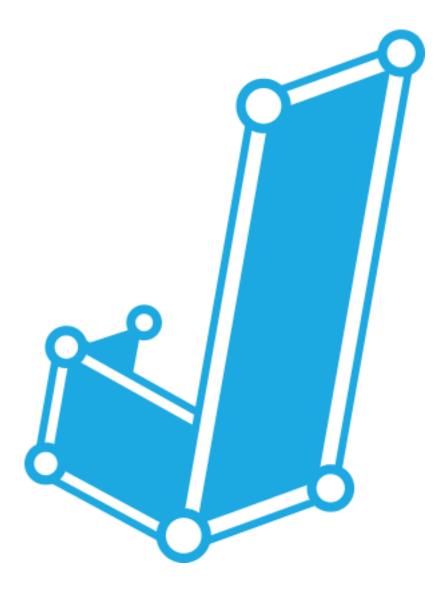
JASP Manual



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WHAT IS JASP?	
FREQUENTLY ASKED QUESTIONS (FAQ):	5
INTRODUCTION TO JASP	6
Downloading JASP Importing Data Into JASP	
DESCRIPTIVE STATISTICS	9
Reliability Analysis	
T-TEST	
INDEPENDENT SAMPLES Paired Samples One Sample	
ANOVA	
REPEATED MEASURES ANOVA	
REGRESSION	25
CORRELATION	-
FREQUENCIES	
BINOMIAL TEST Contingency Tables Log-Linear Regression	
FACTOR ANALYSIS	
Principal Component Analysis Exploratory Factor Analysis	
BAYESIAN METHODS	
Bayesian Independent Samples T-Test Bayesian Paired Samples T-Test Bayesian Repeated Measures ANOVA Bayesian Correlation Matrix Bayesian Correlation Pairs Bayesian Linear Regression Bayesian Binomial Test Bayesian Contingency Tables. Bayesian Log-Linear Regression	
AUTHOR INFORMATION:	

Table of Contents

Welcome to JASP!

The following manual is intended to guide you through running various statistical analyses. We have provided step-by-step instructions on how to run each test along with pictures for you to reference.

What is JASP?

JASP stands for Jeffreys' Amazing Statistical Program, after the Bayesian pioneer Sir Harold Jeffreys. It is a new, intuitive substitute for statistical programs, like SPSS. Unlike other programs commonly used in research labs, colleges, and universities, JASP is completely free. It is an open-source program supported by the University of Amsterdam, developed with the user in mind. Gone are the days of entering in code to run a program. JASP utilizes a point-and-click user interface to make analysis fun and easy.

What functions can JASP perform?

JASP offers both classical and Bayesian analysis procedures. A complete list of the functionality is included below:

Analysis	Classical	Bayesian
ANOVA	✓	\checkmark
ANCOVA	\checkmark	\checkmark
Binomial Test	\checkmark	\checkmark
Multinomial Test	\checkmark	
Contingency Tables (Chi-squared included)	\checkmark	1
Correlation: Pearson, Spearman, Kendall	\checkmark	\checkmark
Exploratory Factor Analysis (EFA)	\checkmark	
Linear Regression	✓	\checkmark
Log-Linear Regression	✓	\checkmark
Logistic Regression	✓	
Principal Component Analysis (PCA)	\checkmark	
Repeated Measures ANOVA	\checkmark	1
Reliability Analyses: α , $\gamma\delta$, and ω	✓	
Structural Equation Modeling (SEM)	\checkmark	
Summary Statistics		\checkmark
T-Tests: Independent, Paired, One-Sample	✓	1

It's important to remember that as an open-source program, JASP is constantly evolving based on the needs of its users. If you see an analysis is missing, their website offers a link to submit a feature request: <u>https://jasp-stats.org/feature-requests-bug-reports/</u>. Another important feature to point out is all of the output tables are in APA format. This makes it easy to copy and paste tables directly from your JASP output.

What are some of the limitations?

As a result of the program constantly increasing its repertoire of analyses, it's important to check the JASP website for updated software. Also remember to consider the version your class or instructor is using when you download the software. **IMPORTANT: JASP does not offer automatic system updates.**

You cannot directly import data from excel. In order to import your data, you must save your file as a .csv file, then open the file in JASP. Step-by-step directions are provided for you reference.

Frequently Asked Questions (FAQ):

How do I cite JASP?

```
JASP Team (2018). JASP (Version 0.9)[Computer software].
```

And the BibTeX entry : @MISC{JASP2017, AUTHOR = {{JASP Team}}, TITLE = {{JASP (Version 0.8.4)[Computer software]}}, YEAR = {2017}, URL = {https://jasp-stats.org/}

What programming language is JASP written in?

JASP is written in C++, R, and javascript.

What do the symbols in the corner of variable boxes represent?

These symbols tell you what type of variable (scale, ordinal, or categorical) can go in the particular category.

Is there anything JASP cannot do at this point?

At present, JASP does not have a sufficient system for producing histograms. It is better to utilize excel for this feature. JASP also does not run a MANOVA or MANCOVA.

How do I save my output?

There are a few different options. (1) You can save your output in JASP. Simply, select File > Save and select a destination. (2) You can export your data. At the moment, JASP only allows you to export the output as an HTML file, meaning you can only open it in a web browser. However, you are still able to print this page normally. (3) You can copy and paste your output from JASP into word. Simply, click the downward arrow next to the analysis test name and select copy, then paste your output into Word or PowerPoint.

Introduction to JASP

Step	Action	Result
Downlo	bading JASP	
1.	Open web browser (recommended: Google Chrome) and type in <u>https://jasp-</u> <u>stats.org/download/</u>	
		JASP is available for: Microsoft Windows Mac OS X Max OS X Linux See also: JASP 0.8.2 RELEASED August 21th, 2017. New features and improvements of this JASP version include split-by descriptives, exact p-values, post-hoc analysis Bayesian ANOVA and more. For additional details see the release notes below.
2.	Select your OS (whether Windows or MAC) and click "Download." If you have a Linux processor, continue to the download instructions listed on the webpage.	JASP is available for: Mac OSX Mac O
3.	As the file downloads, open file, and select "YES" to run the installation, and then click "Next" and "I Agree."	ASP 0.8.2.0 Welcome to the JASP 0.8.2.0 Setup Wizard This wizard will guide you through the installation of JASP 0.8.2.0. This recommended that you close all other applications before starting Setup. This will make it possible to update relevant system files without having to reboot your computer. Click Next to continue. Next > Cancel

Step	Action	Result				
4.	Select "Finish"	Image: Insertion of the section of				
	ing Data Into JASP					
1.	In order to analyze data in JASP, the file must be in CSV format. To convert an Excel, or spreadsheet File into a CSV extension file, click "File," "Save As" and "CSV" and save the file somewhere on your computer.	Format: Comma Separated Values (.csv)				
2.	The top row in your excel file will become your variable names once the file is opened in JASP.	Diet Crackers Calories 1 1 850 1 1 875 1 1 825 Diet Crackers Calories 1 1 850 2 1 1 875 3 1 1 825				

Step	Action	Result
3.	To import into JASP, open JASP select "File," "Open," and "Computer" and select the saved CSV File. For this manual, will we be using the Examples that JASP includes. To follow along, this is where we would click "Examples" and select the appropriate file for the test we are conducting.	JASPFileCommonOpenRecentSaveComputerSave AsOSFExport ResultsOSFExport DataExamplesSync DataLineCloseLine
4.	Once the file is open, the data should appear in the "Common" tab.	Bookt Common Common

End of procedure

Descriptive Statistics

Descriptive statistics and reliability tell us basic features about our data. Descriptive statistics can give us things such as measures of central tendency and measures of dispersion to describe the data. Reliability refers to a scale's ability to consistently measure a variable.

Step	Action	Result						
Descri	ptive Statistics							
We will use Descriptive Statistics to discover basic information about our data. This can be used to simply get an idea about the data or to compare groups as follow-up information to an omnibus test. We will be using the example data set " Bugs ."								
1.	Click "Descriptives" at the top left- hand corner and "Descriptive Statistics" from the drop-down menu.	File Common Descriptives Image: Common model Image: Commodel Image						
2.	Add the variables you are interested in to Variables. We can add a single variable or as many as we wish. When comparing two variables, JASP will provide us with common descriptive information for the variables. Explore other options given in the menu space to include any other descriptive information you would like. For the sake of information purposes, all possible characteristics are included in the example. When you have made your selections, click "OK."	File Common Descriptives Image: Common image: C						

Step	Action	Result
3.	Your results will be displayed in the right-hand space.	Descriptives
	Among other things, this provides	Descriptive Statistics
	information about measures of	Lo D, Lo F Lo D, Hi F
	central tendency, measures of	Valid 93 91
	dispersion, and the normal curve.	Missing 0 2 Mean 5.715 7.379
		Std. Error of Mean 0.2810 0.2641
		Median 6.000 8.000
		Mode 3.000 ^a 10.00 Std. Deviation 2.710 2.519
		Variance 7.345 6.346
		Skewness -0.1318 -0.9472
		Std. Error of Skewness 0.2500 0.2527
		Kurtosis –0.7606 0.1598
		Std. Error of Kurtosis 0.4952 0.5003
		Range 10.00 9.500 Minimum 0.000 0.5000
		Maximum 10.00 10.00
		Sum 531.5 671.5
		25th percentile 3.500 6.000
		50th percentile 6.000 8.000
		75th percentile 7.500 9.500
		25th percentile 3.500 6.000 50th percentile 6.000 8.000
		75th percentile 7.500 9.500
		* More than one mode exists, only the first is reported
4.	If you scroll down further in the results section, plots and visuals will also appear such as the example included here.	Lo D, HI F
		Lo D, Lo F

Reliab	ility Analysis				
We wi	Il be using the example data set "Bug	IS."			
1.	Click "Descriptives" at the top left- hand corner and "Reliability Analysis" from the drop-down menu.		Common	NOVA Region North North Europe North North Europe Europe	sion Frequencies Frequencies Education some advance college college some some some
2.	The Scale Statistics list refers to a series of test statistics that JASP can produce to determine reliability of the measurements for all the variables on the whole. In other words, with higher reliability, results should be consistent under the same conditions. Individual Item Statistics include the reliability test values (if an item dropped), and the mean, standard deviation and item-rest correlation. In the Reverse Coding Items area, we can have the scale of the variable reverse (e.g. on a scale from 1-7, 1 becomes 7 and 7 becomes 1, etc.). This will cause the scale and item statistics in the output to change to reflect the changes in the way that the variable is scaled.	Gut Gut Gut Gre Gre Gut Gut Gut Me Gut Me Star	enbach's α tmann's λ6 Donald's ω eatest lower bound erage interitem corr an ndard deviation tal Item Statistics enbach's α (if item dr tmann's λ6 (if item d	opped) ropped)	

3.	Your results will be displayed in the right-hand space.	Reliability Analysis							
		Scale Reliability	Statistics						
			mean	sd	Cronbach's α Gutma	ann's λ6 McDona	ds'ω Average in	teritem correlation	
		scale	6.547	2.302	0.731	0.577 0.	732	0.577	
		Note. Of the ob	oservations,	91 were u	sed, 2 were excluded, and	193 were provided.			
		Item Reliabilit	y Statistics				If item dropped		
			mean	sd	item-rest correlation	Cronbach's α	Gutmann's λ6	McDonalds' ω	
		Lo D, Lo F	5.714	2.666	0.577	0.333	NaN		
		Lo D, Hi F	7.379	2.519	0.577	0.577	0.333		

T-Test

A T-test is a parametric test that is used to evaluate the difference between two means. The means can be independent from one another (there are different participants across the two groups) or they can be dependent (the same participants were measured on two occasions). Alternatively, a researcher may only have one mean and he or she would like to test it against a known or hypothesized mean. These three analyses, Independent Samples, Paired Samples, and One Sample, are discussed below.

Step	Action	Result									
Indepe	Independent Samples										
	The Independent Samples T-Test allows you to test the null hypothesis that the means of two independent groups are equal.										
	l be using the example data set " Kito nce in age based on sex. Here, sex g										
1. Click "T-Tests" at the top left- hand corner and "Independent Samples T-Test" from the drop- down menu.											
		Participa	Independent Samples T-Test Paired Samples T-Test	6	4 q1_NEO						
		2 2	One Sample T-Test	5	5						
		Bayesi	Bayesian Independent Samples T-Test Bayesian Paired Samples T-Test	3	4						
		4 4	Bayesian One Sample T-Test	5	3						
		5 5	1 3	3	1						
		6 6	2 3	1	3						
				25.01 12							

Step	Action	Result
2.	 Add your Dependent Variable and Grouping Variable. As mentioned, in this example, we will be using Age as a dependent variable and Sex as a grouping variable. Keep in mind the following constraints: The dependent variable must be scale The grouping variable must have exactly 2 levels Check "Student" under Tests and "Group 1 ≠Group 2" under Hypothesis. Explore other options given in the menu space. For instance, it may be useful to look at Assumption Checks, as an Independent Samples <i>t</i>-Test assumes Normality and Equality of Variance. It is also helpful to select "Effect Size," under Additional Statistics. When you have made your selections, click "OK." 	g5_NEO g6_NEO g7_NEO g8_NEO g9_NEO g10_NEO g11_NEO g12_NEO g3_check g4_check g1c_Neco g1c_Neco
3.	Your results will be displayed in the right-hand space. In this case, there is no evidence of a significant difference in age between males and females, as our <i>p</i> -value is greater than .05.	T-Test Independent Samples T-Test t df p Age -1.329 100.0 0.187 Note. Student's T-Test.

Paired Samples

The Paired Samples T-Test allows you to test the null hypothesis that the means of two dependent groups are equal.

We will be using the example data set "Bugs."

1.	Click "T-Tests" at the top left- hand corner and "Paired Samples T-Test" from the drop-down menu.	Fil Desc 1 1 2 2 3 3	criptives	Lo D, Lo F 6 10 5			
		4 4		-	ne Sample T-Test		6
		5 5 6 6		Female Female	North Europe	some some	3 2
2.	Add your conditions to right box. In this example, we will be using Lo D, Lo F and Lo D, Hi F as the levels that we would like to compare. Keep in mind the following constraints: • The variable must be scale Check "Student" under Tests and "Group 1 ≠Group 2" under Hypothesis. Explore other options given in the menu space. For instance, it may be useful to look at Assumption Checks, as a Paired Samples <i>t</i> - Test assumes Normality. It is also helpful to select "Effect Size," under Additional Statistics. When you have made your selections, click "OK."		Hypothesis Measure Measure 	n signed rank 1 ≠ Measure 2 1 > Measure 2 1 < Measure 2 1 < Measure 2	N E C C U U Missir Sir E E E	ional Statistics Mean difference Effect size Confidence inter Interval Descriptives Descriptives ploi Confidence i /ovk-Sellke Max ng Values	rval 95 % ts interval 95 % kimum p-Ratio

3.	Your results will be displayed in the right-hand space.	T-Test
	In this case, there is evidence of	Paired Samples T-Test
	a significant difference between	t df p
	the two conditions, with a <i>p</i> -value less than .05.	Lo D, Lo F - Lo D, Hi F -6.649 90 < .001
		Note. Student's T-Test.
One S	ample	
The Or mean.	ne Sample T-Test allows you to test t	the mean of a set of data against a known or hypothesized
	I be using the example data set " Kitc from the average age of a general po	chen Rolls." Suppose we wondered if this group's age opulation.

1.	1. Click "T-Tests" at the top left- hand corner and "One Sample T- Test" from the drop-down menu.		File Cor	nmon	ANOVA *	Regression Fre	equencies Facto
			📏 Participar		ident Samples T amples T-Test	-Test	🚴 q2_check
		1	1	Paired Samples T-Test One Sample T-Test		6	
		2	2	Bavesia	n Independent S	Samples T-Test	5
		3	3		n Paired Sample		3
		4	4	Bayesia	n One <mark>Sample</mark> T	-Test	5
		5	5		1	3	3
		6	6		2	3	1

2.	 Add your dependent variable. As mentioned, in this example, we will be using Age as a dependent variable. Keep in mind the following constraints: The dependent variable must be scale 	 q9_NEO q10_NEO q11_NEO q12_NEO q3_check q4_check include Rotation Sex Student Major.Occupation 	Age OK
	Check "Student" under Tests and "≠ Test Value" under Hypothesis. Under Test Value, input the number that you would like your group to be compared against. In this example, you may find research indicating that the general population in the United States has an average age of 36 years.	Tests ✓ Student Mann-Whitney U Test value: 36 Hypothesis ✓ Test value ✓ Test value ✓ Test value	Additional Statistics Mean difference Effect size Confidence interval Interval 95 % Descriptives Descriptives plots Confidence interval 95 % Vovk-Sellke Maximum p-Ratio Missing Values Exclude cases analysis by analysis
	Explore other options given in the menu space. For instance, it may be useful to look at Assumption Checks, as a One Sample <i>t</i> -Test assumes Normality. It is also helpful to select "Effect Size," under Additional Statistics. When you have made your selections, click "OK."	Assumption Checks	Exclude cases listwise
3.	Your results will be displayed in the right-hand space. In this case, there is evidence of	T-Test One Sample T-Test	
	a significant difference between this group and our test value, as	t	df p
	our <i>p</i> -value is less than .05. By going back to Step 3 and clicking	Age -30.42	101 < .001
	"Descriptives" under Additional Statistics, you will see that the group's average age was 22.10, which our test has just told us is significantly less than 36.	Note. Student's T-Test. Note. All tests, hypothesi different from 36.	is is population mean is

ANOVA

An ANOVA is a parametric test that is used to evaluate the difference between groups. It can be used with at least two groups (e.g. men and women), and is preferred over a t-Test when there are three or more groups (e.g. primary school, high school, graduate school). It can also be used to evaluate more than one independent variable (e.g. gender and educational achievement) and see if there is an interaction between the variables. The ANOVA is an omnibus test which means that the outcome of the test only tells you if there is a difference somewhere among the groups. For a variable with more than two levels (e.g. educational achievement, here), follow-up tests are required to determine which groups differ.

Step	Action	Res	sult		
ANOV	A				
We wil	NOVA allows you to test the null hype I be using the example data set " Toc ength and how those variables may	oth G	rowth.'	" Suppose	d we wondered about variables that
1.	Click "ANOVA" at the top left- hand corner and "ANOVA" from the drop-down menu.		File	Common T-Tests VC VC VC VC VC VC VC VC VC VC	Image: ANOVA Image: ANOVA ANOVA Repeated Measures ANOVA ANCOVA Bayesian ANOVA Bayesian ANOVA Bayesian ANOVA Bayesian ANOVA Bayesian ANOVA 500 500 500 500

Step	Action	Result
2.	 Add your Dependent Variable and Fixed Factors. For the ANOVA, "Fixed Factors" refers to your independent variable(s). As mentioned, in this example, we will be using Len as a dependent variable and Supp and Dose as independent variables. Keep in mind the following constraints: The dependent variable must be scale The fixed factors most be ordinal categorical Explore other options given in the menu space. For instance, it may be useful to look at Assumption Checks, as an ANOVA assumes Normality (Q-Q plot of residuals) and Homoscedasticity (Homogeneity tests). It is also helpful to select "Estimates of effect size" and "partial η²." When you have made your selections, click "OK." 	Descriptives Image: State of the state of
3.	Your results will be displayed in the right-hand space.	ANOVA
	la dela seconda de la 1970 - A	ANOVA - len
	In this case, there is a significant	Cases Sum of Squares df Mean Square F p η_p^2
	main effect of Supp, a significant	supp 205.4 1 205.35 15.572 <.001 0.224 dose 2426.4 2 1213.22 92.000 <.001
	main effect of Dose, and a	supp * dose 108.3 2 54.16 4.107 0.022 0.132
	significant interaction between	Residual 712.1 54 13.19
	Supp and Dose. All these effects	Note. Type III Sum of Squares
	have a <i>p</i> -value less than .05.	

Step	Action	Result
4.	As the ANOVA showed a significant overall effect in each main effect case and in the case of the interaction, some follow-up tests are required. Supp only has two levels, so we do not need a follow-up test. However, dose has three levels. Click on the bar labeled Post Hoc Tests. In the left box there will be a list of your factors and below that there will be a list of four tests (Tukey, Scheffe, Bonfferoni, and Holm). In this case, we will select "Tukey" and move Dose to the right.	 Model Assumption Checks Contrasts Post Hoc Tests supp dose dose Correction Tukey Scheffe Bonferroni Holm Descriptives Plots
5.	The results of this test will be displayed in the right-hand space. The Tukey gives each combination of Dose level comparisons. The p _{tukey} column gives the p-value. The differences between all the combinations are significantly different, as our p- values are less than .05. It is also information to ask for	Mean Difference SE t ptukey 500 1000 -9.130 1.148 -7.951 < .001
	Marginal means. This can be found under Additional Options. Move the effects to the right side.	Marginal means Supp dose supp * dose □ Compare main effects Confidence interval adjustment None □ Display □ Descriptive statistics ✓ Estimates of effect size ✓ η² □ partial η² ✓ Novk-Sellke Maximum p-Ratio

	Action	Result					
7.	The results will be displayed in the right-hand space.	Marginal	Means				
	Marginal Means for Supp is tells		eans – supp				_
	us that the mean for OJ is higher	supp	Marginal M		Lower Cl	Upper Cl	-
	than the mean for VC. This is informative, as, with two levels,	OJ VC	20.6 16.9		19.33 15.63	21.99 18.29	
	we did not perform a Tukey. We now know which level is higher	Marginal M	eans – dose				
	and that the difference is	dose	Marginal M	ean SE	Lower Cl	Upper Cl	-
	significant because we had a main effect of Supp from the original output. This output, as	500 1000 2000	10.6 19.7 26.1	0.812	8.977 18.107 24.472	12.23 21.36 27.73	
with the latter two outputs, give us means and confidence		Marginal M	eans - supp *	dose			
	intervals that we should report	supp	dose	Marginal Mean	SE	Lower CI	Upper Cl
	when describing the analysis.	OJ	500 1000 2000	13.230 22.700 26.060	1.148 1.148 1.148	10.928 20.398 23.758	15.53 25.00 28.36
		VC	500 1000 2000	7.980 16.770 26.140	1.148 1.148 1.148	5.678 14.468 23.838	10.28 19.07 28.44
Repea	ted Measures ANOVA					25.050	20.44
The Romann	ted Measures ANOVA epeated Measures ANOVA differs fro le conditions. Thus, while the above A A is for within-subjects.	ANOVA wa	DVA beca	use the sar	ne partic	cipants a	re in
The Romultipl ANOV We wi	epeated Measures ANOVA differs fro le conditions. Thus, while the above / A is for within-subjects. Il be using the example data set " Bug	ANOVA wa gs."	DVA beca is for betw	use the sar	ne partic	cipants a	re in
The R multipl ANOV	epeated Measures ANOVA differs fro le conditions. Thus, while the above <i>A</i> A is for within-subjects. Il be using the example data set " Bug Click "ANOVA" at the top left- hand corner and "Repeated Measures ANOVA" from the	ANOVA wa	DVA beca	use the sar veen-group	ne partic s, the Re	cipants a	re in
The Ro nultipl ANOV We wi	epeated Measures ANOVA differs fro le conditions. Thus, while the above <i>i</i> A is for within-subjects. Il be using the example data set " Bug Click "ANOVA" at the top left- hand corner and "Repeated	ANOVA wa	DVA beca s for betw Common	use the sar veen-groups	ne partic s, the Re Regression F Reasures ANG	cipants a epeated I	re in Measures
The Romultipl ANOV We wi	epeated Measures ANOVA differs fro le conditions. Thus, while the above <i>A</i> A is for within-subjects. Il be using the example data set " Bug Click "ANOVA" at the top left- hand corner and "Repeated Measures ANOVA" from the	ANOVA wa	OVA beca is for betw Common	ANOVA Repeated M ANOVA Bayesian At Bayesian At Bayesian At Bayesian At Bayesian At	ne partic s, the Re Regression F leasures ANG	cipants a epeated I requencies	re in Measures
The Romultipl ANOV We wi	epeated Measures ANOVA differs fro le conditions. Thus, while the above <i>A</i> A is for within-subjects. Il be using the example data set " Bug Click "ANOVA" at the top left- hand corner and "Repeated Measures ANOVA" from the	ANOVA was gs." File Descriptives Sut 1 1 2 2 3 3	Common Common Liff T-Tests oject Cemal Femal Femal	ANOVA ANOVA ANOVA ANOVA ANOVA Repeated M ANCOVA Bayesian At Bayesian At Bayesian At Bayesian At Bayesian At Bayesian At	ne partic s, the Re regression F leasures ANG NOVA epeated Mea NCOVA	cipants a epeated I Prequencies	re in Measures

Step	Action	Result
2.	For each repeated measure or the within-subjects variable, label the variable under Repeated Measures Factors in "RM Factor" and then the levels of the variable in "Level."	 Subject Gender Region Education Lo D, Lo F Lo D, Hi F Hi D, Lo F Hi D, Hi F Hi D, Hi F Hi D, Hi F
3.	The repeated measures of Lo D, Lo F, Lo D, Hi F, Hi D, Lo F, and Hi D, Hi F should be placed in each section within the Repeated Measures Cells box, aligning with the appropriate labeling which JASP created based on your Repeated Measures Factors. Keep in mind the following constraints: • The variables must be scale If we were using a mixed- methods design, with both within- subjects and between-subjects variables, we could put the between-subjects variables in Between Subject Factors, as well. For these variables, keep in mind the following constraints: • The variables must be ordinal or categorical Explore other options given in the menu space. For instance, it may be useful to look at Assumption Checks, as a Repeated Measures ANOVA assumes Sphericity. It is also helpful to select "Estimates of effect size"	Repeated Measures Factors D Lo Hi Level 3 F Lo Hi Level 3 RM Factor 3 Repeated Measures Cells No D, Lo F Lo D, Lo F Lo D, Hi F Lo D, Hi F Hi D, Hi F Hi Hi D, Hi F Hi Hi D, Hi F Between Subject Factors Image: Comparison of the
	and "partial η^2 ." When you have made your selections, click "OK."	

Step	Action	Result						
4.	Your results will be displayed in the right-hand space.	Repeated	Measures	ANOVA	· · · · · · · · · · · · · · · · · · ·			
	la this same those is a simulticant	Within Subject	ts Effects					
	In this case, there is a significant		Sum of Squares	df	Mean Square	F	р	$\eta_{\rm P}^{\rm 2}$
	main effect of D and a significant main effect of F, with <i>p</i> -values	D	48.753	1	48.753	12.175	< .001	0.123
	less than .05. There is not	Residual F	348.372 177.557	87 1	4.004 177.557	41.630	< .001	0.324
	evidence for a significant	Residual	371.068	87	4.265	41.050	0.001	0.524
	interaction.	D * F	6.545	1	6.545	2.152	0.146	0.024
		Residual Note. Type III	264.580 Sum of Squares	87	3.041			
	Follow the steps given in the							
	ANOVA "Steps 4-7." In this							
	example, we do not need a post-							
	hoc test because there are only							
	two levels of each of our							
	variables.							
Step	Action	Result						
Step ANCO		Result						
ANCO The Al control		pothesis the	outcome		• •	are the	e same	while
ANCO The Al control We wil	VA NCOVA allows you to test the null hy ing for an extraneous variables that a I be using the example data set " Toc	pothesis the	outcome \ h."		• •	are the	e same	while
ANCO The Al control	VA NCOVA allows you to test the null hy ing for an extraneous variables that a I be using the example data set " Toc Click "ANOVA" at the top left- hand corner and "ANCOVA" from	pothesis the affects the oth Growtl	outcome \ h." mon +	∕ariabl	e.	≠	.	while
ANCO The Al control We wil	VA NCOVA allows you to test the null hy ing for an extraneous variables that a I be using the example data set " Toc Click "ANOVA" at the top left-	pothesis the affects the oth Growtl	outcome \ h." mon + T-Tests A	variabl	e.	≠		while
ANCO The Al control We wil	VA NCOVA allows you to test the null hy ing for an extraneous variables that a I be using the example data set " Toc Click "ANOVA" at the top left- hand corner and "ANCOVA" from	pothesis the affects the oth Growtl	outcome v h." mon + T-Tests A	variabl	e.	Juencies	.	while
ANCO The Al control We wil	VA NCOVA allows you to test the null hy ing for an extraneous variables that a I be using the example data set " Toc Click "ANOVA" at the top left- hand corner and "ANCOVA" from	pothesis the affects the oth Growtl	outcome \ h." mon + T-Tests A	variabl	e. Regression Free Red Measures AN	Juencies	.	while
ANCO The Al control We wil	VA NCOVA allows you to test the null hy ing for an extraneous variables that a I be using the example data set " Toc Click "ANOVA" at the top left- hand corner and "ANCOVA" from	pothesis the affects the oth Growtl Descriptives	outcome v h." mon + T-Tests A VC VC VC	ANOVA Repeate ANCOV	e. Regression Free Red Measures AN	Juencies	.	while
ANCO The Al control We wil	VA NCOVA allows you to test the null hy ing for an extraneous variables that a I be using the example data set " Toc Click "ANOVA" at the top left- hand corner and "ANCOVA" from	pothesis the affects the oth Growtl	outcome v h." mon + T-Tests A VC VC	Ariabl	e. egression Fred ed Measures AN	Juencies	Factor	while

Step	Action	Result
2.	Add your Dependent Variable and Fixed Factors. For the ANOVA, "Fixed Factors" refers to your independent variable(s). This time, the display gives you the option to add a Covariate. In this example, we want to know the effect of Supp on Len, while controlling for the effects of Dose. Keep in mind the following constraints: • The dependent variable must be scale • The fixed factors most be ordinal or categorical • The covariate can be scale or categorical Explore other options given in the menu space. For instance, it may be useful to look at Assumption Checks, as an ANOVA assumes Normality (Q-Q plot of residuals) and Homoscedasticity (Homogeneity tests). It is also helpful to select "Estimates of effect size" and "partial η^2 ." When you have made your selections, click "OK."	Dependent Variable
3.	Your results will be displayed in the right-hand space. In this case, there is evidence of a significant effect of the Covariate, Dose. There is also a significant effect of Supp over and above the effect of Dose, as	ANCOVA - len Cases Sum of Squares df Mean Square F p supp 205.3 1 205.35 11.45 0.001 dose 2224.3 1 2224.30 123.99 <.001
	these <i>p</i> -values are less than .05. Follow the steps given in the ANOVA "Steps 4-7." In this example, we do not need a post- hoc test because there are only two levels of each of our variables.	

End of procedure

Regression

Correlation and regression tell you about the relationship between variables. Correlation describes if the variables tend to increase or decrease together or go in opposite directions. Regression is a form of statistical modeling that examines how much variance is explained by the model we have created with various variables.

Step	Action	Result
Correla	ation	
correla We wil	ted (they move in opposing direction	y related (they move in the same direction) or negatively s). 5." Suppose we were interested in the relationship
1.	Click "Regression" at the top left- hand corner and "Correlation Matrix" from the drop-down menu.	FileCommonDescriptivesImage: TrestsImage: Trests<
2.	Add your variable of interest. As mentioned, in this example, we will be using Openness and Agreeableness. Keep in mind the following constraints: • The variables must be scale or ordinal Explore other options given in the menu space. For instance, the correlation coefficient will depend on what type of variables we are working with. In this case, both variables are scale, so we will use Pearson. We can also ask it to report significant (<i>p</i> -values) and to flag significant correlations to tell us when a <i>p</i> -value indicates significance. When you have made your selections, click "OK."	Neuroticism Extraversion Conscientiousness

Step	Action	Result					
3.	Your results will be displayed in the right-hand space.	Correlation Matrix					
	The matrix matches each variable with the others that we selected.	Pearson Correlations	5				
	In this case, there is evidence of			Openness	Agreeableness		
	a significant positive correlation between Openness and Agreeableness, as the <i>p</i> -value is less than .05. This is a small	Openness	Pearson's r p-value	_	0.159***		
		Agreeableness	Pearson's r p-value		_		
	effect (0.159).	* p < .05, ** p < .01	-				
Regres	sion						
We will	l be using the example data set "Big	5."					
1.	Click "Regression" at the top left-	File Common					
	hand corner and "Linear Regression" from the drop-down menu.	Descriptives		Regression Frequ	encies Factor		
	incha.	Neuroticism	Straversion	Correlation Mat	eeableness		
		1 2.47917	4.20833	Bayesian Corre	0.05839		
		2 2.60417	3.1875	Bayesian Corre	lation Pairs		
		3 2.8125	2.89583	Bayesian Linea	r Regression		
		4 2.89583	3.5625	3.52083	3.16667		
		5 3.02083	3.33333	4.02083	3.20833		
		6 2.52083	3.29167	3.4375	3.70833		

Step	Action	Result							
2.	Add your variable of interest. In this example, we will be using Neuroticism as the Dependent Variable and Extraversion as the Covariate. In Regression, a "Covariate" refers to variables that will be includes in the model. Keep in mind the following constraints: • The variables must be scale Explore other options given in the menu space. For instance, the Method can be "Enter," "Backward," "Forward," or "Stepwise," referring to the order in which our Covariates (if we have multiple) will be entered into the model. When you have made your selections, click "OK."		ess bleness entiousness		•	Dependent Neuroticism Method Covariates Extraversio WLS Weights	Enter		OK
3.	Your results will be displayed in the right-hand space. In this case, R^2 tells us that 12.3% of the total variance is explained by the model including Extraversion. The Coefficients output tells us that Extraversion is a significant predictor for the model, as the <i>p</i> -value is less than .05.	Linear R Model Summa Model 1 ANOVA Model 1 Coefficients Model 1	Regression Regression Regression Residual Total	R ² 0.123 Sum o	Adjuste 0.1 12.53 89.71 102.24 andardized 4.383 -0.446	0.424	an Square 12.530 0.180 Standardized -0.350	F 69.56 t 23.366 -8.340	p < .001

Frequencies

Step	Action	Result			
Binomi	al Test				
We wil	I be using the example data set "Kito	hen Rolls."			
1.	Click "Frequencies" at the top left- hand corner and "Binomial Test" from the drop-down menu.	File Common	ANOVA Regres	sion Frequencies	Factor
		ParticipantNumber 1 1 2 2 3 3 4 4 5 5	Condition	Contingency Ta 2 Log-Linear Reg 3 Bayesian Binom 7 Bayesian Contin	nial Test
		6 6	2	3 1	3
2.	 Add your variable of interest. Keep in mind the following constraints: The variable must be ordinal or categorical Explore other options given in the menu space. When you have made your selections, click "OK." 	 ParticipantNumber Condition q1_check q2_check q1_NEO q2_NEO q3_NEO q4_NEO q5_NEO q6_NEO 		Student	OK
3.	Your results will be displayed in the right-hand space.	Binomial Test			
	In this case, we wondered how many participants are students and how many are not. The output gives us the count for each level of this categorical variable.	Level Student N Y Note. Proportions tested	7 95	Total Proportion 102 0.069 102 0.931 5. 5.	p < .001 < .001

Contin	gency Tables				
We wi	I be using the example data set "Kite	chen Rolls."			
1.	Click "Frequencies" at the top left- hand corner and "Contingency Tables" from the drop-down menu.	File Common Descriptives Image: Common market ParticipantNumber 1 2 3 4 5 6	Image: Anotal state sta	Bayesian Binomial Tes Bayesian Contingency Bayesian Log-Linear R 5 3 3	t Tables
2.	Add your variables of interest (two different variables) to Rows and Columns. Keep in mind the following constraints: • The variables must be ordinal or categorical Explore other options given in the menu space. When you have made your selections, click "OK."	 q1_NEO q2_NEO q3_NEO q4_NEO q5_NEO q6_NEO q7_NEO q9_NEO q10_NEO q11_NEO q12_NEO q3_check q4_check Include Rotation Age Major.Occupation 	Rows Rows Sex Column Column Column Layers Layers	📣 👔	ОК

3.	Your results will be displayed in the right-hand space.	Conting	genc	y Tal	bles		
	The main output tells us the association between two variables. In this case, we see	Contingend	y Table	es			
	that there is a large discrepancy between the number of women		Stu	dent			
	included and the number of men included. Related to Student, the	Sex	Ν	Y	Total		
	majority of both women and men	F	6	71	77		
	are students.	М	1	24	25		
		Total	7	95	102		
		Chi-Square		s Ilue	df	р	
		X ² N	0	.425 102	1	0.515	_
•	near Regression I be using the example data set " Kitc	hen Rolls."					
1.	Click "Frequencies" at the top left- hand corner and "Log-Linear Regression" from the drop-down	File Comm	-	<u>ĮĮ</u> ,	· <u>*</u> * = _ *		
	menu.	Descriptives T-		ANOVA Re	gression Frequencies Binomia	Factor	
		1 1	Number	1	Conting	ency Tables ear Regression	4
		2 2 3 3		2 3	7 Bayesia	n Binomial Test n Contingency Ta n Log-Linear Reg	
		4 4		4	4 Bayesian	5	3
		5 5		1	3	3	1
		6 6		2	3	1	3

2.	 Add your variable(s) of interest to Factors. Keep in mind the following constraints: The variables must be ordinal or categorical Explore other options given in the menu space. When you have made your selections, click "OK." 	 q8_NEO q9_NEO q10_NEO q11_NEO q12_NEO q3_check q4_check Include Rotation Age Major.Occu 	pation		Counts (option Factors		OK
3.	Your results will be displayed in the right-hand space.	Log-Linear	Regre	ssion			
		ANOVA					
	In this case, we were interested		df	Deviance	Residual df	Residual Deviance	p
	in a model produced by these	NULL			3	118.658	
	categorical variables. There is	Sex	1	27.797	2	90.860	< .001
	evidence of a significant influence	Student Sex * Student	1	90.387 0.474	1	0.474 0.000	< .001 0.491
	of Sex and Student with <i>p</i> -values less than .05.		_		-		

Factor Analysis

Factor analysis is a way to describe the variability of correlated variables. The idea is that two or more variables might reflect a single unknown variable. In other words, the aim is to create a fewer number of factors by combining two or more variables. This technique is particularly useful for personality theories, biology, or marketing. These fields may easily hypothesize too many variables, so factor analysis helps to find essentials of a theory. JASP can do two types of factor analysis: principal component analysis (PCA) and exploratory factor analysis (EFA).

	Action	Result	:								
Princip	al Component Analysis										
We wil	I be using the example data set " Pol	itical De	emocra	acy."							
1.	Click "Factor" at the top and "Principal Component Analysis" from the drop-down menu.	4 4 8.9 8.8 10 9.19999 8.90795 8.12798 10 4.61509 6.286 5 5 10 3.3333 10 6.66667 7.5 3.3333 10 6.66667 5.86363							x1 4.44265 5.3845 5.96101		
2.	Add the variables of interest into Included Variables. Keep in mind the following constraints: • The variables must be scale Explore other options given in the menu space. When you have made your selections, click "OK."		Manua	el Analysis values genvalues al imber of F	above	1	Rota	Variable tion Orthogo one Oblique romax	-		OK

Step	Action	Result								
3.	Your results will be displayed in the right-hand space.	Principal	Con	npor	nent	Ana	lysis	;		
	The primary output for a principal component analysis shows the correlation between each variable	Component L	oadings	5						
	of a principal component and the principal component itself		RC	1	RC	2	Uniqu	ieness	;	
	(displayed under the RC 1 and RC 2 columns for the two	x1 x2	0.9					0.100 0.068		
	principal components).	x3 y1	0.9		0.8		(0.008 0.129 0.180		
		y2 y3			0.8	19	(0.356 0.300		
	atory Factor Analysis	itical Democra	acy."							
1.	Click "Factor" at the top and "Exploratory Factor Analysis" from the drop-down menu.	File Common Descriptives Image: Common marked state		- 	vy4 0 0 9.19999 9.19999 6.66667 6.66667	y5 1.25 6.25 8.75 8.90795 7.5 6.25		atory Fact	nent Analys or Analysis 3.33333 0.736999 8.21181 4.61509 6.66667 0.3685	4.44265
2.	Add the variables of interest into Included Variables. Keep in mind the following constraints: • The variables must be scale Explore other options given in the menu space. When you have made your selections, click "OK."	Fite Common Decembers If if if V 1 If if V 2 2 V 3 2 V 4 2 V 7 2 V 8 2 V 9 2 V 1 2 Statist Analysis Egenvalues above (Damber of Factors Rumber of Factors Image: Number of Factors Image: Number of Factors Image: Output options Image: Number of Factors	Reptation Reptation Retation Other pom	nogonal	Sitial Democracy ()	Results Exploratory Factor Loadings Factor Loadings Chi-squared Test	/ Factor Anal			i:

Step	Action	Result								
3.	Your results will be displayed in the right-hand space.	Exploratory Factor Analysis								
	The primary output shows how the variables load into factors.	Factor Loadings								
			Factor 1	Factor 2	Uniqueness	-				
		x1	0.885		0.152					
		x2 x3	0.979 0.895		0.049 0.236					
		y1	0.055	1.007	0.035					
		y2		0.615	0.620					
		y3		0.677	0.514	-				
		Chi-square —	d Test Value	df	p					
		Model	3.761	4	0.439					

Bayesian Methods

JASP offers Bayesian alternatives to many of the analyses. Particularly, there are alternatives for Independent Samples T-Test, Paired Samples T-Test, One Sample T-Test, ANOVA, Repeated Measures ANOVA, ANCOVA, Correlation Matrix, Correlation Pairs, Linear Regression, Binomial Test, Contingency Tables, and Log-Linear Regression. Below we include some examples of procedures for these analyses. Bayesian statistics are an alternative to null hypothesis testing that is becoming increasingly preferred in fields that use statistics.

Step	Action	Res	sult							
Bayesia	Bayesian Independent Samples T-Test									
We will	be using the example data set "Kito	chen	Rolls."							
1.	Click "T-Tests" at the top left- hand corner and "Bayesian Independent Samples T-Test" from the drop-down menu.		scriptives Partic 1 2 3 4 5 6	Independent Sar Paired Samples One Sample T-To Bayesian Indepe Bayesian One Sa 4 1 2	mples T-Test T-Test est ndent Sample Samples T-Te	s T-Test	g2_check 6 5 3 5 3 1			

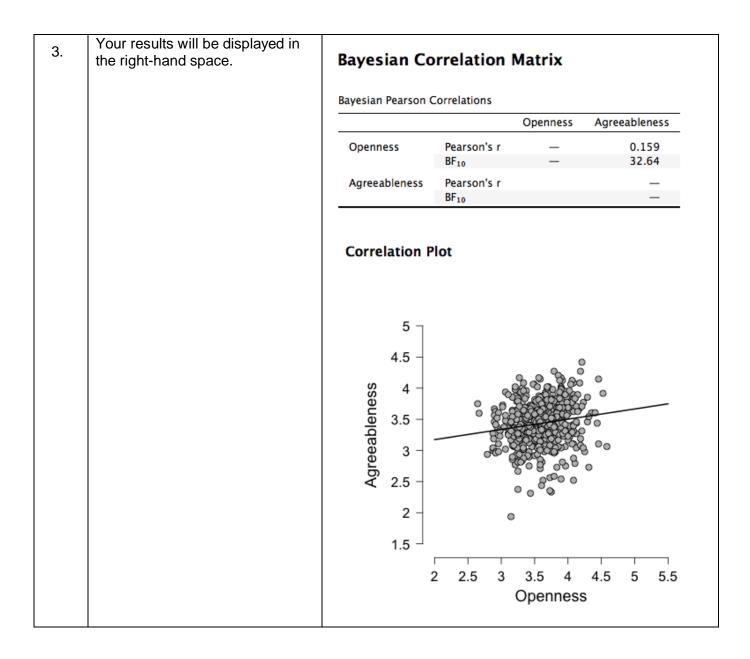
Step	Action	Result		
2.	Add your Dependent Variable and Grouping Variable. In this example, we will be using Age as a dependent variable and Sex as a grouping variable. Keep in mind the following constraints: • The dependent variable must be scale • The grouping variable must have exactly 2 levels When you have made your selections, click "OK."	$ \begin{array}{c} \begin{array}{c} \label{eq:g5_NEO} \\ \label{eq:g6_NEO} \\ \label{eq:g6_NEO} \\ \label{eq:g7_NEO} \\ eq:g7_g7_g7_g7_g7_g7_g7_g7_g7_g7_g7_g7_g7_g$	Sequenti Robust Descripti Credible ir Missing Values Exclude o	iable iable istics ves posterior nal info ctor robustness check al analysis ness check ves plots nterval 95 %
3.	Your results will be displayed in the right-hand space.	Bayesian T-		T-Test
			BF10	error %
		Age	0.509	7.833e – 5

Bayesi	an Paired Samples T-Test					
We wil	I be using the example data set "Bug	js."				
1.	Click "T-Tests" at the top left- hand corner and "Bayesian Paired Samples T-Test" from the drop-down menu.	File Descriptives Subject 1 1 2 2 3 3 4 4 5 5 6 6	t Paired Samp One Sample Bayesian Ind Bayesian Pai		es T-Test	Factor Factor 6 10 5 6 3 3 2
2.	 Add the combination of variables of interest. Keep in mind the following constraints: The variable must be scale When you have made your selections, click "OK." 	Subject Lo D, Lo F Lo D, Hi F Hi D, Lo F Hi D, Hi F Gender Region Education		Lo D, Lo Lo D, Lo Lo D, H Hi D, Lo	o F Hi D, Lo i F Hi D, Hi	F F
		Hypothesis Measure 1 Measure 1 Bayes Factor BF ₁₀ BF ₀₁ Log(BF ₁₀) Prior Cauchy prior	> Measure 2 < Measure 2	A Bayes Sequer Descrip Cred Missing Valu	nd posterior dditional info factor robustness ntial analysis obustness check ptives plots lible interval 95	%

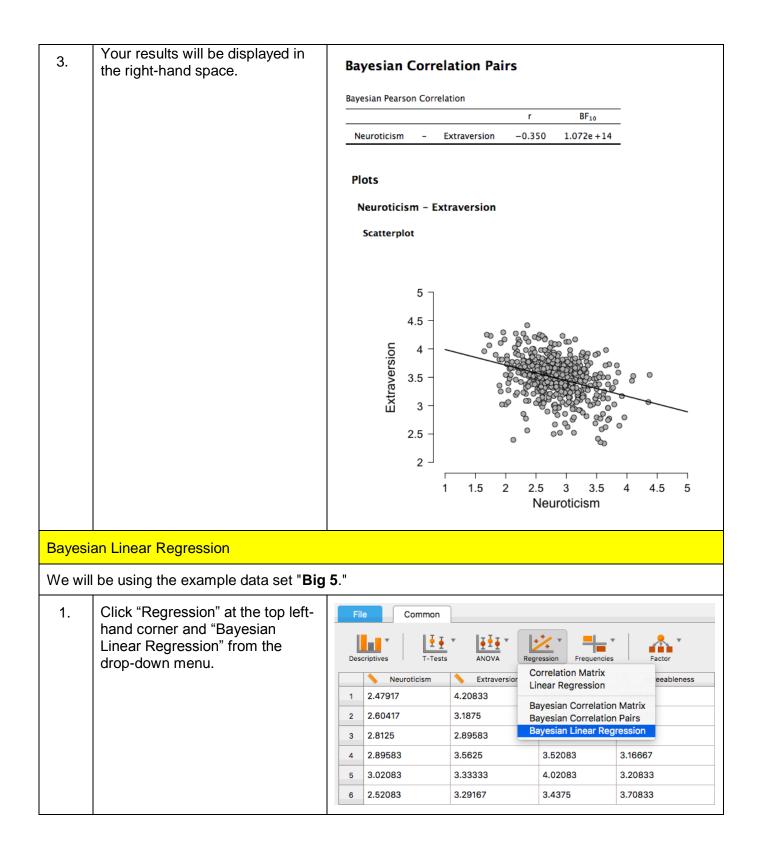
3. Your results will be displayed in the right-hand space.		Bayesian T-Test					
		Bayesian Paired Samples T-Test					
		BF ₁₀ error %					
		Lo D, Lo F – Lo D, Hi F 4.762e +6 1.595e –14					
		Lo D, Lo F – Hi D, Lo F 24.734 3.449e –9					
		Lo D, Hi F – Hi D, Hi F 0.574 2.236e –7					
		Hi D, Lo F – Hi D, Hi F 43.486 1.413e –9					
Bayesi	an Repeated Measures ANOVA						
We wil	l be using the example data set "Bug	gs."					
1.	Click "ANOVA" at the top left- hand corner and select "Bayesian Repeated Measures ANOVA" from the drop-down menu.	File Common Descriptives T-Tests Subject Image: Common Frequencies ANOVA Repeated Measures ANOVA ANCOVA Bayesian ANOVA Bayesian Repeated Measures ANOVA Bayesian ANOVA Femal Bayesian ANCOVA Femal Bayesian ANCOVA					
2.	For each repeated measure or the within-subjects variable, label the variable under Repeated Measures Factors in "RM Factor" and then the levels of the variable in "Level."	Repeated Measures Factors OK D D Lo Hi Level 3 F Lo Hi Lo F Mi E Lo F Mi E Lo Hi Hi E Hi E Hi E Hi E Evel 3 E RM Factor 3 E					

3.	The repeated measures of Lo D, Lo F, Lo D, Hi F, Hi D, Lo F, and Hi D, Hi F should be placed in each section within the Repeated Measures Cells box, aligning with the appropriate labeling which JASP created based on your Repeated Measures Factors. Keep in mind the following constraints:	Repeated Measures Factors OK D Lo Hi Level 3 F Co Hi Hi
	 The variables must be scale 	Level 3 RM Factor 3
		Repeated Measures Cells
	We will use a mixed-methods design, with both within-subjects	📏 Lo D, Lo F 🛛 Lo, Lo
	and between-subjects variables.	📏 Lo D, Hi F 🛛 Lo, Hi
	Add between-subjects variables in Between Subject Factors, as	Hi D, Lo F Hi, Lo
	 The variables must be ordinal or categorical 	Ni D, Hi F Hi, Hi
	Explore other options given in the menu space. When you have made your selections, click "OK."	Between Subject Factors Between Subject Factors Between Subject Fac
4.	Your results will be displayed in the right-hand space.	Bayesian Repeated Measures ANOVA Model Comparison Mull model (incl. subject) Bucation Education Education + D * Education Region Region Education + D * Education + Region Education + D * Education + Region + D * Region Education + Region + D * Region + D * Region Education + Region + Education * Region Education + D * Education + Region + Education * Region Education + D * Education + Region + Education * Region Education + D * Education + Region + Education * Region Education + D * Education + Region + Education * Region Education + D * Education + Region + Education * Region Education + D * Education + Region + Education * Region Education + D * Education + Region + D * Region + Education * Region Education + D * Education + Region + D * Region + Education * Region Education + D * Education + Region + D * Region + Education * Region Education + D * Education + Region + D * Region + Education * Region Education + D * Education + Region + D * Region + Education * Region Education + D * Education + Region + D * Region + Education * Region Education + D * Education + Region + D * Region + Education * Region + D * Education * Region Educat

Bayesi	Bayesian Correlation Matrix					
We wil	l be using the example data set "Big	5."				
1.	Click "Regression" at the top left- hand corner and "Bayesian Correlation Matrix" from the drop- down menu.	2 2 3 2 4 2 5 3	Common Lives Lives Neuroticism .47917 .60417 .8125 .89583 .02083 .52083	Extraversion 4.20833 3.1875 2.89583 3.5625 3.33333 3.29167	Regression Frequencie Correlation Matrix Linear Regression Bayesian Correlation Bayesian Correlation Bayesian Correlation Bayesian Correlation Bayesian Linear Regression 3.52083 4.02083 3.4375	eeableness
2.	Add your variables of interest. Keep in mind the following constraints: • The variable must be scale or ordinal Select "Correlation matrix" under Plots. Explore other options given in the menu space. When you have made your selections, click "OK."	Cor Hyp Bay	Neuroticism Extraversion Conscientiousne Conscientiousne Pearson's rho Rendall's tau-b bothesis Correlated Correlated posi Correlated neg res Factor BF ₁₀ BF ₀₁ Log(BF ₁₀)	ents Plots itively atively	Openness Agreeableness Agreeableness Report Bayes factors Flag supported corre s Correlation matrix Densities for var Posteriors under etched beta prior wide	lations riables r H ₁



Bayesi	an Correlation Pairs				
We wil	be using the example data set "Big	5."			
1.	Click "Regression" at the top left- hand corner and "Bayesian Correlation Pairs" from the drop- down menu.	File Common Descriptives Image: Common for the second se		Regression Correlation Matrix Linear Regression Bayesian Correlat Bayesian Linear R 3.52083 4.02083 3.4375	eeableness ion Matrix ion Pairs
2.	Add your variables of interest. Keep in mind the following constraints: • The variable must be scale Select "Scatterplot" under Plots. Explore other options given in the menu space. When you have made your selections, click "OK."	6 2.52083 3.29167 3.4375 3.4375 S. the Agreeableness Conscientiousness Conscientiousness Correlation Coefficient Pearson's rho Kendall's tau-b Plots Plots Hypothesis Plots Scatterplot Prior and posteri Correlated negatively Additional in Bayes Factor Bayes factor rob BF10 BF201 Log(BF10) Missing Values		nal info r robustness check inalysis es analysis by analysis	



2.	Add your variable of interest. In this example, we will be using Neuroticism as the Dependent Variable and Extraversion as the Covariate. In Regression, a "Covariate" refers to variables that will be includes in the model. Keep in mind the following constraints: • The variables must be scale When you have made your selections, click "OK."	Openness Agreeableness Conscientiousness Covariates Extraversion
3.	Your results will be displayed in the right-hand space.	Model Comparison - Neuroticism Models P(M) P(M data) BFM BF10 error % Null model 0.500 1.415e - 13 1.415e - 13 1.000 Extraversion 0.500 1.000 7.069e + 12 7.069e + 12 1.464e - 6
Bayesi	an Binomial Test	
We will	be using the example data set "Kito	hen Rolls."
1.	Click "Frequencies" at the top left-hand corner and "Bayesian Binomial Test" from the drop- down menu.	FileCommonDescriptivesImage: Common matrixImage: Common matrix
2.	Add your variable of interest. Explore other options given in the menu space. When you have made your selections, click "OK."	q11_NEO q12_NEO mean_NEO q3_check q4_check include Rotation Kotation Kotation

3.	Your results will be displayed in the right-hand space.	Bayesian Binom Bayesian Binomial Test Level Student N Y Note. Proportions tested	Counts Total 7 102 95 102		BF ₁₀ 2.666e + 18 2.666e + 18
	an Contingency Tables	chen Rolls."			
1.	Click "Frequencies" at the top left-hand corner and "Bayesian Contingency Tables" from the drop-down menu.	File Common Descriptives Image: Common marked state sta	Condition Regression 1 2 3 7 4 4 1 3 2 3	Frequencies Binomial Test Contingency Tabl Log-Linear Regres Bayesian Binomial Bayesian Continge Bayesian Log-Line 5 3 1	ssion I Test ency Tables
2.	Add your variables of interest (two different variables) to Rows and Columns. Keep in mind the following constraints: • The variables must be ordinal or categorical Explore other options given in the menu space. When you have made your selections, click "OK."	 q1_NEO q2_NEO q3_NEO q4_NEO q5_NEO q6_NEO q7_NEO q8_NEO q9_NEO q10_NEO q11_NEO q12_NEO q3_check q4_check Include Rotation Age Major.Occupation 	Rows Sex Columns Counts Counts	ent Layer 1	

3.	Your results will be displayed in the right-hand space.	Bayesia	n Co	ontin	gency [·]	Tables	
		Bayesian Contingency Tables					
			Stu	dent		_	
		Sex	Ν	Y	Total		
		F M	6 1	71 24	77 25	_	
		Total	7	95	102		
Bayesi	an Log-Linear Regression	Bayesian Co BF10 inde N		-		Value 0.158 102	
We will	be using the example data set "Kito	hen Rolls."					
1.	Click "Frequencies" at the top left-hand corner and "Bayesian Log-Linear Regression" from the drop-down menu.		rests A		2 Log-Lin 3 Bayesia 7 Bayesia		

2.	Add your variable(s) of interest to Factors. Keep in mind the following constraints: • The variables must be ordinal or categorical Explore other options given in the menu space. When you have made your selections, click "OK."	• q8_NEO • q9_NEO • q10_NEO • q11_NEO • q12_NEO • q12_NEO • q12_NEO • q12_NEO • q3_check • q4_check • q4_check • q4_check • linclude • Rotation • Age • Major.Occupation • Bayes Factor • BF ₁₀ • Shape -1 • Display best 1 model(s) • g1 • Display best 1 model(s) • d1 • OK • O
3.	Your results will be displayed in the right-hand space.	Bayesian Log-Linear Regression
		Models P(M data) BF10
		1 Sex + Student 0.873 1.000
		<i>Note.</i> Total number of models visited = 2

End of procedure

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