EXAMINING YOUR DATA

Chapter 2: Multivariate Insurance

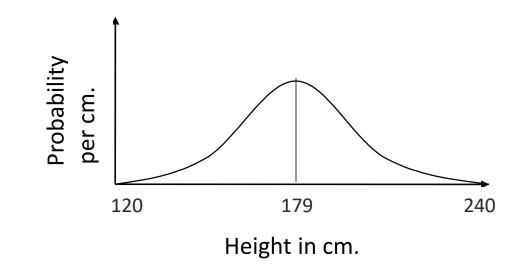
Topics

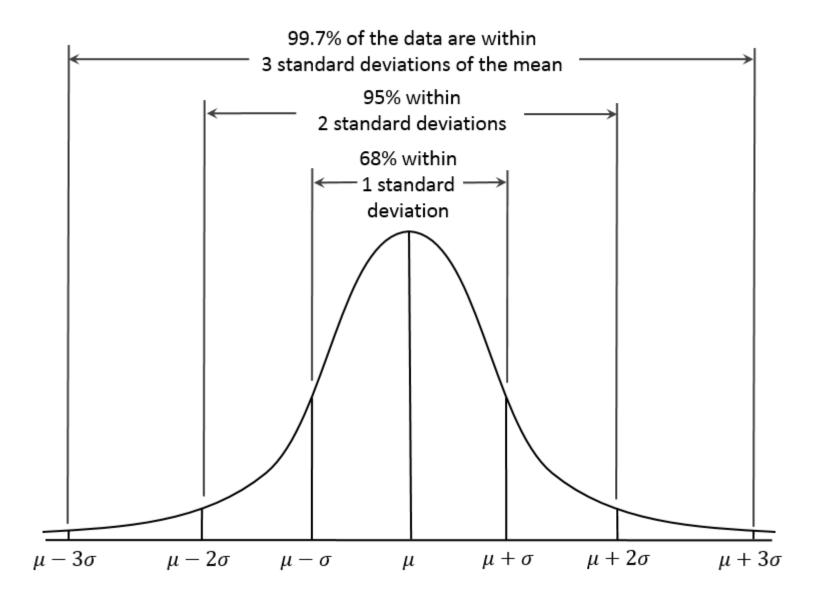
- 1. The normal distribution and normality.
- 2. Relationships between variables (correlation).
- 3. Outliers.
- 4. Missing data.

1. The Normal Distribution

A probability sample of male, 19 year old Swedes had their height measured. The majority were 179 cm tall, with some shorter and some taller.

The distribution is symmetric around the average (the mean \overline{X}).





Checking Normality

You can examine normality visually or statistically. There is no, "one best way", so it is often good to *consider several*.

- Skewness & Kurtosis
- Histograms: not so good with small samples.
- Statistical tests like Kolmogorov-Smirnov or Shapiro-Wilk (n<50).
- Box plots: Handy for identifying outliers.

How normal does the data need to be?

- It depends on the multivariate technique.
- e.g. Linear regression is quite robust against violations of normality.

Now we get SPSS going for everyone in the class

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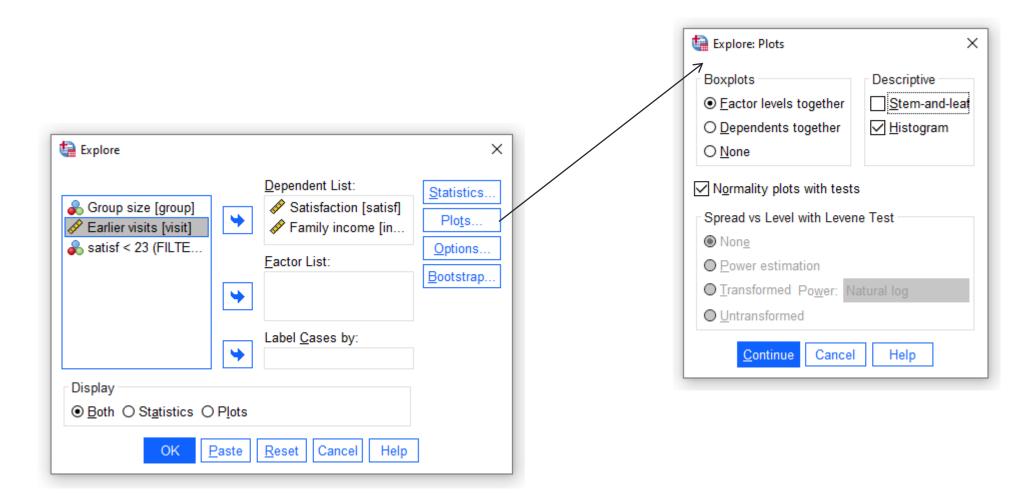
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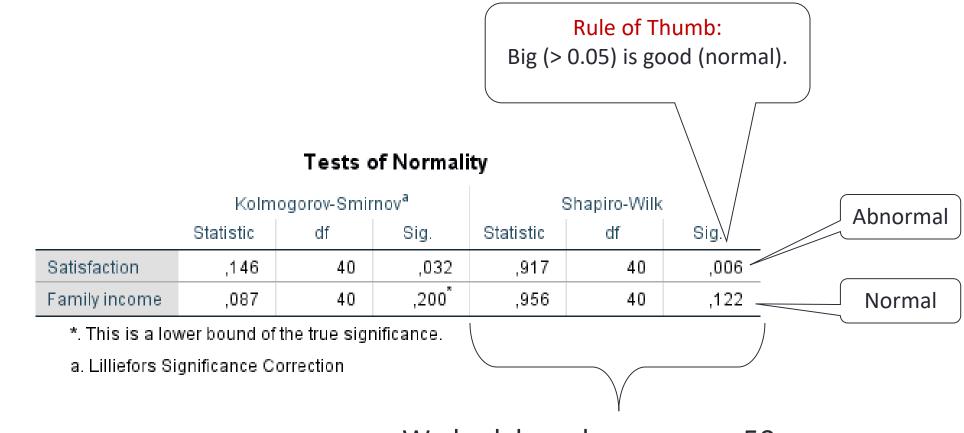




Skewness & Kurtosis

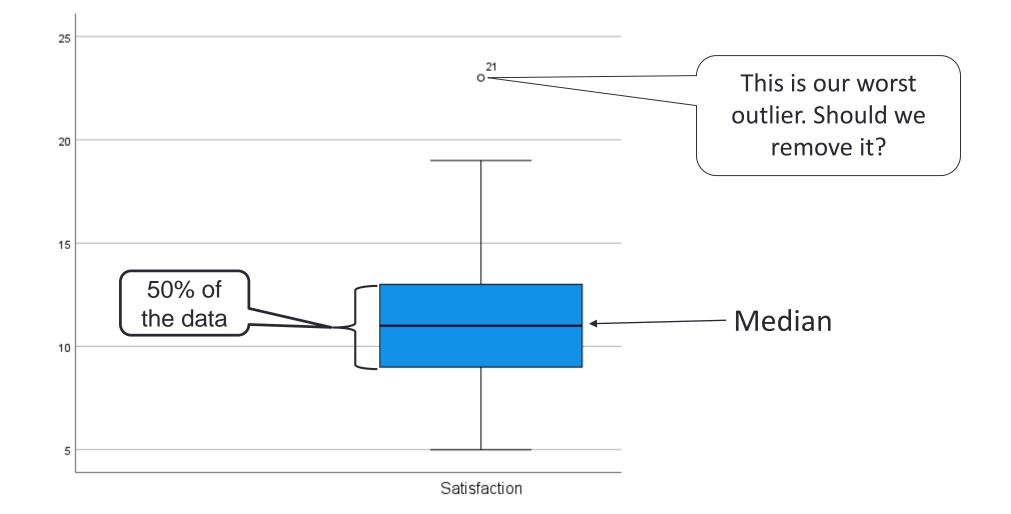
	Desc	riptives			
			Statistic	Std. Error	Rule of Thumb:
Satisfaction	Mean		11,33	,535	Less than 1 (absolute
	95% Confidence Interval	Lower Bound	10,24		
			•		value) is normal
	Skewness		1,188	,374	
	Kurtosis		2,865	,733	
Family income	Mean		369,75	16,316	
	95% Confidence Interval for Mean	Lower Bound	336,75		
	Iormean	Upper Bound	402,75		
	5% Trimmed Mean		365,00		
	Median		370,00	/	Side note: If you ask a statistician
	Variance		10648,654		-
	Std. Deviation		103,192		who calculates skewness by hand,
	Minimum		200		s/he will say the cutoff is +/- 3. most
	Maximum		700		software standardizes both
	Range		500	\vee	
	Interquartile Range		148		measures to +/- 1.
	Skewness		,624	,374	
	Kurtosis		1,359	,733	

Statistical Tests

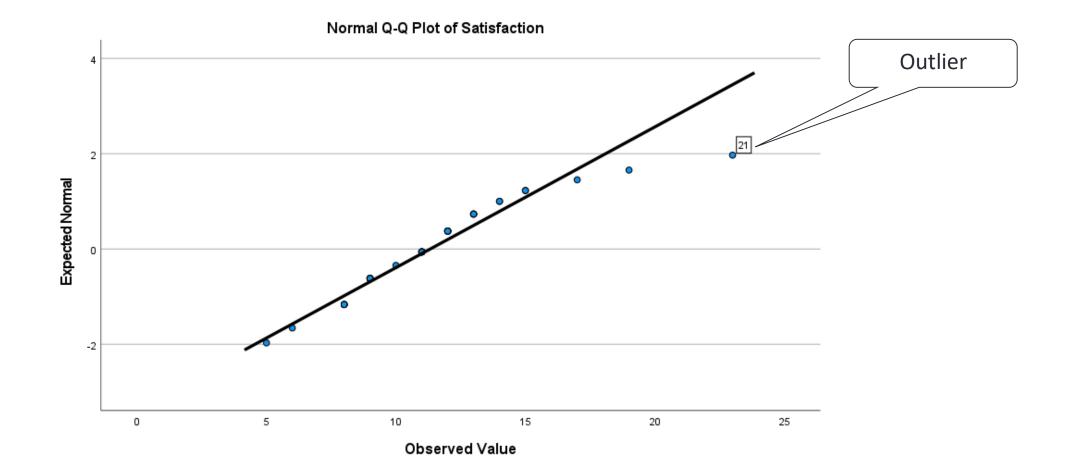


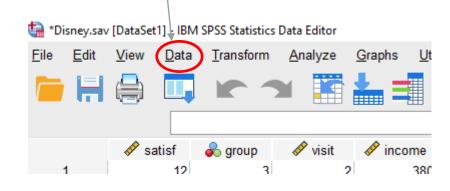
We look here because n < 50.

Boxplot



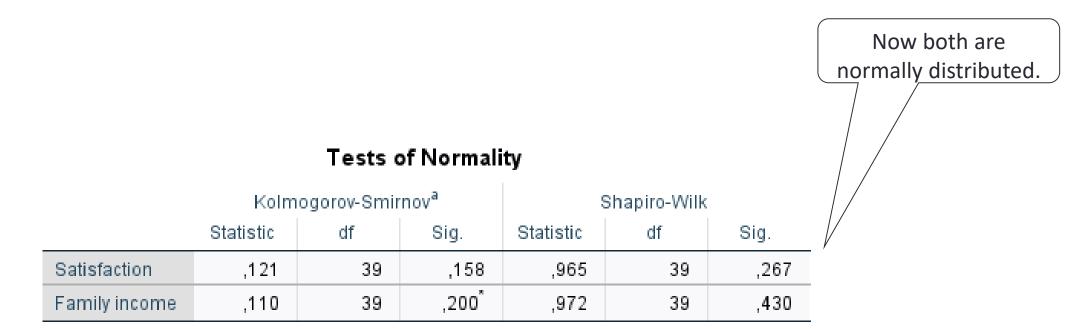
Normal Q-Q Plot





	Select
 Satisfaction [satisf] Group size [group] Earlier visits [visit] Family income [in 	Select O <u>A</u> ll cases O If <u>c</u> ondition is satisfied <u>[f</u> O Ran <u>d</u> om sample of cases <u>Sample</u> O <u>B</u> ased on time or case range <u>Range</u> O <u>U</u> se filter variable: (v) Satisf < 23 (FILTER) [filter_\$]
	Output <u>F</u>ilter out unselected cases Copy selected cases to a new dataset Dataset name: O Delete unselected cases
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Without the Outlier (21)



*. This is a lower bound of the true significance.

a. Lilliefors Significance Correction

Rules of Thumb!

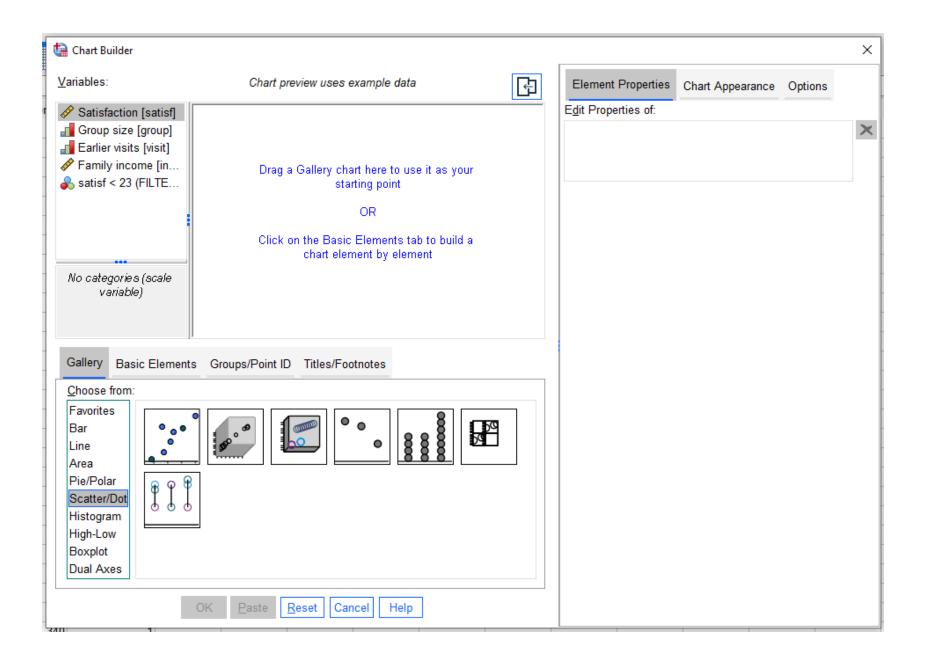
- Consider which technique you are using before worrying too much about normality.
- Remove outliers.
- Consider transforming the data.
- With larger sample sizes normality is less of an issue.

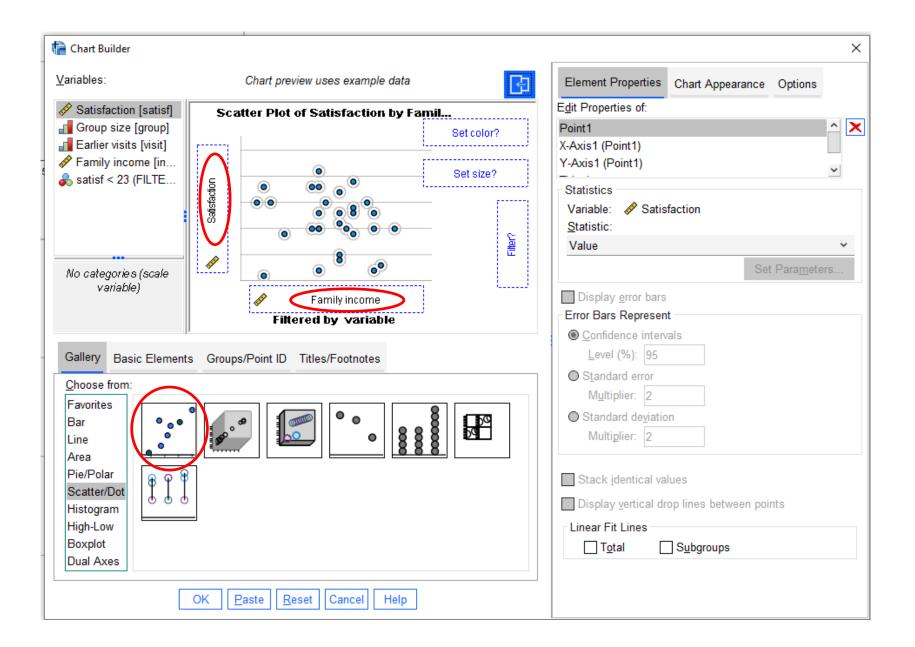
2. Relationships Between Variables

In, for example, regression we want strong linear relationships between the independent variables and the dependent variable, but weak linear relationships between independent variables.

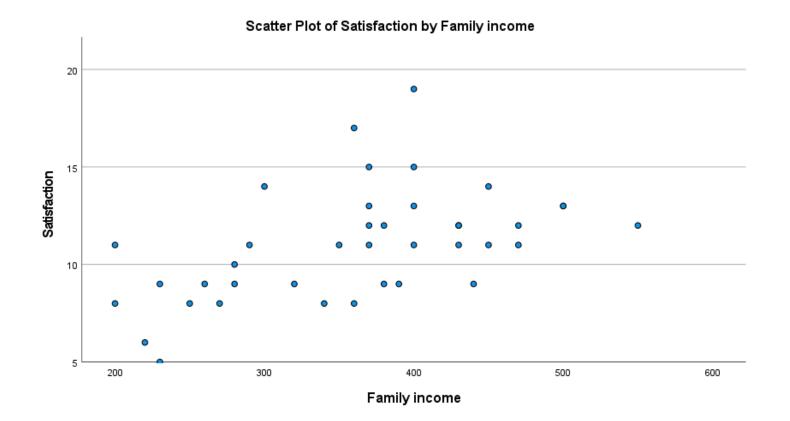
- Examine scatterplots.
- Examine bivariate correlations.

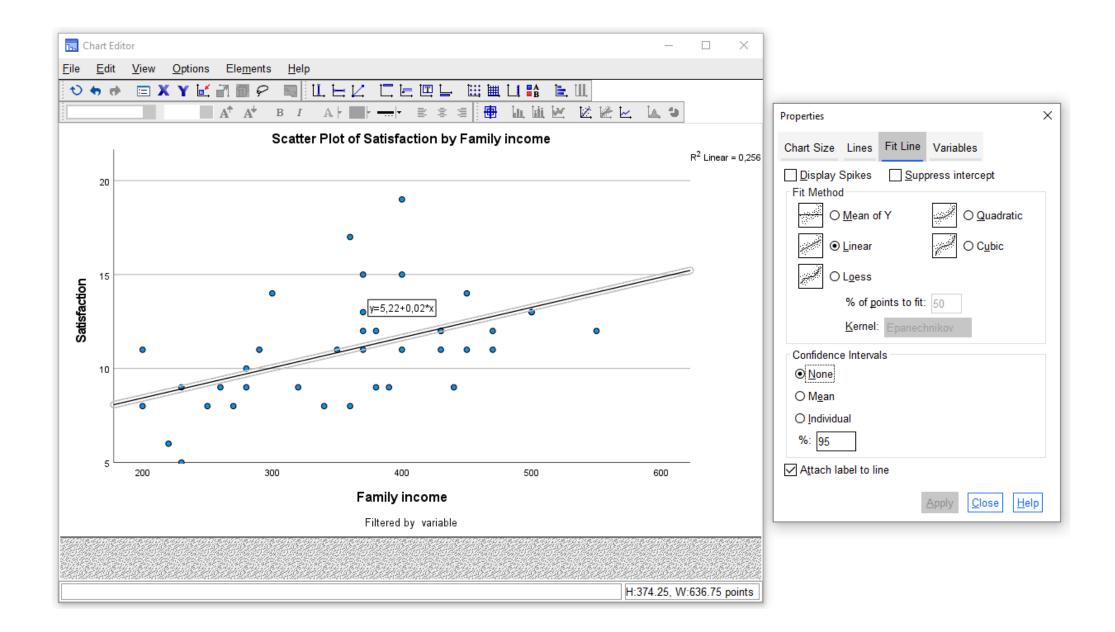
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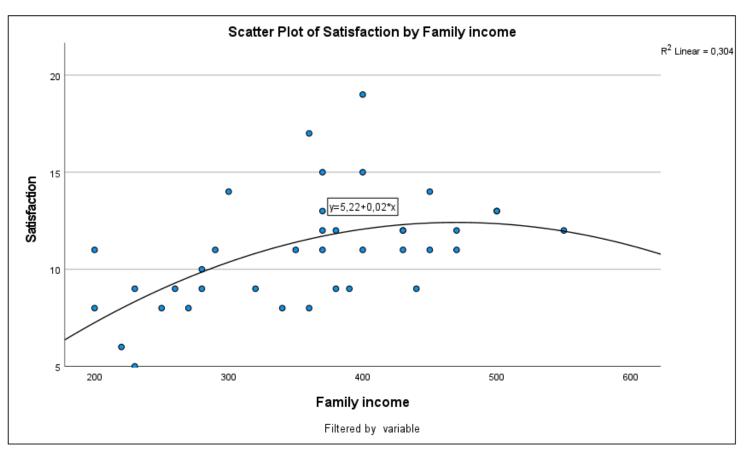


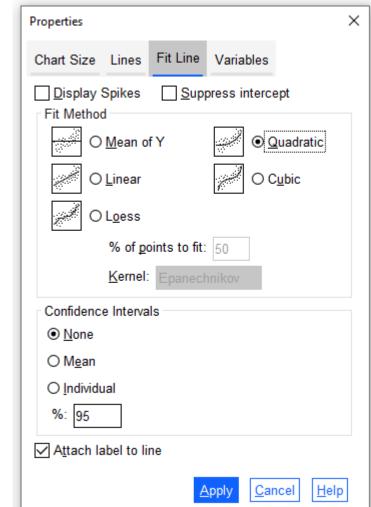


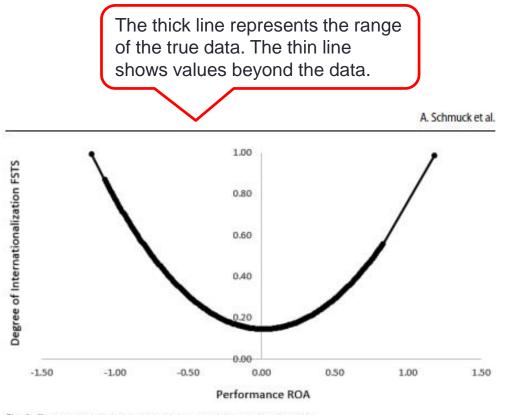
Scatterplot



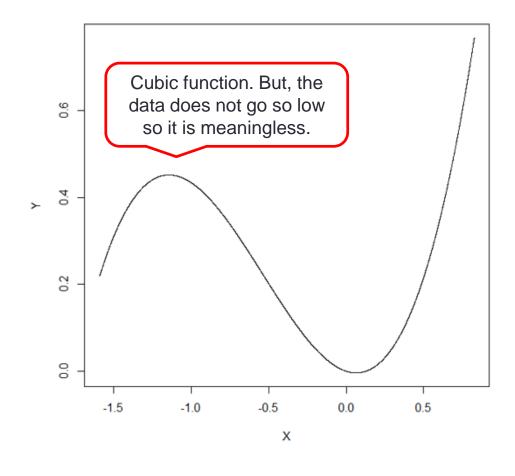












	ta Bivariate Correlations	×
Parametric: When people just say, "correlation", they mean Pearson correlation	✓ Satisfaction [satisf] Group size [group] Earlier visits [visit] ✓ Family income [income] Satisf < 23 (FILTER) [filter_\$]	

Correlations

Family Satisfaction Group size Earlier visits income ,506 -,573** Satisfaction Pearson Correlation 1 ,282 ,001 Sig. (2-tailed) ,081 <.001 N 39 39 39 39 ,764 Pearson Correlation Group size ,282 -,132 1 Sig. (2-tailed) ,081 ,423 <,001 N 39 39 39 39 -,573 -,345 Pearson Correlation Earlier visits -,132 1 Sig. (2-tailed) ,031 <,001 ,423 N 39 39 39 39 ,506 ,764 -,345 Family income Pearson Correlation 1 Sig. (2-tailed) ,001 <,001 ,031 N 39 39 39 39

Correlations

**. Correlation is significant at the 0.01 level (2-tailed).

*. Correlation is significant at the 0.05 level (2-tailed).

Correlations – when considering regression

		Y T		# of earlier	Family
		Satisfaction	Group size	visits	income
Satisfaction	Pearson Correlation				
	Sig. (2-tailed)				
	Ν	39			
Group size	Pearson Correlation	.282			
X	Sig. (2-tailed)	.081			
λ_1	Ν	39			
# of earlier visits	Pearson Correlation	573 **	132		
Xa	Sig. (2-tailed)	.000	.423		
- 2	Ν	39	39		
Family income	Pearson Correlation	.506 **	.764 **	345	
X	Sig. (2-tailed)	.001	.000	.031	
~3	Ν	39	39	39	
War	nt high correlations		4		
١	Want low correlatior	ns (under I. 9	ÐI)		

Correlations

Rule of Thumb!

- Correlations between independent variables should not exceed 0.9. If they do you most likely have problems with multicolliniarity.
- Consider the substantive meaning of correlations. Small correlations, even if they are significant, are meaningless. By small I mean below I.3 I.
- Cohen (1988) established practical guidelines for interpreting statistically significant correlations.

Correlation strength	Absolute value of correlation coefficient
Small	0.10-0.29
Medium	0.30-0.49
Large	0.50-1.00

Table 8.14 Correlation strength

3. Outliers

• There is no rule of thumb. You simply have to look at them and decide if they are an aberration or if they are a natural part of the data.

• Examine how they influence your results.

Article

Best-Practice Recommendations for Defining, Identifying, and Handling Outliers Organizational Research Methods 16(2) 270-301 © The Author(s) 2013 Reprints and permission: sagepub.com/journalsPermissions.nav DOI: 10.1177/1094428112470848 orm.sagepub.com

SAGE

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4. Missing Data

Two questions:

- 1. Why is it missing?
 - You need to determine if the missing data is missing at random, or if there is some sort of pattern as to why it is missing.
- 2. How much is missing?
 - You have to consider how much data is missing. If not so much, just run your analyses with the default "listwise deletion".
 - Pallant often suggests pairwise deletion this way you don't lose so much data.
 - If a lot is missing, you may opt to replace missing values.
 - You may drop variables with a lot of missing values.

Four Steps

- 1. Type of missing data: ignorable not ignorable?
 - Ignorable if substantively missing at random.
- 2. Extent of missing data: If not ignorable, is there so much missing as to bias results?
 - 10% cutoff for no problem.
- 3. Diagnose randomness (many months & large brain).
- 4. What to do?
 - Listwise deletion (complete case approach).
 - Pairwise deletion.
 - Imputation.

Rule of Thumb!

In SPSS you can test for differences between missing and non-missing data across all variables.

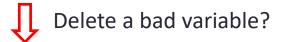
If there are significant patterns – hmm.

- Try to avoid missing data.
- If the % of missing values is low, ignore the patterns.
 - Around 10%.
- Are they variables you really need?
- Are the missing values concentrated in a few cases? Consider deleting them.

Use the "Banking" data

talue Analysis Value Analysis		×
	Quantitative Variables:	Patterns Descriptives Estimation Listwise Pairwise En
	Ma <u>x</u> imum Categories: 25 Case La <u>b</u> els:	☐ Regression Variables EM Regression
Use <u>A</u> ll Variables	•	
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talue Analysis: Patt	terns		×
Display			
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Omit patterns with less	s than 1	% of cases	
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Cases with <u>m</u> issing val	lues, sorte	d by missing value patterns	5
So <u>r</u> t variables by mi			
All cases, optionally so	orted by se	elected variable	
Variables			
Missing Patterns for:		Additional Information for:	
education			
gender	-		
satisfac trust			
commit		Sort by:	
loyalty	-	_ ,	
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Missing data

			D	ata Patte	rns (all c	ases)				
		5	Missing and Extreme Value Patterns							
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1	0	0,								
2	0	0,								
3	0	0,								
4	0	0,								
5	0	,0								
								-		
226	1	14,3						Α		
227	0	0,								
228	3	42,9				Α	Α	Α		
229	0	,0								
230	0	,0								
231	1	14,3				Α				
232	0	,0								
233	0	,0								
234	5	71,4			A	A	А	A	А	
235	1	14,3						Α		

Delete a bad case?