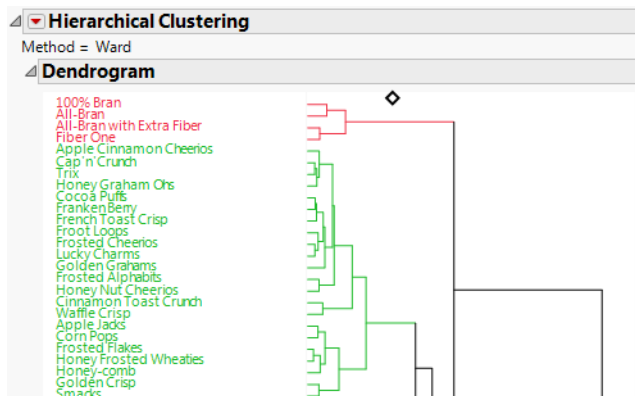
- Which cluster of cereals provides little nutritional value?
- Which cluster of cereals is high in vitamins and minerals and contains a low amount of sugar and fat?
- Which cluster of cereals is high in fiber and low in calories?

Create the Hierarchical Cluster Graph

1. With Cereal.jmp displayed, select **Analyze > Clustering > Hierarchical Cluster**.
2. Select Calories through Enriched, click **Y, Columns**, and then click **OK**.

The Hierarchical Clustering report appears. Figure 6.11 shows a portion of the report. The clusters are colored according to the data table row states.

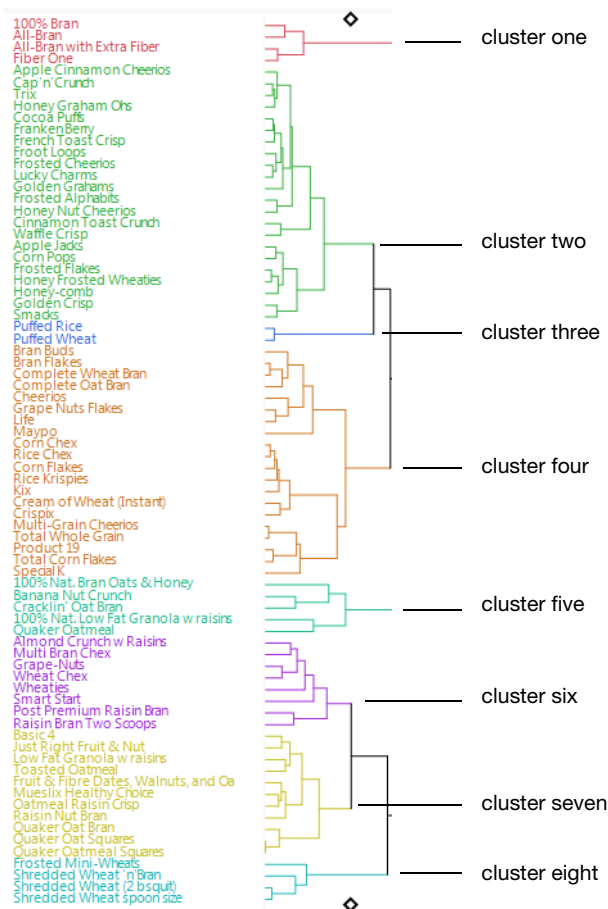
Figure 6.11 Portion of the Hierarchical Clustering Report



3. Select **Color Clusters** from the Hierarchical Clustering red triangle menu.

The clusters are colored according to their relationships in the dendrogram.

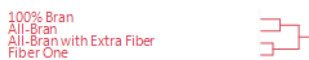
Figure 6.12 Colored Clusters



The cereals have similar characteristics within each cluster. For example, judging by the names of the cereals in cluster one, you guess that the cereals are high in fiber.

Notice how All-Bran with Extra Fiber and Fiber One are grouped in cluster one. These cereals are more similar to each other than the other two cereals in the cluster.

Figure 6.13 Similar Cereals in Cluster One



4. To select cluster one, click on the red horizontal line on the right.

The four cereals are highlighted in red.

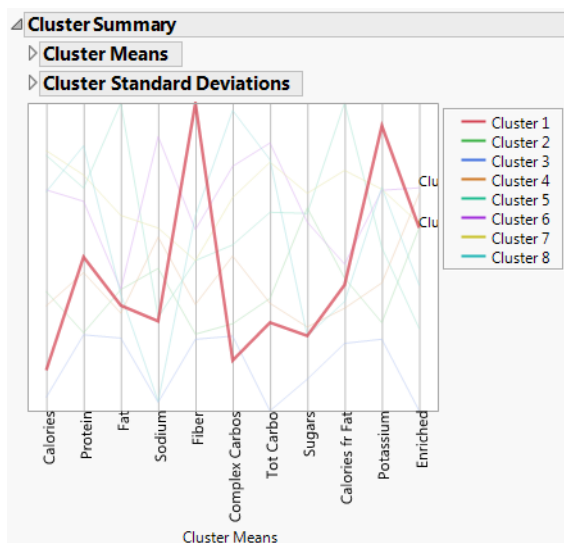
Figure 6.14 Selecting a Cluster



- To see the similar characteristics in the cluster, select **Cluster Summary** from the red triangle menu.

The Cluster Summary graph at the bottom of the report shows the mean value of each variable across each cluster. For example, the cereals in this cluster contain more fiber and potassium than cereals in other clusters.

Figure 6.15 Cluster Summary

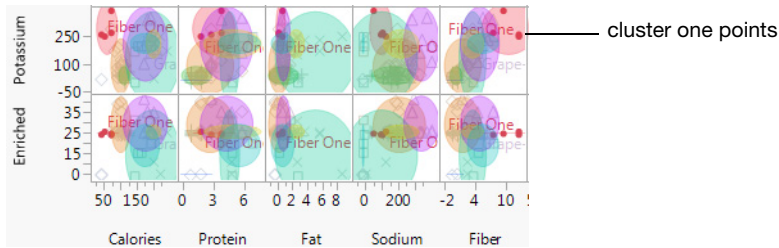


- Select Scatterplot Matrix from the red triangle menu.

This option is an alternative to creating a scatterplot matrix in the Multivariate platform.

Note the Fiber plot in the Potassium row. The selected cereals are located on the right side of the plot between 8 and 13 grams. This location indicates that the cereals in cluster one are high in fiber and potassium.

Figure 6.16 Cluster One Characteristics



Note: The points are also selected in the previous scatterplot matrix that you created if it is still open.

Interpret the Results

Clicking through the clusters and looking at the Cluster Summary report, you can see the following characteristics:

- Cluster one cereals, such as Fiber One and All-Bran, contain high fiber and potassium and low calories.
- Cluster two cereals, which contain many favorite children's cereals, are high in sugar and low in fiber, complex carbohydrates, and protein.
- Cluster three cereals (Puffed Rice and Puffed Wheat) are low in calories but provide little nutritional value.
- Cluster four cereals, such as Total Corn Flakes and Multi-Grain Cheerios, provide 100% of your daily requirement of vitamins and minerals. They are low in fat, fiber, and sugar.
- Cluster five cereals are high in protein and fat and low in sodium. The cluster consists of cereals such as Banana Nut Crunch and Quaker Oatmeal.
- Cluster six cereals are low in fat and high in sodium and carbohydrates. Traditional cereals such as Wheaties and Grape-Nuts are in this cluster.
- Cluster seven cereals are high in calories and low in fiber. Many cereals that include dried fruit are in this cluster (Mueslix Healthy Choice, Low Fat Granola w Raisins, Oatmeal Raisin Crisp, Raisin Nut Bran, and Just Right Fruit & Nut).
- Cluster eight cereals are low in sodium and sugar, and high in complex carbohydrates, protein, and potassium. Shredded Wheat and Mini-Wheat cereals are in this cluster.

By looking at the joins in the dendrogram, you can see which cereals in each cluster are most similar.

- In cluster one, Fiber One is similar in nutritional value to All-Bran with Extra Fiber. 100% Bran and All-Bran are also similar. Each pair of similar cereals are made by different companies, so the cereals are competing against each other.

- In cluster two, Frosted Flakes and Honey Frosted Wheaties are similar even though one is a corn flake and the other is a wheat flake. Lucky Charms and Frosted Cheerios are similar. Cap'n'Crunch and Trix are also similar.

Draw Conclusions

Based on your desire to eat more fiber and fewer calories, you decide to try the cereals in cluster one. You will avoid cereals in cluster three, which consists of puffed wheat and puffed rice and have little nutritional value. And you will try cereals in the highly nutritious cluster four.

Chapter 7

Save and Share Your Work

Save and Re-create Results

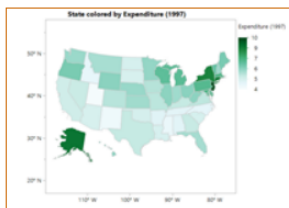
Once you have generated results from your data, JMP provides you with multiple ways to share your work with others. Here are some of the ways that you can share your work:

- Saving platform results as journals, projects, or web reports
- Saving results, data tables, and other files in projects
- Saving scripts to reproduce results in data tables
- Creating Adobe Flash (SWF) versions of platform results
- Saving results as Interactive HTML (.htm, html)
- Saving results as a PowerPoint presentation (.pptx)
- Sharing results in a dashboard

Figure 7.1 Example of a Web Report

SAT By Year Reports

Click any of the thumbnails below to launch interactive JMP reports.



09/28/2017 10:04 AM

Graph Builder State Colored by Expenditure

Place your cursor over the state to see the mean expenditure.



09/28/2017 10:04 AM

Bubble Plot of SAT Verbal by SAT Math Sized by % Taking (2004) Across Year ID State

Click the Play button to animate the plot

Contents

Save Platform Results in Journals	185
Example of Creating a Journal	185
Add Analyses to a Journal	186
Create a Project	186
Save Files in a Project	190
Move Files to a Project	192
Share the Project	194
Save and Run Scripts	194
Example of Saving and Running a Script	195
About Scripts and JSL	196
Save Reports as Interactive HTML	196
Interactive HTML Contains Data	197
Example of Creating Interactive HTML	198
Create a Web Report	199
Save a Report as a PowerPoint Presentation	201
Create Adobe Flash Versions of the Profiler, Bubble Plot, or Distribution Platforms	202
Example of Saving an Adobe Flash Version of a Bubble Plot	202
Create Dashboards	204
Example of Combining Windows	205
Example of Creating a Dashboard with Two Reports	206

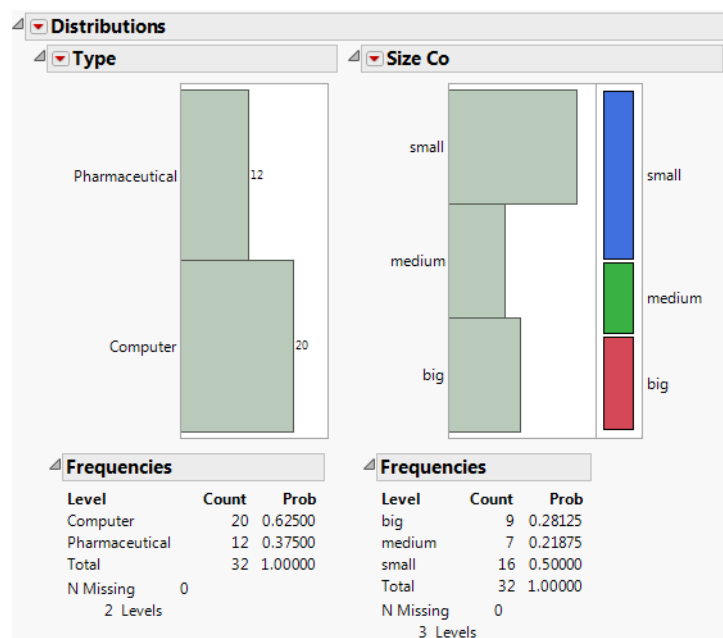
Save Platform Results in Journals

Save platform reports for future viewing by creating a journal of the report window. The journal is a copy of the report window. You can edit or append additional reports to an existing journal. The journal is not connected to the data table. A journal is an easy way to save the results from several report windows in a single report window that you can share with others.

Example of Creating a Journal

1. Select **Help > Sample Data Library** and open **Companies.jmp**.
2. Select **Analyze > Distribution**.
3. Select both **Type** and **Size Co** and click **Y, Columns**.
4. Click **OK**.
5. From the red triangle menu for **Type**, select **Histogram Options > Show Counts**.
6. From the red triangle menu for **Size Co**, select **Mosaic Plot**.
7. Select **Edit > Journal** to journal these results. The results are duplicated in a journal window.

Figure 7.2 Journal of Distribution Results



The results in the journal are not connected to the data table. In the Type bar chart, if you click the Computer bar, no rows are selected in the data table.

Since the journal is a copy of your results, most of the red triangle menus do not exist. A journal does have a red triangle menu for each new report that you add to the journal. This menu has two options:

Rerun in new window If you have the original data table that was used to create the original report, this option runs the analysis again. The result is a new report window.

Edit Script This option opens a script window that contains a JSL script to re-create the analysis. JSL is a more advanced topic that is covered in the *Scripting Guide* and *JSL Syntax Reference*.

Add Analyses to a Journal

If you perform another analysis, you can add the results of the analysis to the existing journal.

1. With a journal open, select **Analyze > Distribution**.
2. Select profit/emp and click **Y, Columns**.
3. Click **OK**.
4. Select **Edit > Journal**. The results are appended to the bottom of the journal.

Create a Project

A JMP project provides a way to organize files that you use in an analysis. You can add JMP files (reports, data tables, scripts, journals, and so on) and non JMP files, such as Microsoft PowerPoint or Adobe PDF files. After you perform an analysis, the graph and report appear in the project on tabs. You can display multiple reports or graphs, display the log, and run scripts from the project. Reports and graphs remain linked to the data table.

Projects help you avoid overlapping or concealed JMP windows. You can maximize the project window to have a bigger view of the project. And when you save the project, the state of the project is saved (for example, the open reports and the layout of the window).

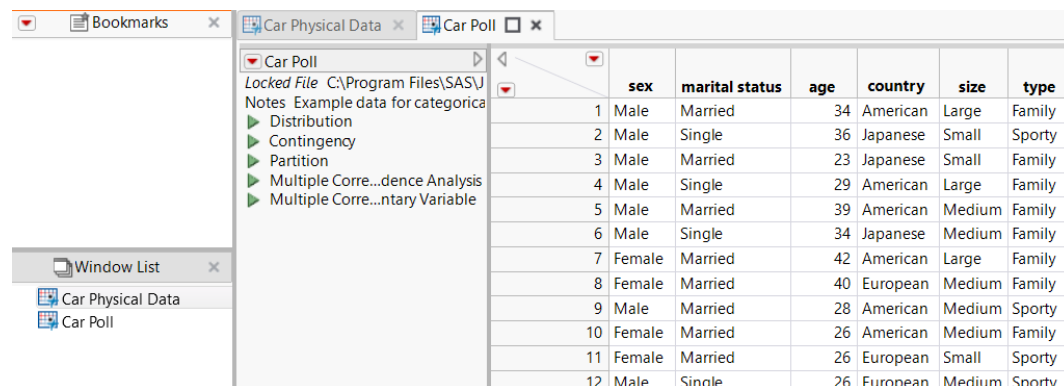
Create a New Project and Add Data Tables

1. To start a new project, select **File > New > Project**. A window appears that shows the untitled project.
2. Select **File > Open** and browse to the JMP Samples/Data folder.
3. Press Shift and click Car Physical Data.jmp and Car Poll.jmp.
4. Click **Open**.

The data tables are added to the project.

5. Select **File > Save Project** and save the file as **Cars.jmpprj**.

Figure 7.3 Initial Project



	sex	marital status	age	country	size	type
1	Male	Married	34	American	Large	Family
2	Male	Single	36	Japanese	Small	Sporty
3	Male	Married	23	Japanese	Small	Family
4	Male	Single	29	American	Large	Family
5	Male	Married	39	American	Medium	Family
6	Male	Single	34	Japanese	Medium	Family
7	Female	Married	42	American	Large	Family
8	Female	Married	40	European	Medium	Family
9	Male	Married	28	American	Medium	Sporty
10	Female	Married	26	American	Medium	Family
11	Female	Married	26	European	Small	Sporty
12	Male	Single	26	European	Medium	Sporty

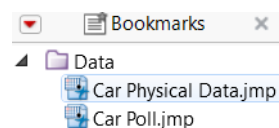
Tip: When you open a file using **File > Open**, and a project is active, the file will be added to the project automatically.

Add and Organize Bookmarks

Bookmarks provide quick access to files that you regularly open.

1. In the Window List, right-click one of the file names and select **Bookmark All**.
You can also drag the files from the Window List to the Bookmarks pane.
2. Create a group for the data tables by selecting **New Group** from the Bookmarks red triangle menu. The group organizes the files but does not move them on your computer.
3. Enter **Data** and click **OK**.
4. In the Bookmarks pane, select the Car data tables and drag them into the Data group.

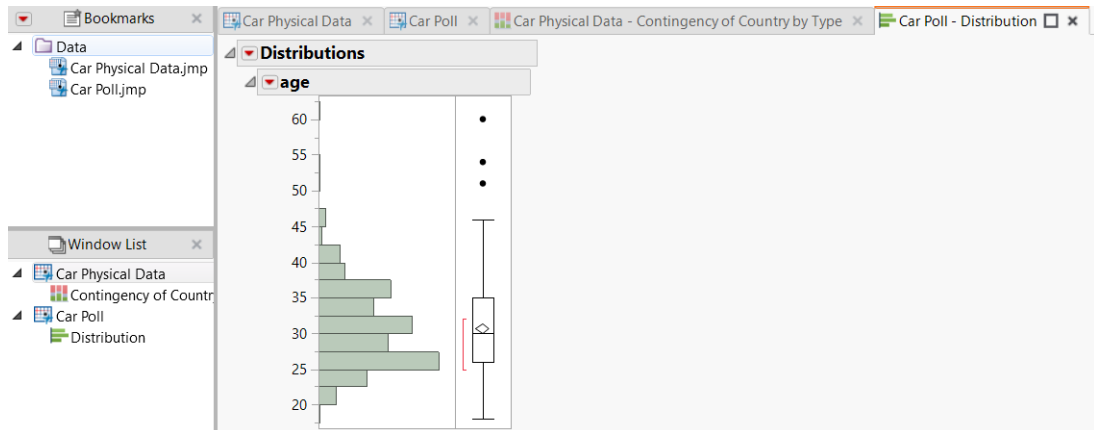
Figure 7.4 Bookmarked Files



Rearrange Reports

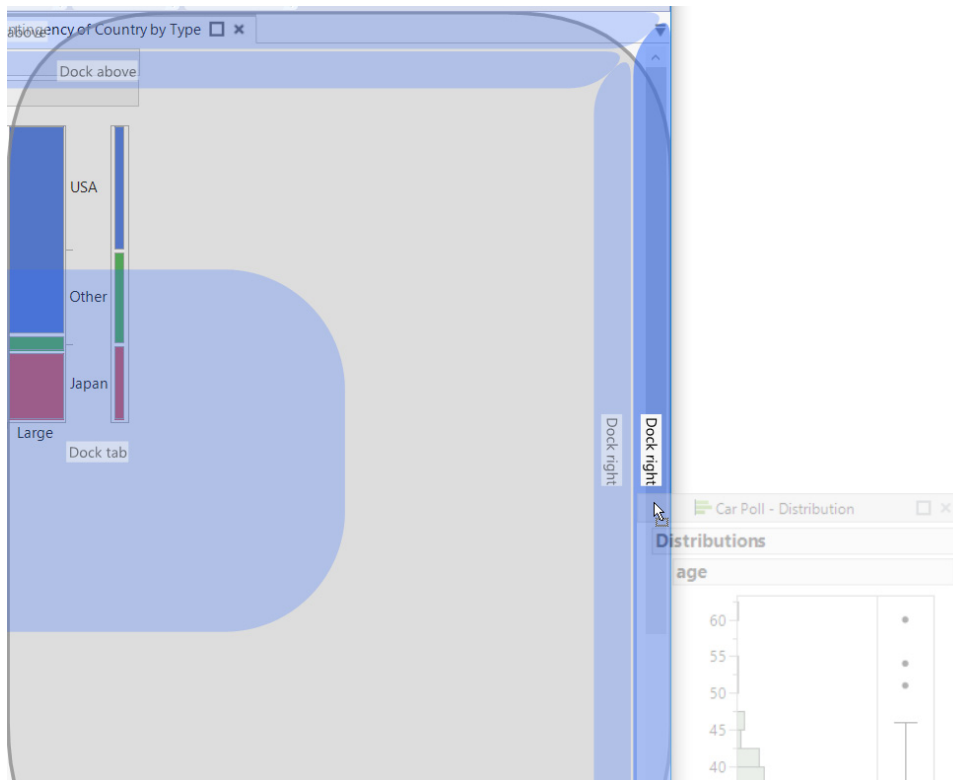
You can rearrange reports to display them in a separate pane. As you drag the report tab, zones appear that show you where to place, or “dock”, the report.

1. In Car Physical Data.jmp, run the Contingency script.
 2. In Car Poll.jmp, run the Distribution script.
- The data tables and reports appear in tabs.

Figure 7.5 Tabbed Reports

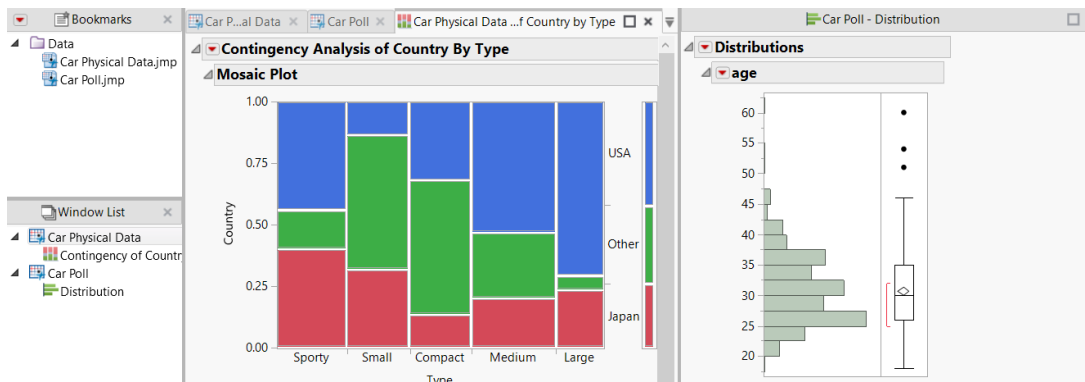
3. Drag the Distribution report tab to the right until the *Dock right* zone appears on the far right.

Figure 7.6 Dragging a Report



4. Drop the report into the *Dock right* zone.
The tab appears in a new pane.

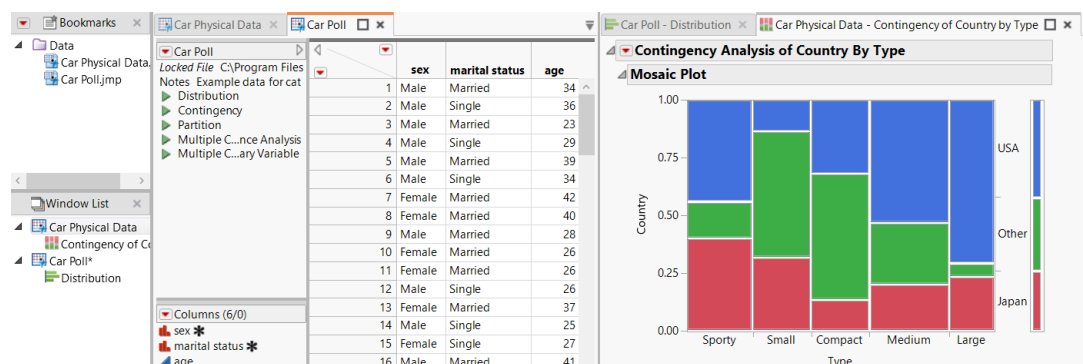
Figure 7.7 Report in New Pane



5. Drag the Contingency tab to the middle of the Distribution report.

6. When the *Dock tab* zone appears, drop the report.

Figure 7.8 Docked Report Tabs



Tips:

- Keyboard shortcuts that work in JMP windows outside of projects work the same in projects. For example, press Ctrl and then W to close a selected pane (for example, the Window List) or press Ctrl and then S to save a document.
- If the data table associated with a report is updated on your computer, the report in your project is updated when you reopen the project.

Save Files in a Project

When you modify a file (for example, a data table or script) inside a project, an asterisk appears on the tab to indicate that the file is not saved. Select **File > Save** and save the file to your computer. Then, select **File > Save Project** to save the project.

You can also create and save new files from within the project.

1. Follow step 1 through step 4 in [“Create a New Project and Add Data Tables”](#) on page 186 to create a project.
2. In the project, select the Car Poll.jmp tab.
3. Select **Tables > Subset**.
4. Select **Subset by**.
5. Select country from the Columns list and click **OK**.

Three new data tables subset by country are created in the project. Notice the asterisk on the tab of each new data table.

Figure 7.9 New Subset Data Tables

	sex	marital status	age	size	type
1	Male	Single	36	Small	Sporty
2	Male	Married	23	Small	Family
3	Male	Single	34	Medium	Family
4	Male	Single	25	Small	Sporty
5	Female	Single	27	Medium	Family
6	Male	Married	28	Small	Family

6. Select each new tab, select **File > Save**, and save the file on your computer.

Tip: If you plan to share the project, save the data table in the same folder as (or in a subfolder of) the project.

7. Select **File > Save Project** to save the data tables in the project.

Save Files in a Different Format and Save to the Project

You can save a graph in another format (such as Interactive HTML or PowerPoint) and then bookmark it in the project.

1. Select **File > New > Project** to create an empty project.
2. Select **File > Open**, browse to the JMP Samples/Data folder, and open Big Class Families.jmp.
3. Run the Graph Builder script.
4. (Windows) Select **File > Save As** and select **PowerPoint Presentation** from the Save as Type list.

Tip: Select **Bookmark in project** to add a bookmark to the file.

Browse to the desktop and click **Save**.

The PowerPoint file opens automatically.

5. (Macintosh) Select **File > Export (Macintosh)** and select **Microsoft PowerPoint**.

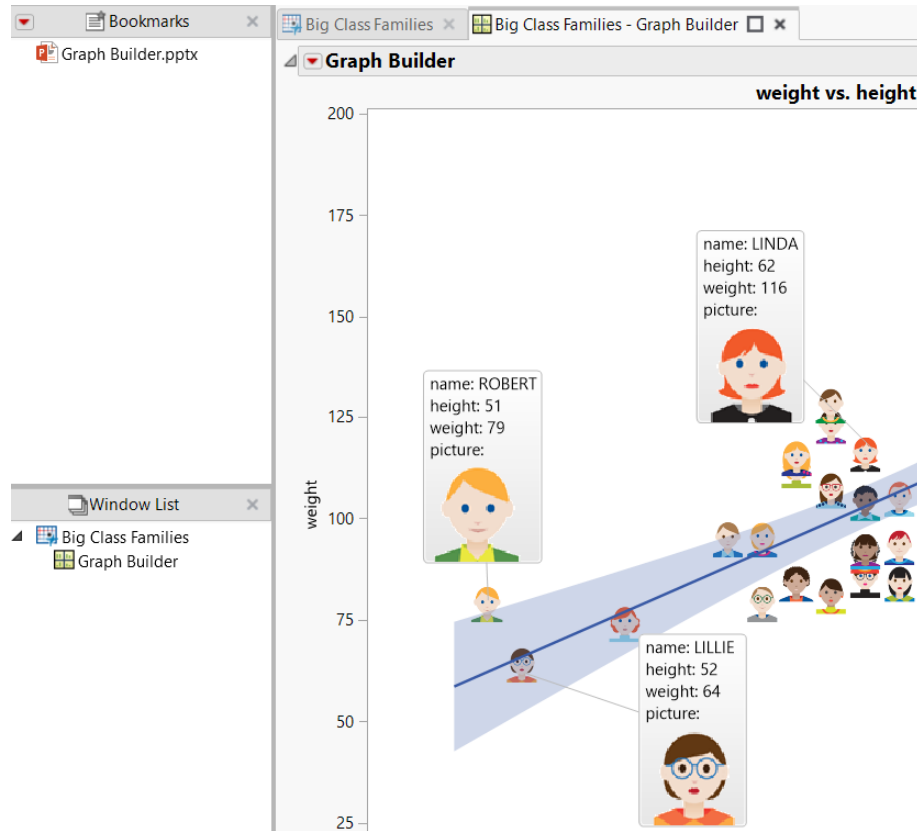
Click **Next**, browse to the desktop, and then click **Export**.

The PowerPoint file opens automatically.

6. In JMP, from the Bookmarks red triangle menu, select **Add Files**, navigate to the **Graph Builder.pptx** file that you created, and then click **Open (Windows)** or **Choose (Macintosh)**.

The PPTX file is added to the Bookmarks pane.

Figure 7.10 Bookmarked PPTX File



Move Files to a Project

Suppose you generated graphs and reports from open data tables and you want to put all the files in a new project. The Move to/from Project option in the Window menu enables you to quickly combine open windows into a project. You can move windows to a new project or move to windows to an open project.

1. Select **Help > Sample Data Library**.
2. Open **Big Class Families.jmp**.
3. Run the Distribution, Text Explorer, and Graph Builder scripts.
4. From any window, select **Window > Move to/from Project**.
5. Leave Source Project as "(None)" because you are not moving files from another project.
6. Leave Destination Project as "(New Project)" because you want to put the files in a new project.

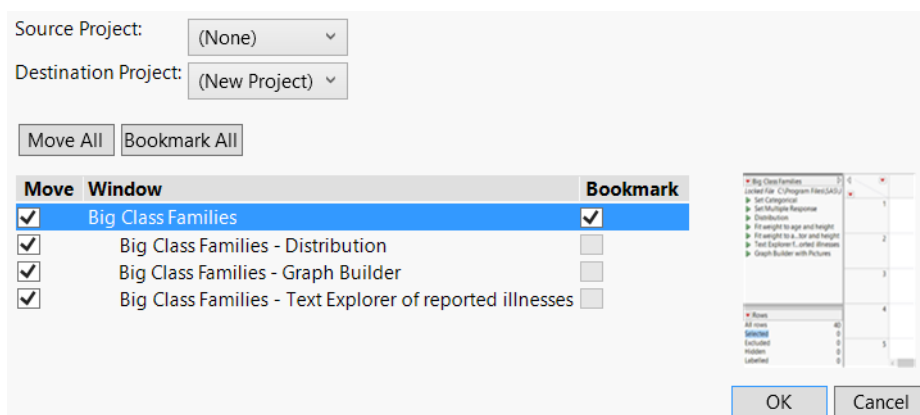
7. Select the check box next to Big Class Families.jmp.

The graphs associated with the data table are also selected.

8. Click **Bookmark** next to Big Class Families.

A bookmark to the data table will be included in the project.

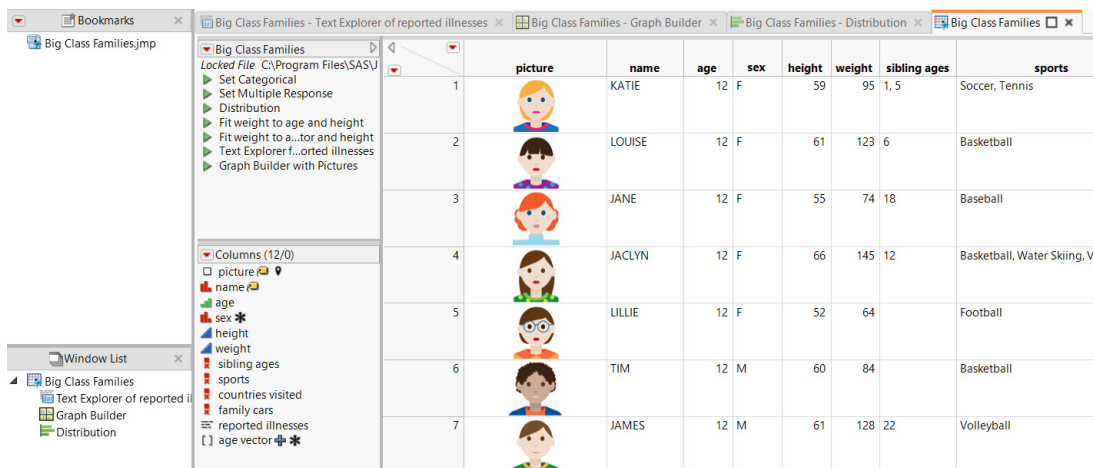
Figure 7.11 Moving Files to a New Project



9. Click **OK**.

A new project with the selected files and bookmark appears.

Figure 7.12 New Project



Share the Project

Sometimes you want to archive a project and share it with others. JMP zips the files in the project and saves the state of the project (the bookmarks, open panes, pane sizes, and so on).

When other users open the archived project on their computer, the links are resolved as follows:

1. If the file is in the same folder as the project or in a subfolder, the link contains a relative path, such as `My Data/Corn Trials.jmp`.
2. If the file is in a directory that JMP represents as a path variable, such as the desktop or documents directory, the links contain the path variable, such as `$DESKTOP/Corn Trials.jmp`. See the Types of Data chapter in the *Scripting Guide* for more information.
3. Otherwise, the link contains an absolute path, such as `C:/2017Data/Corn Trials.jmp`.

Caution: An absolute path does not work on other users' computers unless their file system is set up like yours.

To archive a project

1. Make sure the project is saved.
2. Select **File > Archive Project**.
3. Browse to the folder where you want to save the project files.
4. Click **Save**.

To unarchive a project

1. Use Windows Explorer or Macintosh Finder to create the folder in which you want to extract the files.
2. Open the archive in JMP.
3. Select the folder that you just created and click **Extract**.

Save and Run Scripts

Most platform options in JMP are scriptable, meaning that most actions that you perform can be saved as a JMP Scripting Language (JSL) script. You can use a script to reproduce your actions or results at any time.

Example of Saving and Running a Script

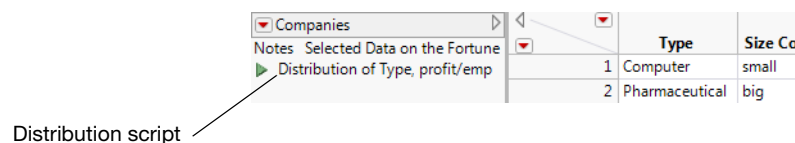
Create a Report

1. Select **Help > Sample Data Library** and open **Companies.jmp**.
2. Select **Analyze > Distribution**.
3. Select **Type** and **profit/emp** and click **Y, Columns**.
4. Click **OK**.
5. From the red triangle menu for **Type**, select these options:
 - **Histogram Options > Show Counts**
 - **Confidence Interval > 0.95**
6. From the red triangle menu for **profit/emp**, select these options:
 - **Outlier Box Plot**, to remove the outlier box plot
 - **CDF Plot**
7. From the red triangle menu for **Distributions**, select **Stack**.

Save the Script to the Data Table and Run It

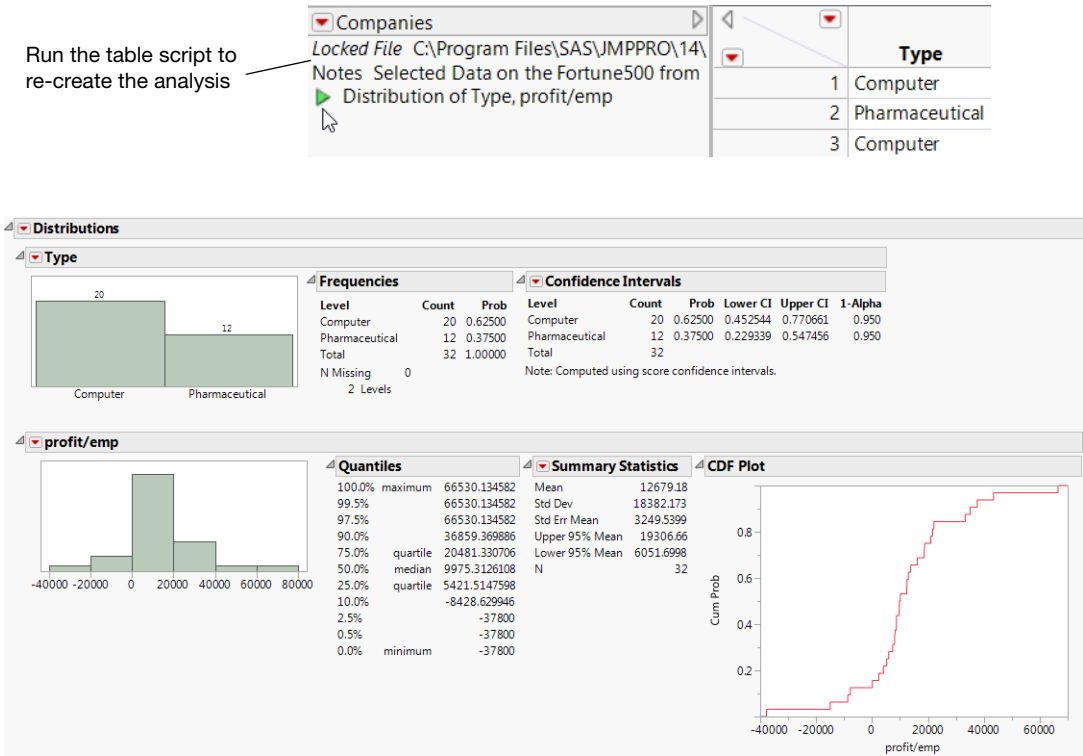
1. To save this analysis, select **Save Script > To Data Table** from the red triangle menu for **Distributions**. The new script appears in the **Table** panel.

Figure 7.13 Distribution Script



2. Close the **Distribution** report window.
3. To re-create the analysis, click the green triangle next to the **Distribution** script.

Figure 7.14 Running the Distribution Script



Tip: Right-click the table script to view more options.

About Scripts and JSL

The script that you saved in this section contains JMP Scripting Language (JSL) commands. JSL is a more advanced topic that is covered in the *Scripting Guide* and *JSL Syntax Reference*.

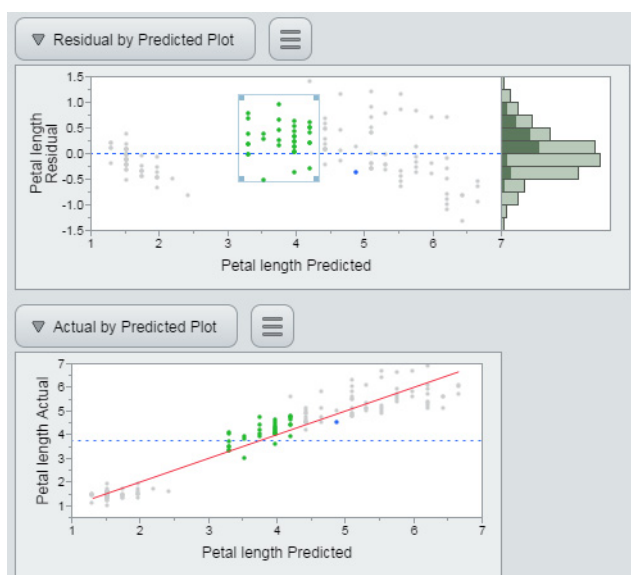
Save Reports as Interactive HTML

Interactive HTML enables JMP users to share reports that contain dynamic graphs so that even non JMP users can explore the data. The JMP report is saved as a web page in HTML 5 format, which you can email to users or publish on a website. Users then explore the data as they would in JMP.

Interactive HTML provides a subset of features from JMP:

- Explore interactive graph features, such as selecting histogram bars and viewing data values.
- View data by brushing.
- Show or hide report sections.
- Place your cursor over the report for tooltips.
- Increase the marker size.

Figure 7.15 Brushing Data in Interactive HTML



Many changes that you make to the graphs, such as ordered variables, horizontal histograms, background colors, and colored data points, are saved in the web page. Graphs and tables that are closed when you save the content remain closed on the web page until the user opens them.

Interactive HTML Contains Data

When you save reports as interactive HTML in JMP, your data are embedded in the HTML. The content is unencrypted, because web browsers cannot read encrypted data. To avoid sharing sensitive data, save your results as a non-interactive web page. (Select **File > Save As > HTML File** on Windows, or **File > Export > HTML** on Macintosh.)

Example of Creating Interactive HTML

Create a Report

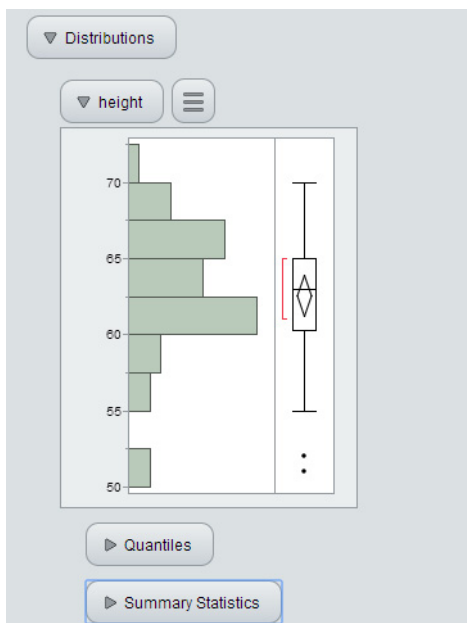
1. Select **Help > Sample Data Library** and open **Big Class.jmp**.
2. Select **Analyze > Distribution**.
3. Select **height** and click **Y, Columns**.
4. Click **OK**.

Save as Interactive HTML

1. On Windows, select **File > Save As** and select **Interactive HTML with Data** from the Save as type list. On Macintosh, select **File > Export > Interactive HTML with Data**.
2. Name and save the file (or export on Macintosh).

The output appears in your default browser.

Figure 7.16 Interactive HTML Output



For information about exploring interactive HTML output, visit

<https://www.jmp.com/support/help/InteractiveHTML/14/ShareJMPReports.shtml>.

Create a Web Report

The File > Publish option creates a web page in which reports, descriptive text, and graphics are displayed. The web page, graphics, and support files are saved in the directory that you specify so that you can zip the files and send them to another user. This feature is particularly helpful for non JMP users.

Some graphs can be displayed as interactive HTML (for example, bubble plots or background maps). When you click on the thumbnail of the graph in a web report, the interactive graph appears on a new web page.

To generate a web report, follow these steps:

1. Select **Help > Sample Data Library** and open SATByYear.jmp.
2. Run the Graph Builder Map and Bubble Plot by State scripts.
3. From any JMP window, select **File > Publish**.

In the Select Reports window that appears, you can select one or more reports to display in the web page. In this example, both reports are displayed in the web page because they are not selected in this window.

Note that the Folder Name box indicates the current date and time. The web report files are saved in this folder.

4. Click **Next**.
5. Enter the descriptions shown in Figure 7.17.

Note: If you selected one report, you have the option of generating an index page. The index page shows a thumbnail of the report. You click the thumbnail to view the report. Figure 7.18 shows an example.

Figure 7.17 Customizing the Web Report

Customize Reports

Title:
Graph Builder State Colored by Expenditure

Description:
Place your cursor over the state to see the mean expenditure.

11/01/2017 10:38 AM

Title:
Bubble Plot of SAT Verbal by SAT Math Sized by % Taking (2004)
Across Year ID State

Description:
Click the Play button to animate the plot.

11/01/2017 10:38 AM

☒ Open published web report

Web Report Title:
SAT By Year Reports

Web Report Description:
Click any of the thumbnails below to launch interactive JMP reports.

► **Customizations**

Previous

Build Report

Cancel

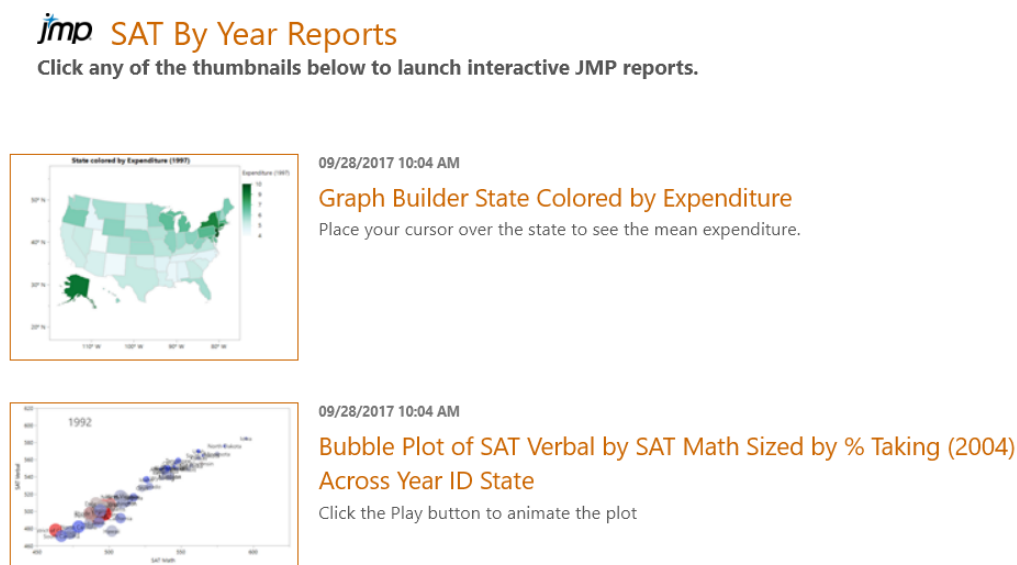
Help

Note: The reports are in the order you created them. In this example, the Graph Builder Map script was run first, so the map is the first report. Use the arrows next to the reports to rearrange them. Also, after you remove a report, a Restore Reports button appears in case you want to add the report again.

6. Enter *SAT By Year Reports* in the Web Report Title box.
7. Click **Build Report**.

The web page opens in your default browser.

Figure 7.18 Web Report



Save a Report as a PowerPoint Presentation

Create a presentation by saving JMP results as a Microsoft PowerPoint presentation (.pptx). Rearrange JMP content and edit text in PowerPoint after saving as a .pptx file. Sections of a JMP report are exported into PowerPoint differently.

- Report headings are exported as editable text boxes.
- Graphs are exported as images. Certain graphical elements, such as legends, are exported as separate images. Images resize to fit the slide in PowerPoint.

Use the selection tool to select the sections that you want to save in your presentation. Delete unwanted content once after you open the file in PowerPoint.

Note: On Windows, PowerPoint 2007 is the minimum version required to open .pptx files created in JMP. On Macintosh, at least PowerPoint 2011 is required.

1. In JMP, create the report.
2. (Windows) Select **File > Save As** and select **PowerPoint Presentation** from the Save as type list.
3. (Macintosh) Select **File > Export > PowerPoint Presentation** and click **Next**.
4. Select a graphic file format from the list.

On Windows, EMF is the default format. On Macintosh, PDF is the default format.

5. Name and save the file (or export on Macintosh).

The file opens in Microsoft PowerPoint because **Open the file after saving** is selected by default.

Note: The native EMF graphics produced on Windows are not supported on Macintosh. The native PDF graphics produced on Macintosh are not supported on Windows. For cross-platform compatibility, change the default graphics file format by selecting **File > Preferences > General**. Then, change the **Image Format for PowerPoint** to either PNG or JPEG.

Create Adobe Flash Versions of the Profiler, Bubble Plot, or Distribution Platforms

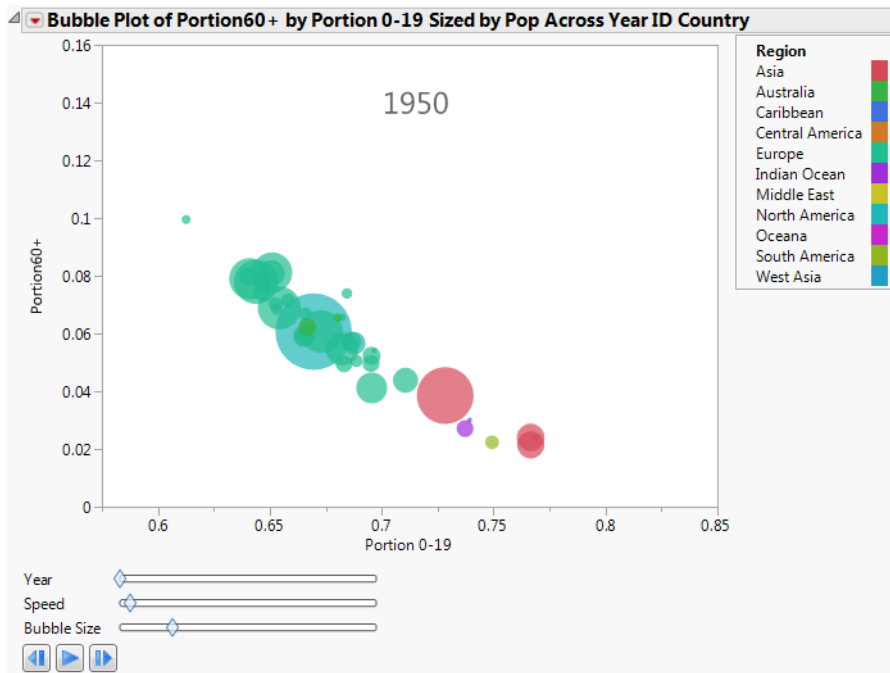
To share the interactive results of a profiler, bubble plot, or distribution outside of JMP, you can export an SWF file and use the Adobe Flash Player to view it. You can import the SWF file into presentations and applications. You can also save the results as an HTML page with the SWF output embedded.

Example of Saving an Adobe Flash Version of a Bubble Plot

Create a Bubble Plot in JMP

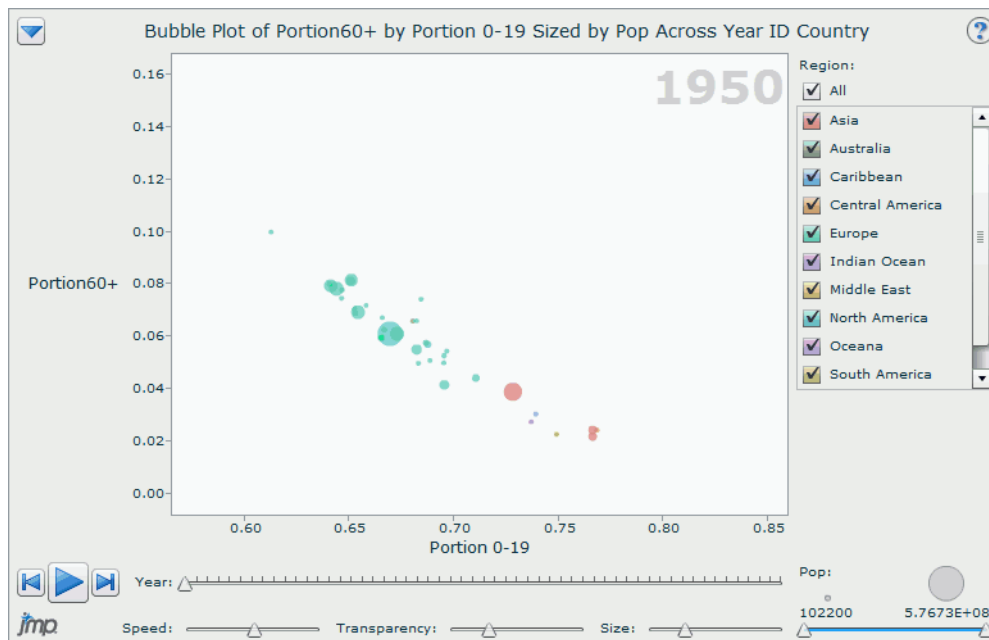
1. Select **Help > Sample Data Library** and open PopAgeGroup.jmp.
2. Select **Graph > Bubble Plot**.
3. Select Portion60+ and click **Y**.
4. Select Portion 0-19 and click **X**.
5. Select Country and click **ID**.
6. Select Year and click **Time**.
7. Select Pop and click **Sizes**.
8. Select Region and click **Coloring**.
9. Click **OK**.

Figure 7.19 Initial Bubble Plot



Save the Bubble Plot as an SWF File

1. Select **Save for Adobe Flash Platform (SWF)** from the red triangle menu.
2. In the Save As SWF window, select the location to which you want to save the file.
3. Click **Save**. The Adobe Flash version of the Bubble Plot is saved as HTML and appears in a web browser.

Figure 7.20 Adobe Flash Version of Bubble Plot**Additional Information**

The JMP website has additional information about the following:

- more details about Adobe Flash versions of the Profiler, Bubble Plot, and Distribution platforms
- instructions on importing the Adobe Flash versions into Microsoft PowerPoint

Visit <https://www.jmp.com/support/swfhelp/en/> for details.

Create Dashboards

A dashboard is a visual tool that lets you run and present reports on a regular basis. You can show reports, data filters, selection filters, data tables, and graphics on a dashboard. The content shown on the dashboard is updated when you open the dashboard.

Example of Combining Windows

You can quickly create dashboards by merging several open windows in JMP. Combining windows provides options to view a summary of statistics and include a selection filter.

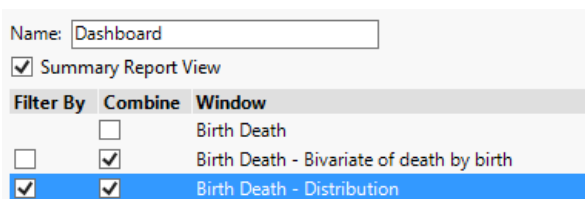
1. Select **Help > Sample Data Library** and open Birth Death.jmp.
2. Run the Distribution and Bivariate table scripts.
3. Select **Window > Combine Windows**.

The Combine Windows window appears.

Tip: On Windows, you can also select Combine Windows from the Arrange Menu option in the lower right corner of JMP windows.

4. Select **Summary Report View** to display the graphs and omit the statistical reports
5. In the Combine column, select **Birth Death - Bivariate of death by birth** and **Birth Death - Distribution**.
6. In the Filter By column, select **Birth Death - Distribution**.

Figure 7.21 Combine Windows Options



7. Click **OK**.


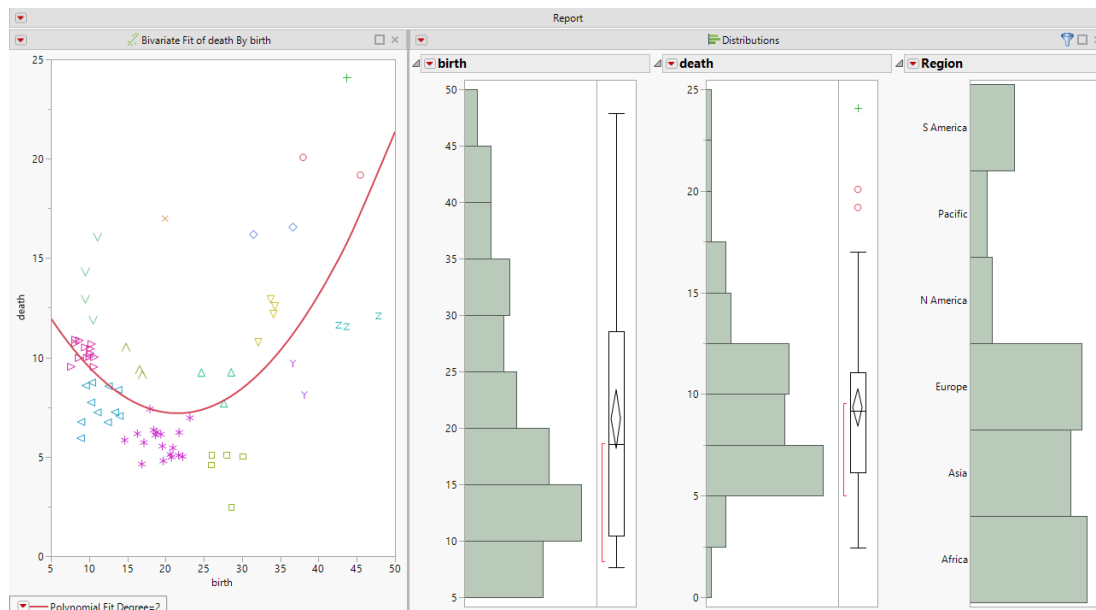
The two reports are combined into one window. Notice the filter icon  at the top of the Distribution report. When you select a bar in one of the histograms, the corresponding data in the Bivariate graph are selected.

Figure 7.22 Combined Windows

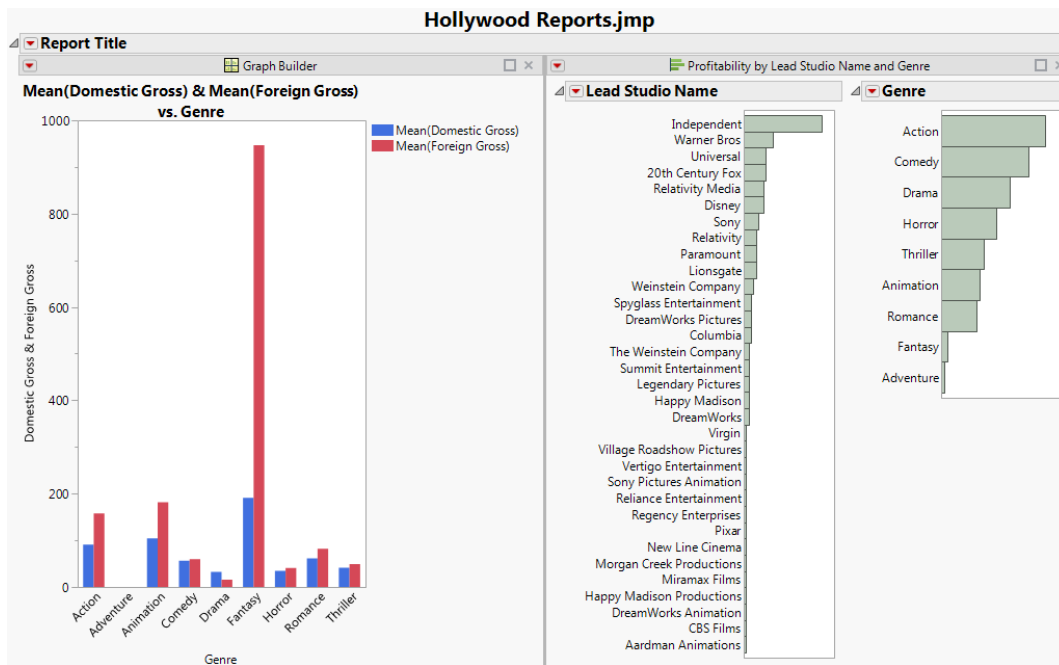


Example of Creating a Dashboard with Two Reports

Suppose that you created two reports and want to run the reports again the next day against an updated set of data. This example shows how to create a dashboard from the reports in Dashboard Builder.

1. Select **Help > Sample Data Library** and open *Hollywood Movies.jmp*.
2. Run the table scripts named “Distribution: Profitability by Lead Studio and Genre” and “Graph Builder: World and Domestic Gross by Genre”.
3. From any window, select **File > New > Dashboard**.
Templates for common layouts appear.
4. Select the **2x1 Dashboard** template.
A box with room for two reports appears on the workspace.
5. In the Reports list, double-click the report thumbnails to put them on the dashboard.
6. Select **Preview Mode** from the Dashboard Builder red triangle menu.
A preview of the dashboard appears. Notice that the graphs are linked to each other and the data table. They also have the same red triangle options as the Distribution and Graph Builder platforms.
7. Click **Close Preview**.

Figure 7.23 Dashboard with Two Reports



For more information about creating dashboards, see the Extend JMP chapter in the *Using JMP* book.

Chapter 8

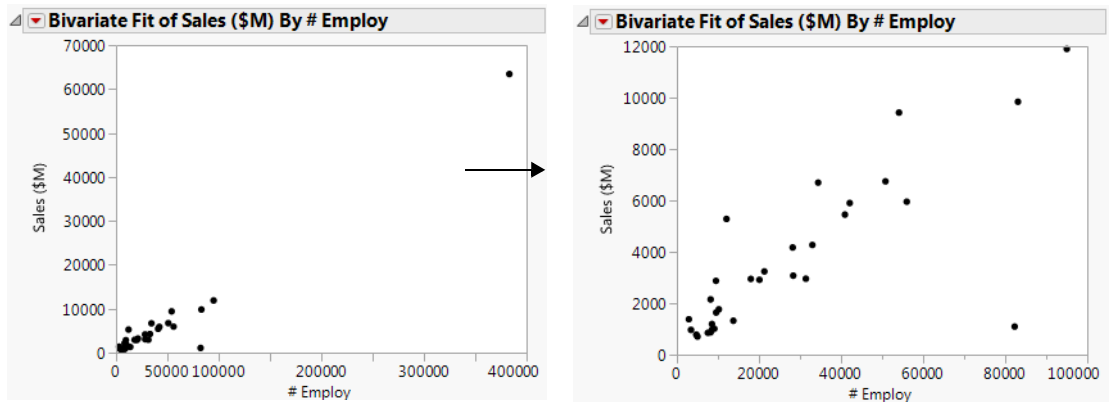
Special Features

Automatic Analysis Updates and SAS Integration

Using some of the special features in JMP, you can do the following:

- Update analyses or graphs automatically
- Customize platform results
- Integrate with SAS to use advanced analytical features

Figure 8.1 Examples of Special Features



```
DATA Candy_Bars; INPUT Calories Total_fat_g Carbohydrate_g Protein_g; Lines;
310 20 28 6
230 12 27 4
220 12 24 3
170 8 21 3
200 2.5 43 1
260 16 26 5
190 1.5 42 2
190 11 21 2
230 12 28 3
;
RUN;

PROC GLM DATA=Candy_Bars ALPHA=0.05;
MODEL Calories = Total_fat_g Carbohydrate_g Protein_g;
RUN;
```

Contents

Automatically Update Analyses and Graphs	211
Example of Using Automatic Recalc	211
Change Preferences	214
Example of Changing Preferences	215
Integrate JMP and SAS	217
Example of Creating SAS Code	218
Example of Submitting SAS Code	218

Automatically Update Analyses and Graphs

When you make a change to a data table, you can use the Automatic Recalc feature to automatically update analyses and graphs that are associated with the data table. For example, if you exclude, include, or delete values in the data table, that change is instantly reflected in the associated analyses or graphs. Note the following information:

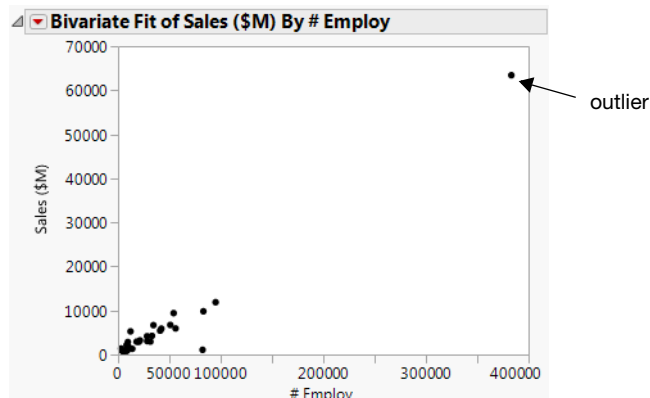
- Some platforms do not support Automatic Recalc. For more information, see the JMP Reports chapter in the *Using JMP* book.
- For the supported platforms in the **Analyze** menu, Automatic Recalc is turned off by default. However, for the supported platforms in the **Quality and Process** menu, Automatic Recalc is turned on by default, except for the Variability/Attribute Gauge Chart, Capability, and Control Chart.
- For the supported platforms in the **Graph** menu, Automatic Recalc is turned on by default.

Example of Using Automatic Recalc

This example uses the Companies.jmp sample data table, which contains financial data for 32 companies from the pharmaceutical and computer industries.

1. Select **Help > Sample Data Library** and open Companies.jmp.
2. Select **Analyze > Fit Y by X**.
3. Select Sales (\$M) and click **Y, Response**.
4. Select # Employ and click **X, Factor**.
5. Click **OK**.

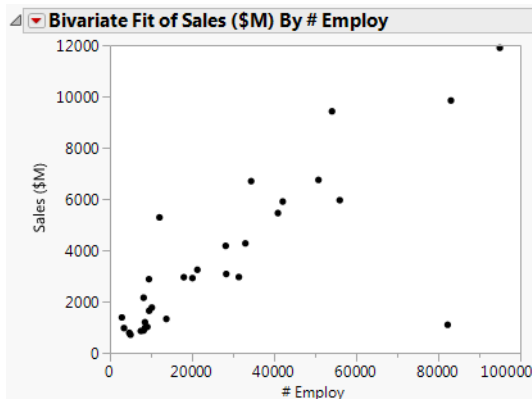
Figure 8.2 Initial Scatterplot



The initial scatterplot shows that one company has significantly more employees and sales than the other companies. You decide that this company is an outlier, and you want to exclude that point. Before you exclude the point, turn on Automatic Recalc so that your scatterplot is updated automatically when you make the change.

6. Turn on Automatic Recalc by selecting **Redo > Automatic Recalc** from the red triangle menu.
7. Click on the outlier to select it.
8. Select **Rows > Exclude/Unexclude**. The point is excluded from the analysis and the scatterplot is automatically updated.

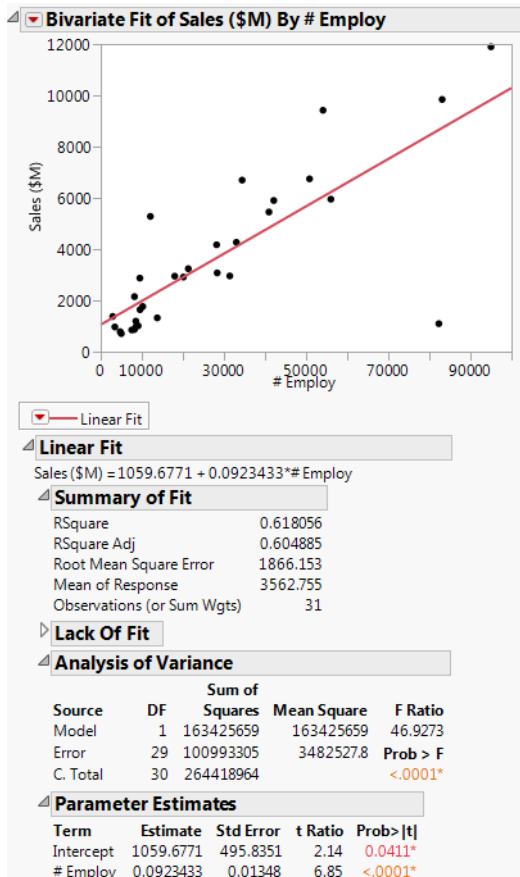
Figure 8.3 Updated Scatterplot



If you fit a regression line to the data, the point in the lower right corner is an outlier, and influences the slope of the line. If you then exclude the outlier with Automatic Recalc turned on, you can see the slope of the line change.

9. Fit a regression line by selecting **Fit Line** from the red triangle menu. Figure 8.4 shows the regression line and analysis results added to the report window.

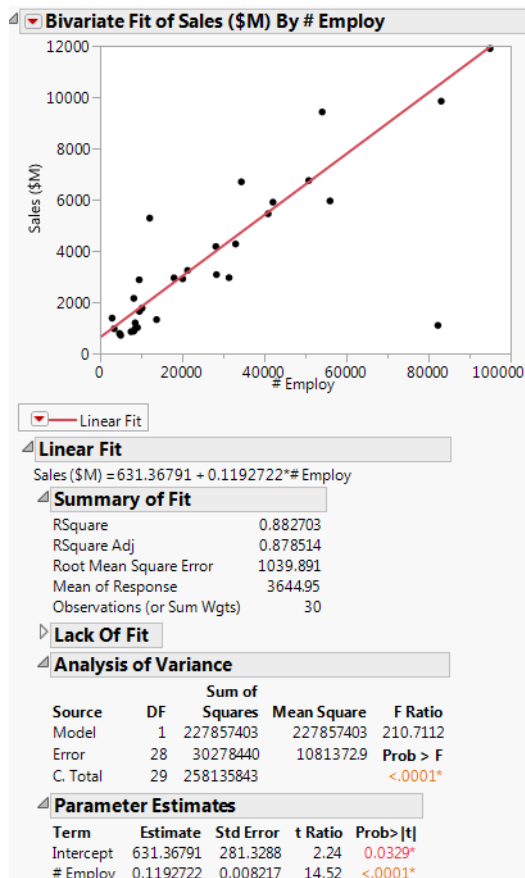
Figure 8.4 Regression Line and Analysis Results



10. Click on the outlier to select it.
11. Select **Rows > Exclude/Unexclude**. The regression line and analysis results are automatically updated, reflecting the exclusion of the point.

Tip: When you exclude a point, the analyses are recalculated without the data point, but the data point is not hidden in the scatterplot. To also hide the point in the scatterplot, select the point, and then select **Rows > Hide and Exclude**.

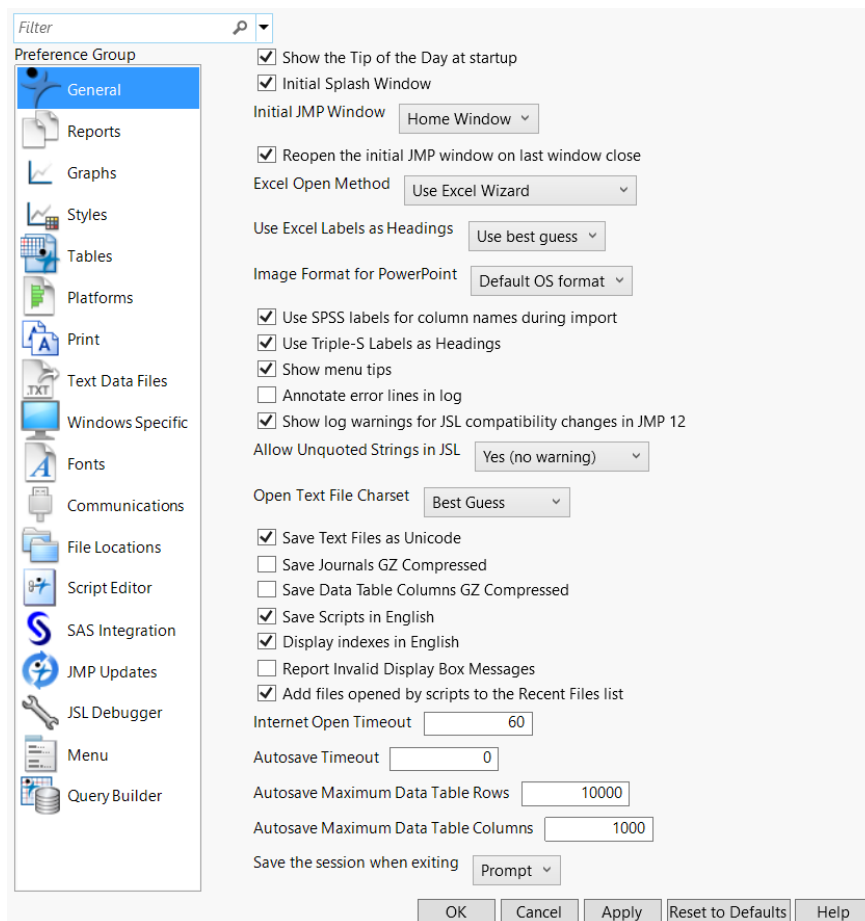
Figure 8.5 Updated Regression Line and Analysis Results



Change Preferences

You can change preferences in JMP using the Preferences window. To open the Preferences window, select **File > Preferences (Windows)** or **JMP > Preferences (Macintosh)**.

Figure 8.6 Preferences Window



On the left side of the Preferences window is a list of Preference groups. On the right side of the window are all of the preferences that you can change for the selected category.

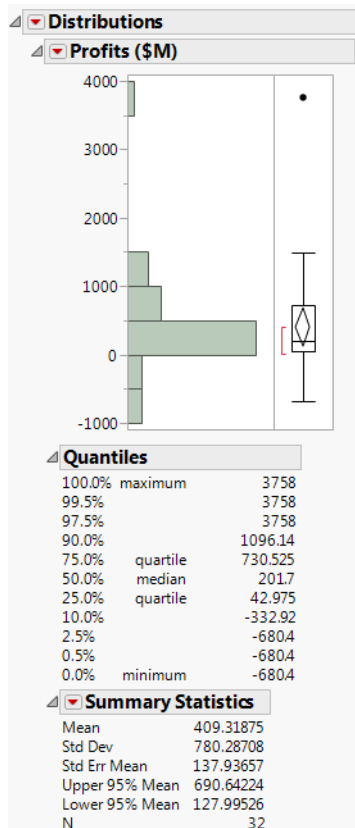
Example of Changing Preferences

Every platform report window has options that you can turn on or off. However, your changes to these options are not remembered the next time you use the platform. If you want JMP to remember your changes every time you use the platform, change those options in the Preferences window.

This example shows how to set the Distribution platform so that an Outlier Box Plot is not added to the initial report.

Create a Distribution Using the Default Preference Setting

1. Select **Help > Sample Data Library** and open Companies.jmp.
2. Select **Analyze > Distribution**.
3. Select Profits (\$M) and click **Y, Columns**.
4. Click **OK**.

Figure 8.7 Distribution Report Window

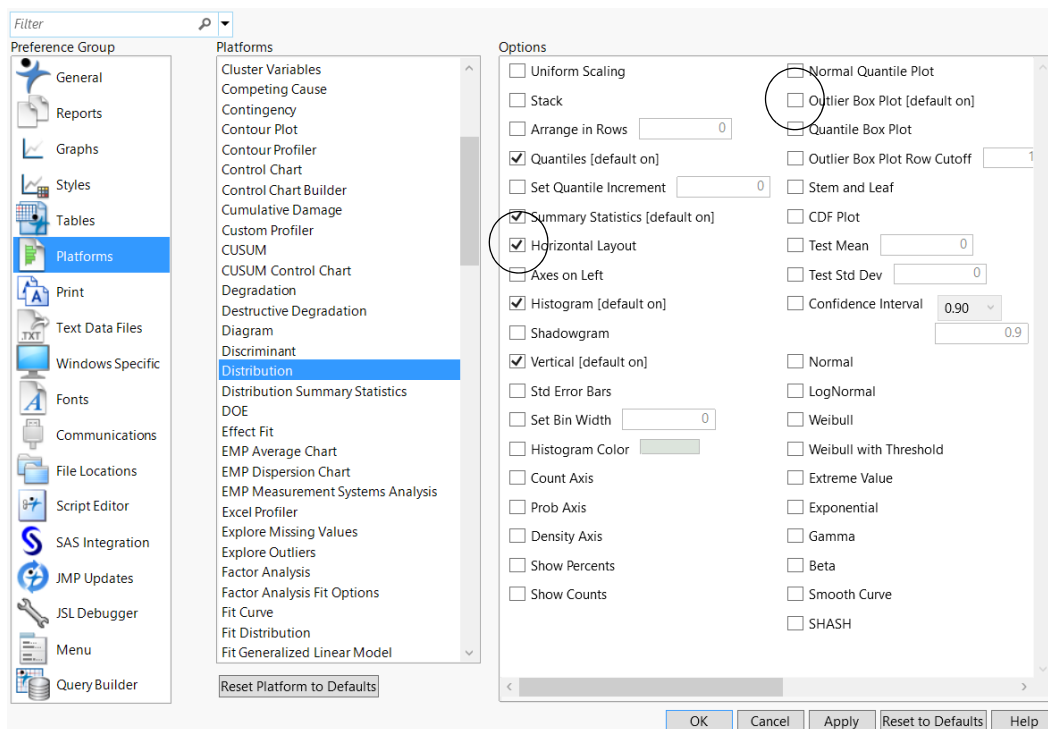
The histogram is vertical, and the graphs includes an outlier box plot. To change the histogram to horizontal and remove the outlier box plot, select the appropriate options from the red triangle menu for Profits (\$M). However, if you want those preferences to be in effect every time you use the platform, then change them in the Preferences window.

Change the Preference for the Outlier Box Plot and Run Distribution Again

1. Select **File > Preferences (Windows)** or **JMP > Preferences (Macintosh)**.
2. Select **Platforms** from the preference group.

3. Select **Distribution** from the Platforms list.
4. Select the **Horizontal Layout** option to turn it on.
5. Deselect the **Outlier Box Plot** option to turn it off.

Figure 8.8 Distribution Preferences



6. Click **OK**.
7. Repeat the Distribution analysis. See [“Create a Distribution Using the Default Preference Setting”](#) on page 216.

The histogram is now horizontal and the outlier box plot does not appear. These preferences remain the same until you change them.

For details about all of the preferences, see the JMP Preferences chapter in the *Using JMP* book.

Integrate JMP and SAS

Note: You must have access to SAS, either on your local machine or on a server, to use SAS through JMP.

Using JMP, you can interact with SAS as follows:

- Write or create SAS code in JMP.
- Submit SAS code and view the results in JMP.
- Connect to a SAS Metadata Server or a SAS Server on a remote machine.
- Connect to SAS on your local machine.
- Open and browse SAS data sets.
- Retrieve and view data sets generated by SAS.

For complete details about integrating JMP and SAS, see the Import Your Data chapter in the *Using JMP*.

Example of Creating SAS Code

This example uses the Candy Bars.jmp sample data table, which contains nutrition data for candy bars.

1. Select **Help > Sample Data Library** and open Candy Bars.jmp.
2. Select **Analyze > Fit Model**.
3. Select Calories and click **Y**.
4. Select Total fat g, Carbohydrate g, and Protein g, and click **Add**.
5. From the red triangle menu for Model Specification, select **Create SAS Job**.

Figure 8.9 shows the SAS code. (Not all of the data is shown.)

Figure 8.9 SAS Code

```
DATA Candy_Bars; INPUT  Calories Total_fat_g Carbohydrate_g Protein_g; Lines;
310 20 28 6
230 12 27 4
220 12 24 3
170 8 21 3
200 2.5 43 1
260 16 26 5
190 1.5 42 2
190 11 21 2
230 12 28 3
;
RUN;

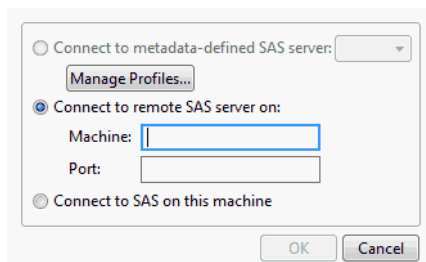
PROC GLM DATA=Candy_Bars ALPHA=0.05;
MODEL Calories = Total_fat_g Carbohydrate_g Protein_g;
RUN;
```

Example of Submitting SAS Code

1. Select **Help > Sample Data Library** and open Candy Bars.jmp.
2. Select **Analyze > Fit Model**.

3. Select **Calories** and click **Y**.
4. Select **Total fat g**, **Carbohydrate g**, and **Protein g**, and click **Add**.
5. From the red triangle menu for Model Specification, select **Submit to SAS**.
6. In the **Connect to SAS Server** window (see Figure 8.10), choose a method to connect to SAS (if you are not already connected). For this example, select **Connect to SAS on this machine**.

Figure 8.10 Connect to SAS Server



7. Click **OK**.

JMP connects to SAS. SAS runs the model and sends the results back to JMP. The results can appear as SAS output, HTML, RTF, PDF, or JMP report format (you can choose the format using JMP Preferences). Figure 8.11 shows the results formatted as a JMP report. For details, see the *Import Your Data* chapter in the *Using JMP* book.

Figure 8.11 SAS Results Formatted as a JMP Report

The SAS System
The GLM Procedure

The GLM Procedure

Data

Number of Observations

Number of Observations Read75
Number of Observations Used75

Dependent Variable: Calories

Analysis of Variance

Calories

Overall ANOVA

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	3	282358	94119.3	3237.58	<.0001*
Error	71	2064.03	29.0709	.	.
Corrected Total	74	284422	.	.	.

Fit Statistics

R-Square	Coeff Var	Root MSE	Calories Mean
0.99274	2.21858	5.39174	243.027

Type I Model ANOVA

Source	DF	Type I SS	Mean Square	F Value	Pr > F
Total_fat_g	1	185260	185260	6372.68	<.0001*
Carbohydrate_g	1	93540.4	93540.4	3217.67	<.0001*
Protein_g	1	3557.86	3557.86	122.386	<.0001*

Type III Model ANOVA

Source	DF	Type III SS	Mean Square	F Value	Pr > F
Total_fat_g	1	111777	111777	3844.97	<.0001*
Carbohydrate_g	1	96756.1	96756.1	3328.28	<.0001*
Protein_g	1	3557.86	3557.86	122.386	<.0001*

Solution

Parameter	Estimate	Standard Error	t Value	Pr > t
Intercept	-5.9643	2.89999	-2.0567	0.0434*
Total_fat_g	8.98995	0.14498	62.0078	<.0001*
Carbohydrate_g	4.0975	0.07102	57.6913	<.0001*
Protein_g	4.40133	0.39785	11.0628	<.0001*

A

- Actual by Predicted plot 163
- Adobe Flash
 - creating bubble plots 202
 - creating Profiler results 202
- analyses
 - adding to journals 186
 - analyzing variable relationships 145
 - automatically updating 211–214
- analyzing data 131
- Anscombe, F.J. 133
- automatic recalc 211–214
- averages, comparing
 - multiple variables 156
 - single variable 151

B

- bar charts 99
 - creating 100
 - example 100–101
 - interacting with 101
 - interpreting 100
- box plots 109
 - changing to line plots 121
 - creating 110
 - Distribution platform and 141
 - example 110
 - interpreting 111
 - outlier box plots 61
- bubble plots 124
 - creating 125
 - creating Adobe Flash version 202
 - example 125
 - interacting with 128
 - interpreting 127
 - options 128
 - trail bubbles 128

C

- categorical variables
 - comparing averages 151
 - comparing proportions 154
 - example 142–145
 - relationships 145
- cells, selecting or deselecting 77
- character data type 80
- charts
 - bar charts 99–101
 - creating separate charts 122
 - variability charts 115
- clustering 177
- code, SAS
 - creating 218
 - submitting 218
- Column Info window 79–80
- column names 80
 - changing 80
 - entering 80
 - unique 80
- Column Properties menu 81
- columns
 - adding 71
 - locking 81
 - naming 80
 - selecting or deselecting 76
- columns panel 58
- combining reports in dashboard 205
- contingency tables 155
- continuous modeling type 80, 136
- continuous variables 136
 - comparing averages 151
 - example of analyzing 140
 - fitting lines 147
 - modeling types 136
 - regression with one predictor 146
 - relationships 145

copying and pasting data 67
correlation 102

D

dashboard, creating 204

data

analyzing 131

changing

modeling types 137

multiple values 74

single value 74

copying and pasting 67

creating patterned 75

deselecting

cells 77

columns 76

rows 76

excluding data points 104

filtering 83

finding means 89

graphing 133

importing 67–68

including data points 104

moving into JMP 67

searching for values 78

selecting

cells 77

columns 76

data points 106

rows 76

unexcluding 104

data filter 83

data grid 57

data points

excluding 104

including 104

selecting 106

data tables 53

See also modeling types

adding

columns 71

rows 71

changing name of 71

Column Properties menu 81

columns panel 58

creating 70

data grid 57

data types 80

example 65

formatting numeric values 80

joining 91

opening 67

opening sample 56

rows panel 58

sorting by columns 93

subsetting 89

table panel 57

typing data into 70

data types 80

disclosure icons 53, 60

Distribution platform 139

box plot and 141

example, categorical variables 142–145

example, continuous variables 140

histograms and 141

Linear Fit report 149

Quantiles report 141

rotating the report window 140

Summary Statistics report 141

t-tests 142

E

Effect Tests report 159

example, analyses 131

excluding data 104

F

Fill options 75

filtering data 83

Fit Model platform

Actual by Predicted plot 163

analyzing variable relationships 145

Effect Tests report 159

launching 162

Parameter Estimates report 163

Prediction Profiler 164

Fit Y by X platform

analyzing variable relationships 145

contingency table 155

launching 151

Means and Std Dev report 153

- Mosaic Plot 156
- Tests report 156
- t-tests 153

fitting lines 147

formula editor 81

G

Graph Builder 118

- changing box plots to line plots 121
- creating graphs 119
- creating separate charts 122
- example 119
- interpreting graphs 123
- launching 157

graphing data 133

graphs

- automatically updating 211–214
- creating 119
- multiple variable 102
- See also* Graph Builder
- single variable 97

grid lines, adding to overlay plots 114

H

hierarchical clustering 177

histograms 97

- creating 98
- Distribution platform and 141
- example 97
- interacting with 99, 144
- interpreting 98

Home window 54

HTML 5, save output as 196

I

icons, modeling type 71

importing data 67

interaction 158

interactive HTML, save report as 196

J

JMP 51

- copying and pasting data into 67
- creating SAS code in 218

- getting started 53–63
- importing data 67
- integrating with SAS 217
- starting JMP 54
- submitting SAS code in 218
- versus Microsoft Excel 63

JMP scripting language 196

joining data tables 91

journals

- adding analyses 186
- creating 185
- saving platform results 185

JSL (JMP scripting language) 196

L

lasso tool 106

launch windows 53

line plots, changing box plots to 121

linear fit report 149

lines

- adding mean lines 152
- fitting 147

locking columns 81

M

mean lines 152

mean, testing 142

Means and Std Dev report 153

Microsoft Excel

- importing data 68
- versus JMP 63

modeling types

- See also* categorical variables
- changing 137
- continuous 80, 136
- example 136
- icons 71
- nominal 80, 136
- ordinal 80, 136

Mosaic Plot 156

multiple regression 160, 162

multivariate analysis 173

N

names

- columns 80
- data tables 71

nominal modeling type 80

See also categorical variables

numeric data type 80

numeric values, formatting 80

O

ordinal modeling type 80, 136

See also categorical variables

outlier box plots 61

outliers 99, 111

- example 146, 149
 - excluding 149
 - implications of 152, 158
 - removing 104
- overlay plots 112–113
- adding grid lines 114
 - connecting points 114
 - example 112
 - interacting with 114
 - interpreting 114
 - legends 115
 - multiple variables and 115

P

Parameter Estimates report 163

patterned data, creating 75

platforms 53, 58

- examples 59–63
- launch windows 53
- launching 59
- report windows and 53, 60, 62
- saving results 185
- viewing results 59

plots

See specific plot type

PowerPoint, saving report as 201

Prediction Profiler 164

predictor variables 145

preferences, changing 214

Profiler platform, creating Adobe Flash

results 202

proportions, comparing 154

Q

Quantiles report 141

R

recoding data 74

red triangle menus 53, 60

regression

- multiple 160, 162
- one predictor 146
- two predictors 160

report windows 60

- closing or opening 62
- description 53
- disclosure icons 53, 60
- interacting with 62
- platform options 53
- red triangle menus 53, 60
- reports in 53
- rotating 140

reports

- Effect Tests 159
- Linear Fit 149
- Means and Std Dev 153
- Parameter Estimates 163
- Quantiles 141
- Summary Statistics 141
- Tests 156

results, saving platform 185

rotating the report window 140

row state data type 80

rows

- adding 71
- clear row states 154
- excluding or unexcluding 104
- returning to default state 154
- selecting or deselecting 76

rows panel 58

S

sample data tables 56

SAS

- creating code 218
- integrating with JMP 217
- submitting code 218
- saving results in journal 185
- scatterplot matrices 106
 - creating 107
 - example 107
 - interacting with 109
 - interpreting 108
- scatterplot matrix 175
- Scatterplot Matrix platform
 - example 161
 - launching 161
- scatterplots 102–103
 - example 103
 - interacting with 106
 - interpreting 104
- scripts
 - creating 194
 - JSL and 196
 - running 194
- searching for values 78
- side-by-side box plots
 - See box plots
- sorting data tables by columns 93
- standard deviations 153
- starting JMP 54
- statistics, summary 85
- subsetting data tables 89
- summary statistics 85
- Summary Statistics report 141

T

- table panel 57
- testing means 142
- Tests report 156
- Tip of the Day window 54
- trail bubbles 128
- t-tests
 - Distribution platform 142
 - Fit Y by X platform 153
 - two-sample 153

V

- values

- changing multiple 74
- changing single 74
- formatting numeric 80
- searching for 78
- variability 110
- variability charts 115
 - creating 116
 - example of 116
 - grouping order 116
 - hiding 117
 - interpreting 118
- variables
 - analyzing relationships 145
 - categorical 142, 145, 151, 154
 - comparing averages for multiple 156
 - comparing averages for single 151
 - continuous 136, 140, 145–147, 151
 - graphs with multiple 102
 - graphs with single 97
 - modeling types 136
 - overlay plots with multiple 115
 - predictor 145
 - types of relationships 145

W-Z

- windows
 - JMP Starter window 54
 - Tip of the Day window 54

