



Authorized Software Value Plus Business Analytics

Business Analytics Award Winner 2012

Maximise Your Investment In SPSS

15th May 2015 – Royal Exchange, London

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Agenda

- 0900 Welcome and Introduction To Smart Vision Europe
- 0915 What's available in each SPSS module? How could each be of value in your own organisation?

What's new in SPSS v23.0 – an overview of the additional functionality included within the latest release of SPSS.

- 1015 Break for Tea & Coffee
- 1030 Getting more from SPSS best practice, effective working and avoiding common pitfalls

Automating and Extending capabilities within SPSS

Training for SPSS – access to specialist support and training options available to both new and experienced users

1115 Summary, Q&A and Close





Predictive Analytics for Smarter Business



- Premium, accredited partner to IBM specialising in the SPSS Advanced Analytics suite.
- Team each has 15 to 20 years of experience working in the analysis, statistics & predictive analytics sector - specifically as senior members of the heritage SPSS team







Maximise your investment in SPSS Statistics

Jarlath Quinn – Analytics Consultant

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An Overview of IBM SPSS Statistics: Add-on Modules

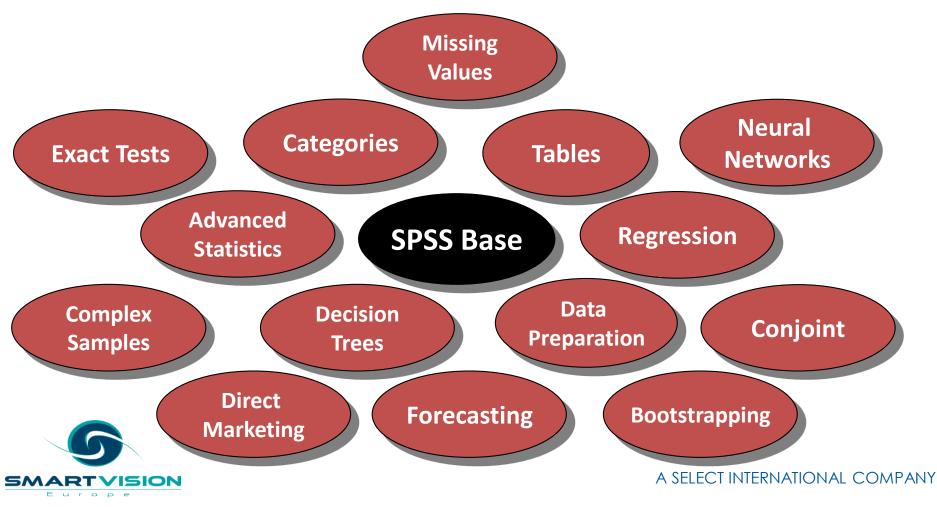
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SPSS Statistics

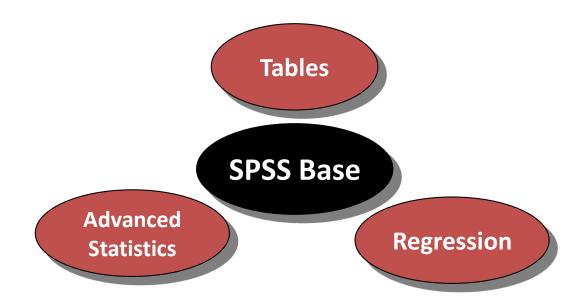
- Since 1968 one of the world's most popular data analysis and statistical interrogation platforms
- Used for everything from epidemiology studies , survey research and business reporting to direct marketing, credit risk, predictive modelling and asset management
- Statistics included in the base software:
 - Descriptive statistics: Cross tabulation, Frequencies, Descriptives statistics
 - Statistical Tests: T-test, ANOVA, Correlation
 - Prediction for numerical outcomes: Linear regression
 - Prediction for identifying groups: Factor analysis, Cluster analysis



IBM SPSS Statistics Base & Associated Modules

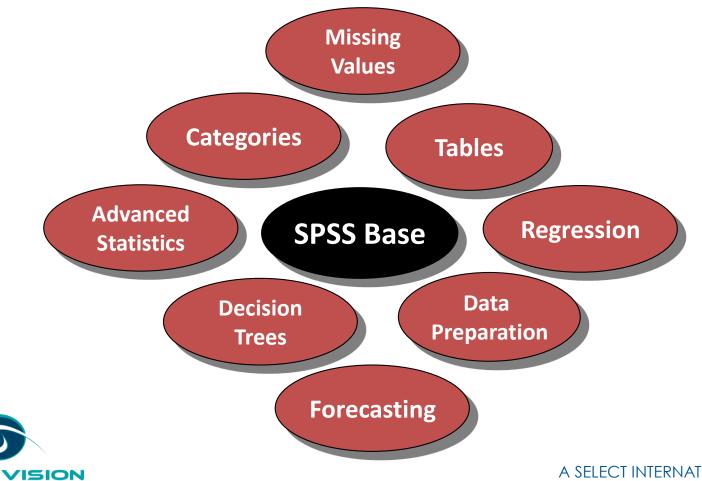


IBM SPSS Statistics Standard



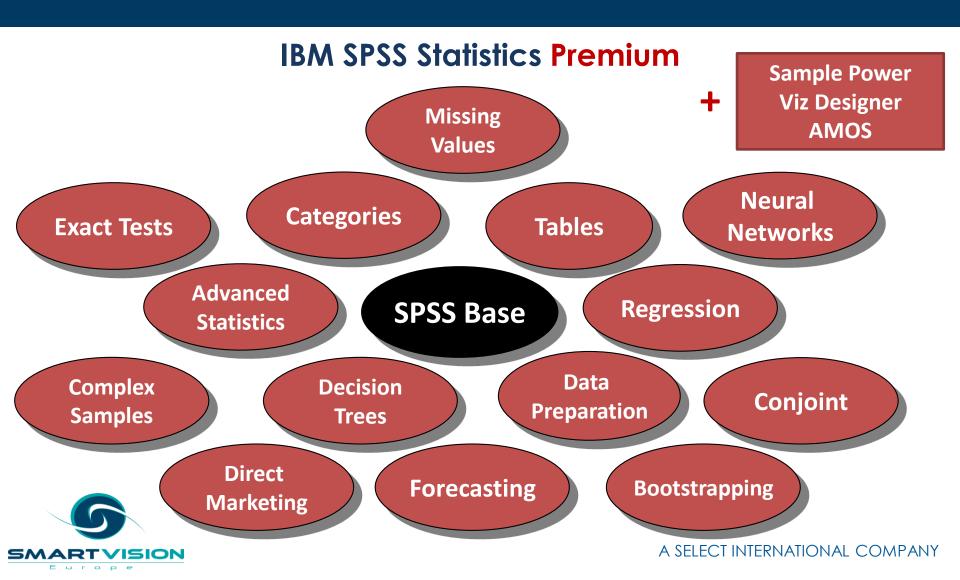


IBM SPSS Statistics Professional



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IBM SPSS Custom Tables

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IBM SPSS Custom Tables

Custom Tables								
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A single dialog box for all tabulation and reports
Preview the table as you drag and drop the fields
into the rows and columns





IBM SPSS Custom Tables

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SPSS Regression Models

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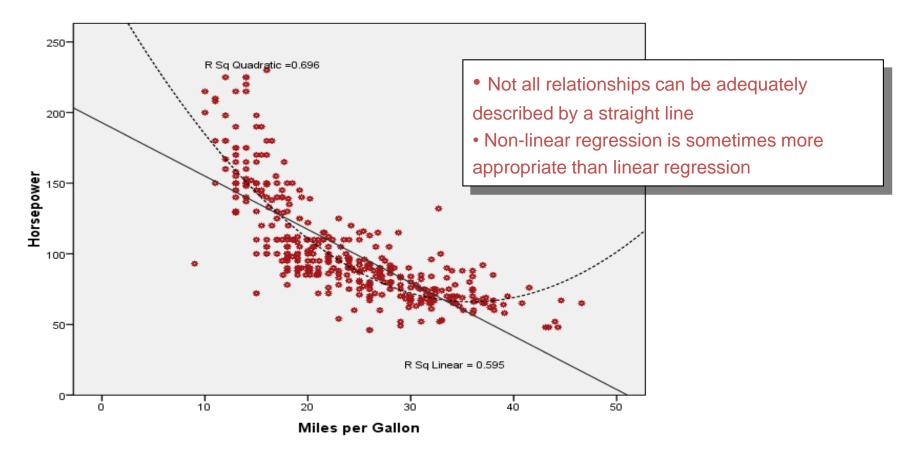
SPSS Regression Models

- The SPSS *Regression Models* module contains a wide range of nonlinear regression models that augment the linear regression functionality in SPSS Base.
- *Regression Models* is a family of classical predictive techniques all of which involve fitting (or *regressing*) a line or curve to a series of observations in order to model effects or predict outcomes.
- *Regression Models* is often used in situations where the Linear Regression functionality in SPSS base is either inappropriate or is too simplistic
- Logistic Regression is a very widely-used technique for predicting categorical outcomes. In Regression Models there are two forms of Logistic regression:
 - Binary Logistic for predicting 2 category outcomes
 - Multinomial Logistic for predicting more than 2 category outcomes
- Regression Models also contains:
 - Nonlinear regression and Constrained Nonlinear Regression



Probit, Weighted Least Squares and Two Stage Least Squares

SPSS Regression Models







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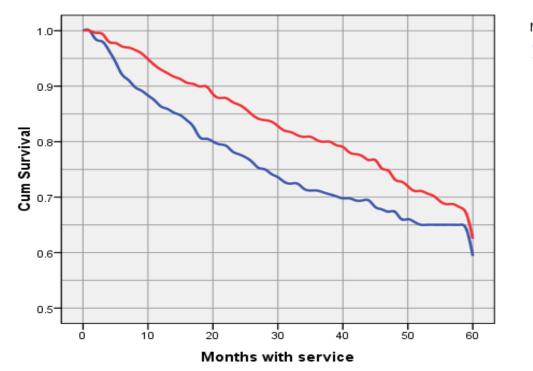
- Advanced Models is the most 'analytically rich' SPSS module. Advanced Models includes a very wide range of multivariate procedures for investigating complex relationships in data.
- A number of the procedures in *Advanced Models* are relatively technical in a statistical sense. In particular, *Advanced Models* encompasses General and *Generalized* Linear Modelling capabilities.
- General Linear Models allow you to model relationships and interactions between many factors. The general linear model incorporates a number of different statistical models: ANOVA, MANOVA, ANCOVA, Repeated Measures etc.
- Generalized Linear Models are an extension of General Linear Models in that they are able to work with a greater range of data distributions. In particular, the model allows for the dependent variable to have a non-normal distribution.
- The Generalized Estimating Equations (GEE) procedure extends the generalized linear model to allow for analysis of repeated measurements or other correlated observations, such as clustered data.



- Advanced Models also includes Linear Mixed Models. If you work with data that display correlation and non-constant variability, such as nested data that represent students within faculties or employees within departments, you can use the linear mixed models procedure to model means, variances, and covariances in your data.
- Advanced Models includes General Loglinear and LOGIT Loglinear analysis.
- Advanced Models also includes a number Survival Analysis algorithms. In recent times, Survival Analysis has also been used in application such as insurance claims and customer churn.
- Advanced Models offers 4 distinct Survival Analysis procedures:
 - Life Tables
 - Kaplan-Meier
 - Cox Regression
 - Cox Regression with time-dependent covariate



Survival Function





• Unmarried customers churn sooner





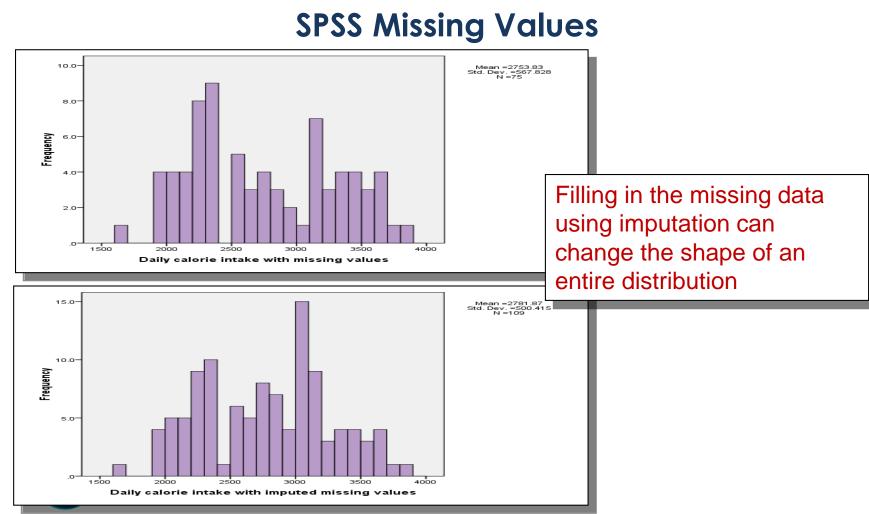
SPSS Missing Values

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SPSS Missing Values

- The *Missing Values* module procedure provides three main functions:
 - It describes the pattern of missing data. Where are the missing values located? How extensive are they? Are values missing randomly?
 - Provides estimates of statistics like means, standard deviations and correlations for data series that contain missing values.
 - Fills in (imputes) missing data with estimated values using special methods like regression or EM (expectation-maximization).
- The *Missing Values* module helps address several concerns caused by incomplete data. By investigating patterns of missing data it can address questions such as 'Why are the data missing?'. The means estimation procedures address questions such as 'How does the missing data affect summary statistics?'





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SPSS Complex Samples

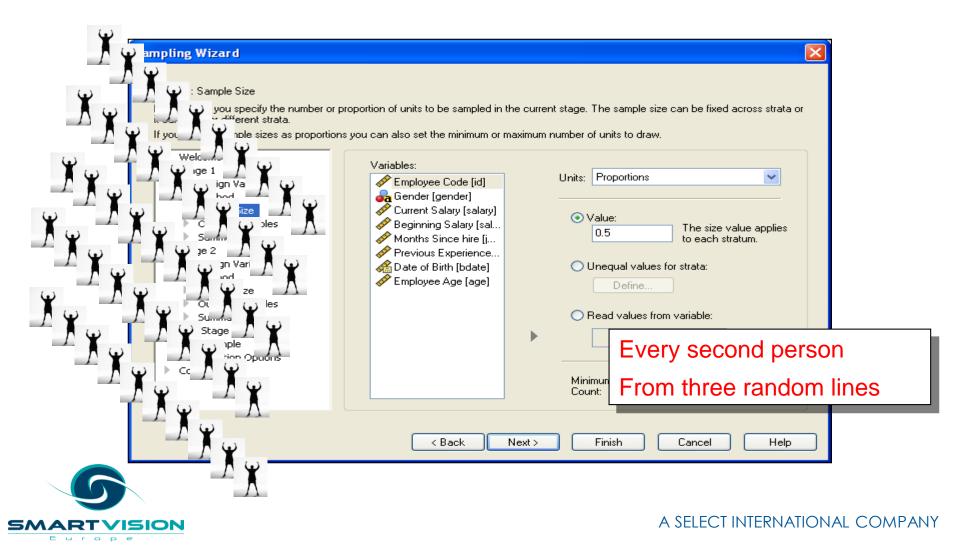
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SPSS Complex Samples

- An inherent assumption of many classical statistical procedures is that the data represents a *simple random sample* drawn from the population of interest.
- The SPSS *Complex Samples* module allows users to draw samples that are more complicated than simple random schemes.
- *Complex Samples* also allows statistical analyses to be carried out that *take account* of the complex sampling scheme used in collecting the data.
- An example of this would be carrying out a chi square test to see if larger households are more likely to recycle glass than smaller households. Using *Complex Samples*, the researchers could calculate a more appropriate test statistic based on a sample where every 3rd house was sampled from a random selection of 50 streets in a town.



SPSS Complex Samples





SPSS Exact Tests

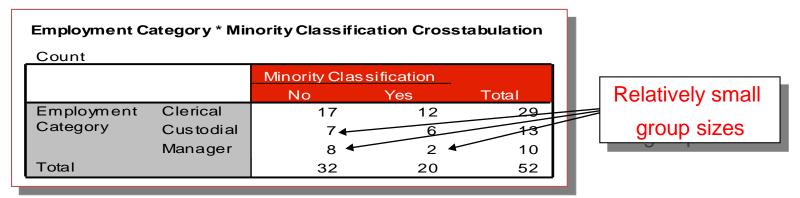
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SPSS Exact Tests

- The SPSS Exact Tests module provides additional methods for calculating the significance levels for the statistical tests available through the Crosstabs and the Nonparametric Tests menus.
- Using the standard tests in SPSS Base (known as asymptotic tests) can lead to misleading or inaccurate results when working with small datasets or sparse groups in the sample data. *Exact Tests* enables users to obtain an accurate significance level without relying on assumptions that might not be met by the data.
- *Exact Tests* offers two extra methods of calculating probabilities on top of the normal asymptotic methods in SPSS Base.
 - Monte Carlo Estimate: An unbiased *estimate* of the *exact* significance level. This method is most useful when the data set is too large to compute exact significance but the data do not meet the assumptions of the asymptotic method.
 - Exact: The probability of the observed outcome or an outcome more extreme is calculated exactly.



SPSS Exact Tests



Chi-Square Tests								
	Value	df	Asymp. Sig. (2-sided)	Exact Sig. (2-sided)	Exact Sig. (1-sided)	Point Probability		
Pearson Chi-Square	1.869 ^a	2	.393	.440				
Likelihood Ratio	2.004	2	.367	.411				
Fisher's Exact Test	1.819			.440				
Linear-by-Linear Association	.937 ^b	1	.333	.374	.217	.092		
N of Valid Cases	52							

a. 1 cells (16.7%) have expected count less than 5. The minimum expected count is 3.85.

b. The standardized statistic is -.968.

Chi Square showing exact probabilities highlighted in red

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SPSS Decision Trees

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SPSS Decision Trees

- Decision trees are used *extensively and widely* within Predictive Analytics
- Decision trees can be used to
 - Build profiles of customers/employees/clients
 - Find key behavioural segments
 - Generate predictive models
- Decision Trees can be expressed as a series of hierarchical rules which means that they can be converted in languages like SQL for database scoring
- Decision Trees are especially popular because
 - they are fairly visual representations of models
 - relatively easy to understand



Understanding Decision Trees – a worked example

• What were the most important factors determining survival during the sinking of the RMS Titanic?

Survival on the RMS Titanic

		Count	Percent %
survive	Did not survive	1490	68%
	Survived	711	32%
	Total	2201	100%



Gender?



Age?



Class?

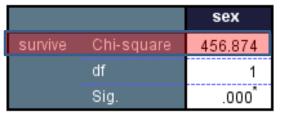


Statistical Tests Like Chi Square help to answer this

		sex					
		f€	emale	male			
		Count	Column Percent %	Count	Column Percent %		
survive	Did not survive	126	26.8%	1364	78.8%		
	Survived	344	73.2%	367	21.2%		
	Total	470	100.0%	1731	100.0%		

Survival on the RMS Titanic

Pearson Chi-Square Tests





Statistical Tests Like Chi Square help to answer this

		age					
		÷	adult	child			
		Count	Column Percent %	Count	Column Percent %		
survive	Did not survive	1438	68.7%	52	47.7%		
	Survived	654	31.3%	57	52.3%		
	Total	2092	100.0%	109	100.0%		

Survival on the RMS Titanic

Pearson Chi-Square Tests

		age
survive	Chi-square	20.956
	df	1
	Sig.	.000



Statistical Tests Like Chi Square help to answer this

			class								
		1st		2nd			3rd	crew			
		Count	Column Percent %								
survive	Did not survive	122	37.5%	167	58.6%	528	74.8%	673	76.0%		
	Survived	203	62.5%	118	41.4%	178	25.2%	212	24.0%		
	Total	325	100.0%	285	100.0%	706	100.0%	885	100.0%		

Survival on the RMS Titanic

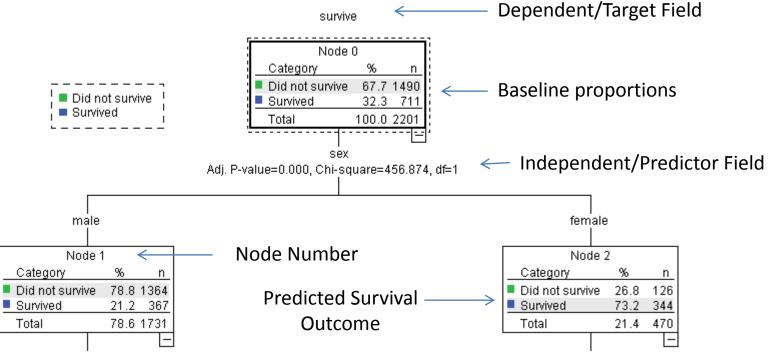
Pearson Chi-Square Tests

		class
survive	Chi-square	190.401
	df	3
	Sig.	.000



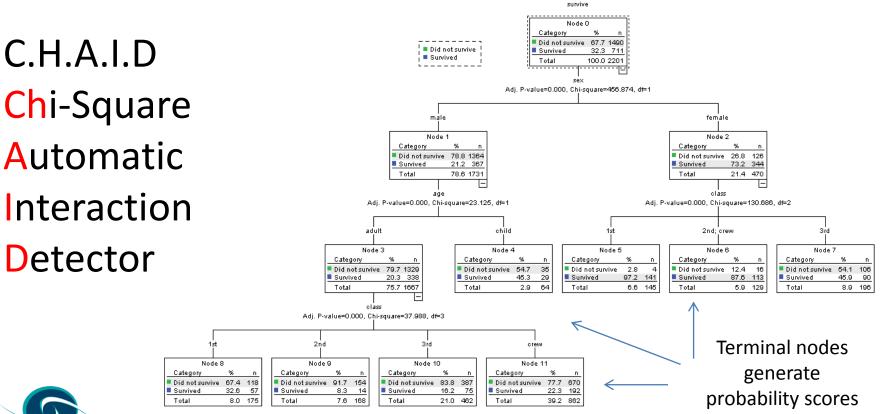
Gender is most important

...and a CHAID Decision tree will reflect this....



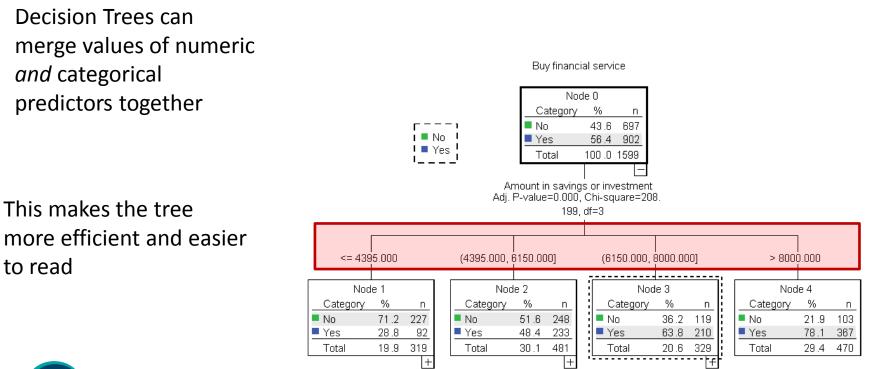


Full CHAID Decision Tree





Merging/Splitting in CHAID Trees

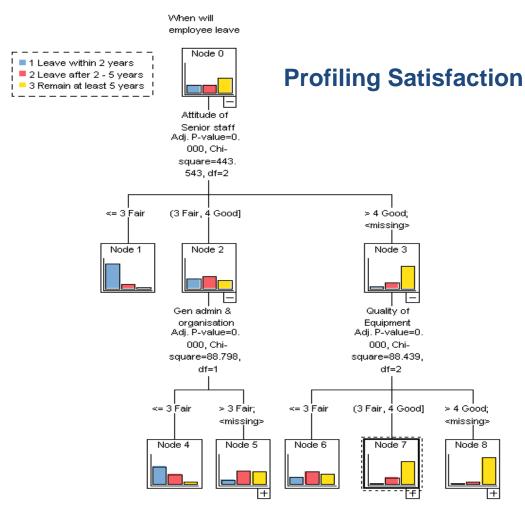




SPSS Classification Trees Purchased Node O Category % n No No 88.7 907 j 🗖 No 🗆 Yes 11.3 116 I 🗕 Yes 🕴 100.0 1023 Total I____ Predicting Employment Category Adj. P-value=0.000, Chi-square=77. **Customer Churn** 213, df=1 Marketing Response Unemployed; Pensioner; Employed; _ Houseperson Student; Part Time Fraud ____ Node 1 Node 2 Category % Category % n n 96.5 No No 78.9 360 No No 547 **Cross Sell** Yes 21.1 96 Yes 3.5 20 Total 44.6 456 Total 55.4 567 Asset Failure Design Adj. P-value=0.000, Chi-square=25. 679, df=1 Design 2 Design 1 Node 3 Node 4 Category - % Category % n n No No 88.4 206 No No 69.1 154 Yes 11.6 27 Yes 30.9 69 21.8 223 Total 22.8 233 Total -I-I Age Group Age Group Adj. P-value=0.001, Chi-square=13. Adj. P-value=0.029, Chi-square=7. 181. df=1 711. df=1 <= 36 to 45 <= 46 to 55 > 46 to 55 > 36 to 45 Node 5 Node 6 Node 7 Node 8 - %6 Category % Category % Category % NΥ Category n n n n RT VISION No. 94.9 94 No. 83.6 112 No 62.5 105 No. 89.1 49 Europe 5.1 5 16.4 22 37.5 63 10.9 6 Yes Yes Yes Yes

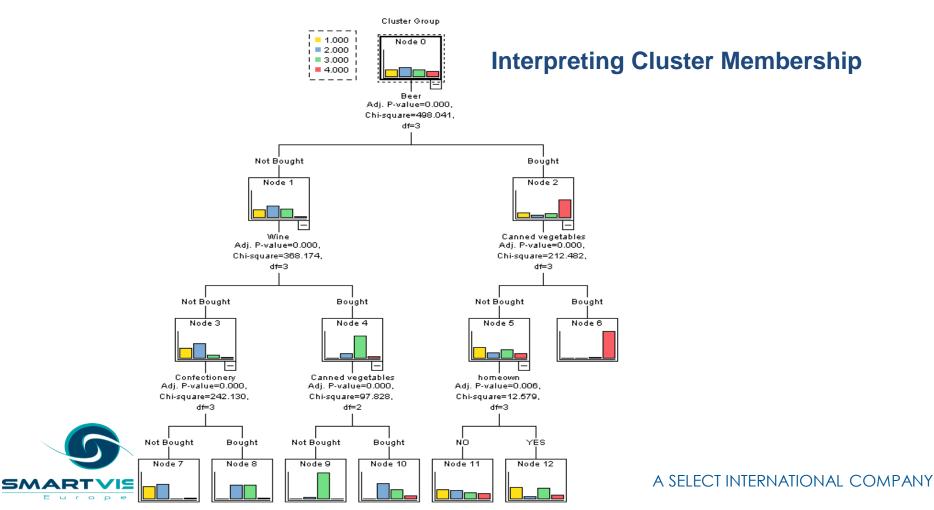
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SPSS Classification Trees



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SPSS Classification Trees





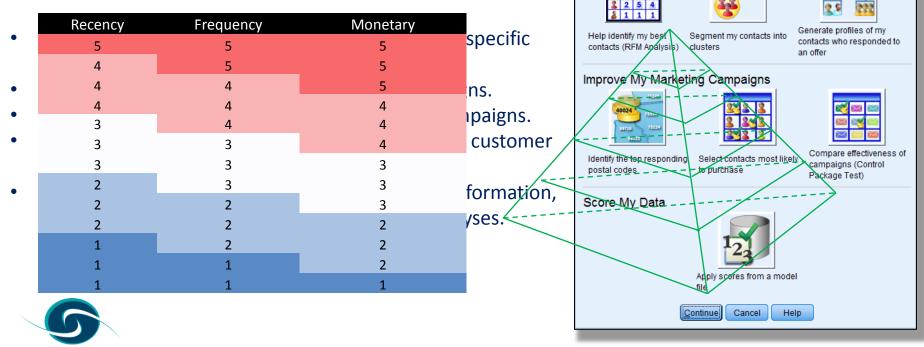
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SPSS Direct Marketing

RFM – Recency, Frequency, Monetary

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Choose one of the following techniques:

Understand My Contacts



SPSS Categories

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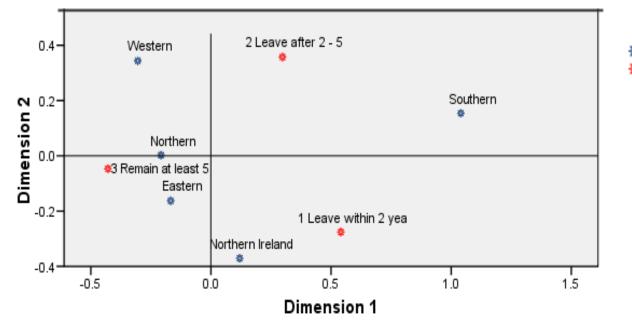
SPSS Categories

- The *SPSS Categories* module provides a number of algorithms based on a family of techniques called *optimal scaling*.
- Optimal scaling attempts to *quantify the* category groups of categorical fields i.e. assign numerical values to the categories *as if they existed on a scale*.
- By quantifying categories can be used as excellent exploratory tools when modelling multivariate categorical data.
- Examples of techniques include:
 - Correspondence Analysis
 - Categorical Regression
 - Categorical PCA



SPSS Categories

• Quantifying the two categorical fields '*Internal Sector*' and '*When will employee leave*' helps us to explore the relationship between the two variables as if they were continuous fields in a scatterplot



🗱 Internal Sector 🔆 When will employee leave



SPSS Conjoint

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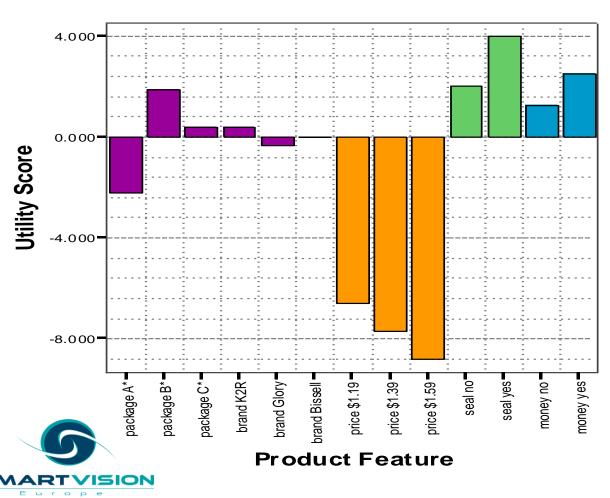
SPSS Conjoint

- Conjoint analysis is a technique pioneered by market research analysts to determine how people value the different features that make up an individual product or service.
- Conjoint analysis can be used to discover the optimal combination of product/service attributes in terms of the combination that is most influential on customer choice or decision making.
- *Conjoint* works by showing respondents a particular set of products (or services) and by analysing how they make preferences between these products.
- By mapping the different features or aspects of the products to the choices that the respondent makes, the *Conjoint* technique is able to infer the ideal set of

characteristics for a product or service.

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SPSS Conjoint



Results of a Conjoint analysis showing utility (preference) scores for different aspects of a cleaning product

Note: Conjoint analysis in SPSS is primarily run via SPSS syntax



SPSS Forecasting

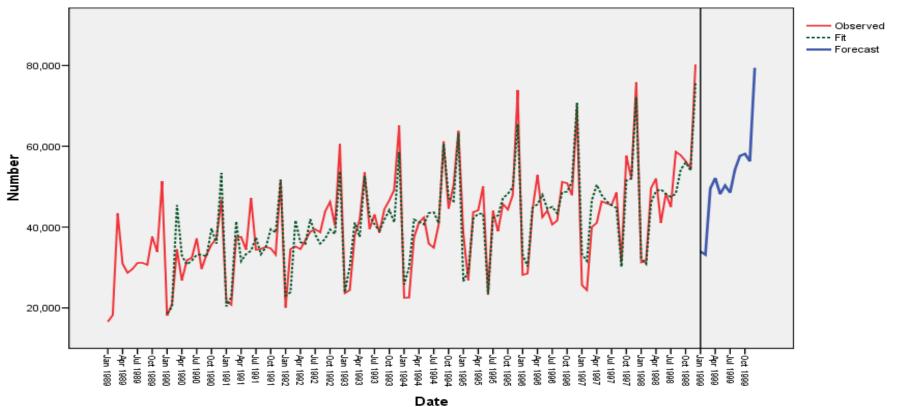
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SPSS Forecasting

- *SPSS Forecasting* is the SPSS *time series* module. Time series forecasting is the use of a model to predict future events based on known past events.
- Examples of time series forecasting include:
 - Predicting the number of staff required on each day for a call centre
 - Forecasting the number of patients visiting the accident and emergency department
 - Predicting demand for a gas or electricity supplier
 - Estimating passenger numbers for a train company
- The time factor, is in itself, a predictor of the dependent variable. In other words, in time series, the *past provides a model for the future*.
- SPSS Forecasting is particularly powerful as it can automatically select and fit a Time Series Model



SPSS Forecasting







SPSS Data Preparation

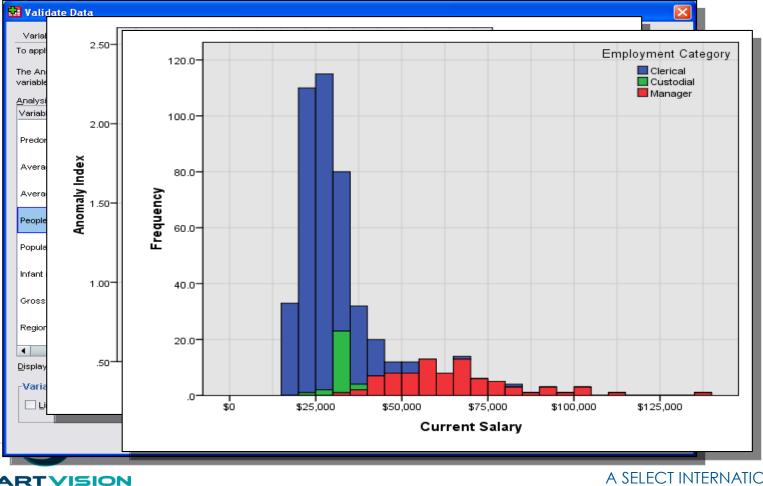
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SPSS Data Preparation

- The SPSS Data Preparation module allows users to identify data errors or unusual cases in their datasets. Using a combination of basic checks, validation rules or anomaly detection algorithms, the Data Preparation module will generate new variables or output reports that identify problematic cases or unusual records.
- It can be used to:
 - Identify records with a high percentage of missing values, a high degree of variability or conversely, too little variability as well as incomplete id fields or duplicate records.
 - Provide a graphical overview of each of the fields and the capability to create validation rules for individual fields. An example of this would be a rule that ensures a field is an integer (i.e. no decimal places) such as age.
 - Create rules that ensure that the values in combinations of variables do not contradict each other or imply errors in the data. An example would be a cross-variable rule that ensures that all car

drivers are at least 17 years old..

SPSS Data Preparation



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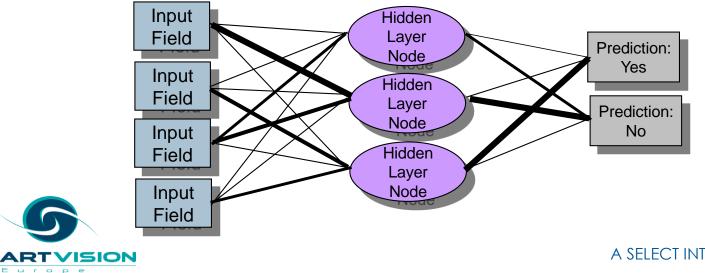


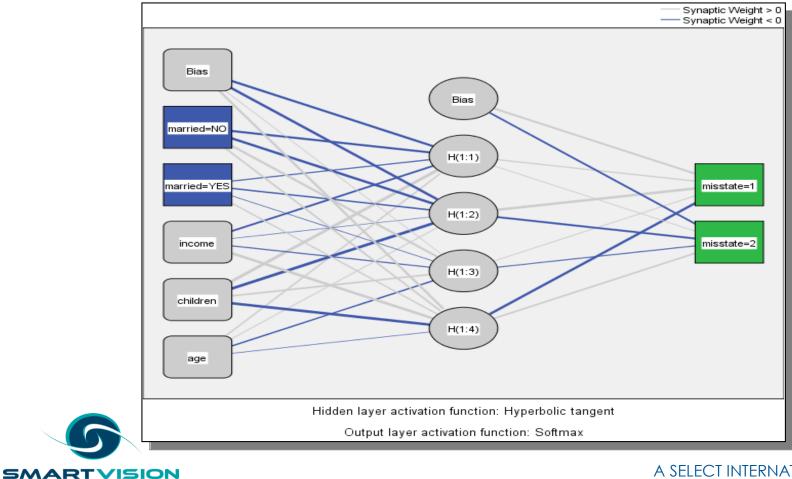
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- SPSS *Neural Networks* provides an alternative predictive capability to approaches such as regression or classification trees. Predictive neural networks are particularly useful in applications where the data from the underlying phenomena is complex such as fraud detection, credit scoring and pattern recognition.
- Neural Networks attempt to 'learn' the outcomes of a target field by constantly updating the model with increasingly smaller changes until model accuracy can no longer be improved
- One of the primary advantages of neural networks when compared to classical statistical techniques, is their flexibility and lack of distributional assumptions.



- A neural network works by taking the values of predictor or input fields and feeding them into the algorithm as an *input layer*.
- The *input layer* is used to create a *hidden layer* containing unseen nodes (or units) where each node is some function of the input fields (in fact some networks can create more than one hidden layer).
- The *output layer* contains the responses or predictions. The network is continually rebuilt or refined so that the *synaptic weights* in the nodes correctly predict the outcome.





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What's new in IBM SPSS Statistics v23?

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- Loads of enhancements enabled via the Python Essentials Pack (available at installation or via a separate download)
- Examples include
 - Manage Datasets
 - Read Triple S Data
 - Connect to Internet Data
 - Weibull Plots
 - Anonymize Variables
 - Simulate Active Dataset

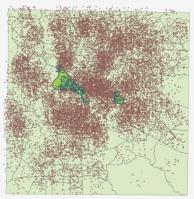


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	🔏 Set Measurement Level for Unknown	
	E Create Value Labels from Data	
	Copy Data Properties	
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	🔝 Identify Unusual Cases	
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	Propensity Score Matching	
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Geospatial Association Rules

- Using geospatial association rules, you can find patterns in data based on both the spatial and non-spatial properties. For example, you might identify patterns in crime data by location and demographic attributes. From these patterns, you can build rules that predict where certain types of crimes are likely to occur.
- This procedure is available in the *Base Statistics* option.







IBM SPSS Statistics v23: Geospatial Association Rules

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-	rt: Confidence 🔹 🔍 🔍	
Rules		Rule 5 : CrimeType = MinorTheft (Predicted By Within PropertyValue ≤ 120,310.000)
1 Rule Popi	e 1 - CrimeType = MinorTheft (Predicted By Within ulation > 8,575.500 & Within PropertyValue ≤ 120,310.0	Events satisfying the rule (both condition and prediction) Context objects that meet the conditions of the rule Events satisfying the prediction but not the condition Context objects that do not meet the conditions of the rule
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3 Rule Hous	e 3 - CrimeType = MinorTheft (Predicted By Within scholds $\leq 2,722.500$ & Within PropertyValue $\leq 120,310$	
4 Rule Popi	e 4 - CrimeType = MinorTheft (Predicted By Within ulation \leq 8,575.500 & Within Households \leq 2,722.500)	
and c And	e 5 - CrimeType = MinorTheft (Predicted By Within pertyValue ≤ 120,310.000)	
6 Rule Popi	e 6 - CrimeType = MinorTheft (Predicted By Within ulation ≤ 8,575.500 & Within PropertyValue ≤ 120,310	
/ Prec	e 7 - CrimeType = MajorTheft (dicted By	
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IBM SPSS Statistics v23: Geospatial Association Rules

🖏 Interactive Output - Rules Map File Edit View Help	
Sort: Confidence	
Rules	Rule 8 : CrimeType = MajorTheft (Predicted By Within Population > 8,575.500 and Within PropertyValue > 120,310.000)
Rule 1 - CrimeType = MinorTheft (Predicted By Within Population > 8,575.500 & Within PropertyValue \leq 120,310.0	Events satisfying the rule (both condition and prediction) Context objects that meet the conditions of the rule Events satisfying the prediction but not the condition Context objects that do not meet the conditions of the rule
2 Rule 2 - CrimeType = MinorTheft (Predicted By Within Population > 8,575.500 & Within Households \leq 2,722.500)	
3 Rule 3 - CrimeType = MinorTheft (Predicted By Within Households \leq 2,722.500 & Within PropertyValue \leq 120,310	
4 Rule 4 - CrimeType = MinorTheft (Predicted By Within Population ≤ 8,575.500 & Within Households ≤ 2,722.500)	
5 Rule 5 - CrimeType = MinorTheft (Predicted By Within PropertyValue ≤ 120,310.000)	
6 Rule 6 - CrimeType = MinorTheft (Predicted By Within Population ≤ 8,575.500 & Within PropertyValue ≤ 120,310	
7 Rule 7 - CrimeType = MajorTheft (Predicted By	
8 Rule 8 - CrimeType = MajorTheft (Predicted By	
9 Rule 9 - CrimeType = MajorTheft (Predicted By	
10 Rule 10 - CrimeType = MajorTheft (Predicted By	
Confidence	

IBM SPSS Statistics v23: Geospatial Association Rules

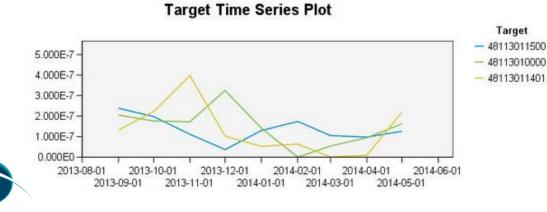
File <u>E</u> dit <u>V</u> iew <u>H</u> elp	
Sort: Rule Support 🔹 🍳 🔍	
Rules	Rule 16 : CrimeType = Violent (Predicted By Within Property\/alue ≤ 120,310.000) Events satisfying the rule (both condition and prediction) Context objects that meet the conditions of the rule Events satisfying the prediction but not the condition Context objects that do not meet the conditions of the rule
2 Rule 3 - CrimeType = MinorThett (Predicted By Within Households ≤ 2,722.500 & Within PropertyValue ≤ 120,3 3 Rule 6 - CrimeType = MinorThett (Predicted By Within PropertyValue ≤ 120,3 4 Rule 4 - CrimeType = MinorThett (Predicted By Within PropertyValue ≤ 120, 4 Rule 4 - CrimeType = MinorThett (Predicted By Within Propulation ≤ 8,575.500 & Within Household) 5 Rule 16 - CrimeType = MinorThett (Predicted By Within Propulation ≤ 8,575.500 & Within Household) 6 Rule 16 - CrimeType = NinorThett (Predicted By Within Propulation ≤ 8,575.500 & Within Household) 7 Rule 17 - CrimeType = NinorThett (Predicted By 7 Rule 13 - CrimeType = NinorThett (Predicted By 8 Rule 13 - CrimeType = NinorThett (
9 = MajorTheft (10 Rule 19 - CrimeType = Violent (Rule Support	

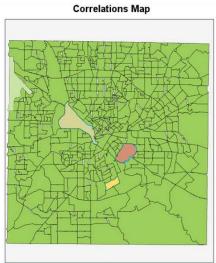


• Spatial Temporal Prediction

ч г о р

- Spatial temporal prediction uses data that contains location data, input fields for prediction (predictors), a time field, and a target field. Each location has numerous rows in the data that represents the values of each predictor at each time interval at each location.
- This procedure is available in the *Base Statistics* option.





• Temporal Causal Models

- Temporal causal modelling attempts to discover key causal relationships in time series data. In temporal causal modelling, you specify a set of target series and a set of candidate inputs to those targets.
- The procedure then builds an autoregressive time series model for each target and *includes only those inputs that have a causal relationship with the target*. This approach differs from traditional time series modelling where you must explicitly specify the predictors for a target series.
- Temporal causal modelling procedures are available in the *Forecasting* option.



• Bulk Loading to a database

• When you export data to a database, bulk loading submits data to the database in batches instead of one record at a time. This action can make the operation much faster, particularly for large data files.

- Bulkloading				
🗸 B <u>u</u> lk Load				
Batch size:	10000			
Batch commit				
© 0 <u>D</u> BC				
◎ <u>R</u> ow-wise binding				
○ Colu <u>m</u> n-wise binding				







Getting more from SPSS - Automating and Extending

John McConnell – Services

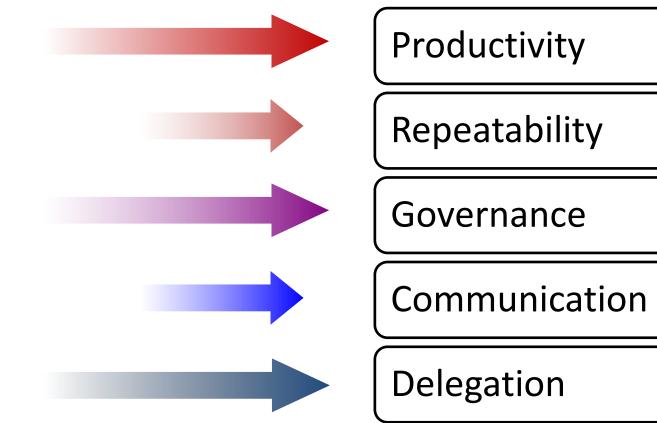
www.sv-europe.com

Contents

- Background
- Levels of automation with syntax
- Automating beyond syntax
- Extensions
- Automating SPSS from the outside
- Support from Smart Vision



Some reasons to automate



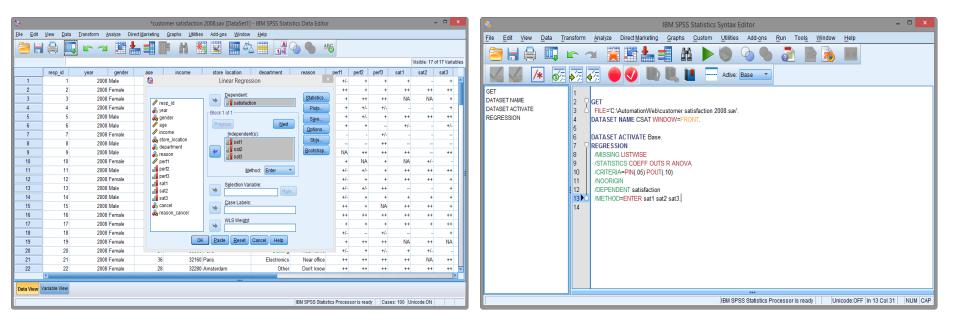


Contents

- Background
- Levels of automation with syntax and streams
- Automating beyond syntax and streams
- Automating SPSS from the outside



Automation – Level 1



From the GUI to ...





Defining and pasting

¶≣	Open Data	×	
Look in: 退 Automatio	nWeb 🔽 🙆 🔯 🔝 🖿		
Census.sav	on 2008.say		
File name: custome	r satisfaction 2008.sav	Open	GET
Files of type: SPSS St	atistics (*.sav)	<u>P</u> aste	FILE='C:\AutomationWeb\customer satisfaction 2008.sav.
Encoding:	Ŧ	Cancel	DATASET NAME CSAT WINDOW=FRONT.
Minimize string wid	ths based on observed values	<u>H</u> elp	
			DATASET ACTIVATE CSAT.
	Retrieve File From Repository		REGRESSION
			/MISSING LISTWISE
			STATISTICS COEFF OUTS R ANOVA
			/CRITERIA=PIN(.05) POUT(.10)
			/NOORIGIN

/DEPENDENT satisfaction /METHOD=ENTER sat1 sat2 sat3.



Getting help

ALL	-
ANOVA	
BCOV	
CHANGE	
CI	
COEFF	
COLLIN	
DEFAULTS	
F	
LABEL	
OUTS	Ŧ

Auto or <ctrl>+<space> Pops up relevant options



Tool to show us the syntax options for the selected command

<u>H</u> elp	
O Topics	
👔 <u>T</u> utorial	
Case Stu <u>d</u> ies	
Working with R	
Statisti <u>c</u> s Coach	
Command Syntax Reference	e
SPSS Commu <u>n</u> ity	
About	
Algorithms	
IBM SPSS Products Home	
Programmability	•

The PDF of all commands and options



Forgot to Paste?

Charts	Pivot Tables	File Locations	Scripts	Multiple Imputations	Syntax Editor
Startup Folder	s for Open and Save Di	alogs			
© <u>S</u> pecified fo	older				
<u>D</u> ata file:	s: C:\Users\jmcco_00	0\Documents			Browse
<u>O</u> ther file	S: C:\Users\jmcco_00	0\Documents			Browse
Last folder	used				
Session Journ	al				
Record syn	itax in Journal				
◎ Append ◎	Overwrite				
Journal file: C	:\temp\statistics.jnl				Bro <u>w</u> se

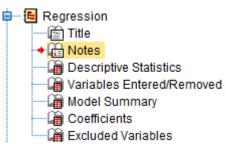
The Journal File is set (in Edit > Options) to record syntax automatically

- until overwritten or deleted



Forgot to Paste?

	Notes	
Output Created		03-DEC-2014 07:12:18
Comments		
Input	Data	C:\AutomationWeb\customer satisfaction 2008.sav
	Active Dataset	Base
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data File	140
Missing Value Handling	Definition of Missing	User-defined missing values are treated as missing.
	Cases Used	Statistics are based on cases with no missing values for any variable used.
Syntax		REGRESSION /DESCRIPTIVES MEAN /MISSING LISTWISE /STATISTICS R COEFF OUTS /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT satisfaction /METHOD= STEPWISE sat1 sat2 sat3.
Resources	Processor Time	00:00:00.02
	Elapsed Time	00:00:00.02
	Memory Required	5088 bytes
	Additional Memory Required for Residual Plots	0 bytes



The (usually hidden) **Notes** table in output contains the syntax for each output



Batch running Syntax – The Production Facility

	Production Facility	×
My Jobs Background Job St	tatus	
Location of production job files	C:AutomationWeb	<u>B</u> rowse
Select a job or create a new on	ne:	
KDA.spj	KDA.spj r Syntax files	
	-	New
	File C:\AutomationWeb\KDA(1).sps	Save
		S <u>a</u> ve As
	Syntax format: Error processing:	🗱 🛛
	Interactive Continue processing after errors	~
	Contant	
		Browse
		ptions
	Print SPSS Statistics Viewer file on completion	Puono
	Note: Printing is not an option when running a job in the background on a server.	
	Runtime values	
	A runtime value is substituted for a symbol used in syntax. A symbol starts with an @ si	ign and
	must not contain any special characters or spaces. If no default value is specified, do n	otuse
	the 'silent' command line keyword when running the job from the command line.	
	Symbol Default Value User Prompt Quote Value	
	_ Default encoding	
	O Unicode (UTF-8) O Local encoding	
	Close Help	

Menu path:

Utilities > Production Facility



Background mode runs production jobs on a server

Production Job	Status	End Time	Get Job Out
			Cancel Jol Remove Jo



The server side batch engine

		10	Data					8.sav [Base] - IB					1 I ala			×
<u>File E</u>	Edit	View	Data	Transform	<u>A</u> na		rect <u>M</u> arketing	<u>G</u> raphs <u>C</u> ust	m ¥i	Utilities	Add- <u>o</u> ns	Window	Help		ABC	
									==h)				14			
									_					Visible: 17	of 17 Vari	ables
		resp_i	id	year		gender	age	income		store_lo	cation	departm	ent	reason	perf1	
1			1		08 M		31	2784	-				4	1		3 📤
2			2		08 F		31		0 Chi				2	1		5
3			3		08 F		42		0 Lor				2	1		4
4			4		08 F		28		0 Chi				1	1		4
5			5		08 M		53						4	1		4
6			6		08 M		32						4	1		4
7			7		08 F		41	4560					3	1		2
8			8		08 M		46						2	3		1
9			9		08 M		31	4440					2	1		9
10			10		08 F		33						2	1		4
11			11		08 M		34	5500					3	1		3
12			12		08 F		35						4	2		3
13			13		08 M		43		0 Lor				3	1		3
14			14		08 M		45		0 Tok				3	2		3
15			15		08 M		27		0 Tok				3	1		5
16			16		08 F		999		1 Tok				2	2		5
17			17		08 F		32						2	1		4
18		1	18	20	08 F		38	4800	0 Par				3	1		3 -
Data V	iew	Variable '	View					***								
			_						IBI	I SPSS St	atistics Pro	essor is rea	dy	Unicode:OFF		

statisticsb



On the server side

🖾 Administrator: Command Prompt
C:\>cd automationweb
C:\AutomationWeb>dir Volume in drive C has no label. Volume Serial Number is 700B-9CFD
Directory of C:\AutomationWeb
02/12/2014 16:10 (DIR) 02/12/2014 16:10 (DIR) 03/12/2014 09:23 78,336 Automating your analyses - the best 03/12/2014 09:03 602 BuildStreamEnd.txt 24/11/2013 17:16 602 BuildStreamEnd.txt- 24/11/2013 11:38 313 BuildStreamV2.txt 13/03/2009 15:48 341,645 census.sav 28/10/2009 13:31 14,055 customer satisfaction 2008.sav 02/12/2014 10:16 833,761 Getting Started with SPSS Statistic 03/12/2014 15:19 302 KDA(1).sps 20/03/2009 14:02 510 Macro 0.sps 20/03/2009 14:52 1,914 macro 2.sps 13/03/2009 15:48 1,761 macro 3.sps 02/12/2014 09:30 153,096 Offer Slide.pptx 153,096 Offer Slide.pptx 153,096 Offer Slide.pptx 1,988,842 bytes 2 Dir(s) 69,334,593,536 bytes free
C:\AutomationWeb>statisticsb -f KDA(1).sps -type text -out KDA.txt
C:\AutomationWeb>_

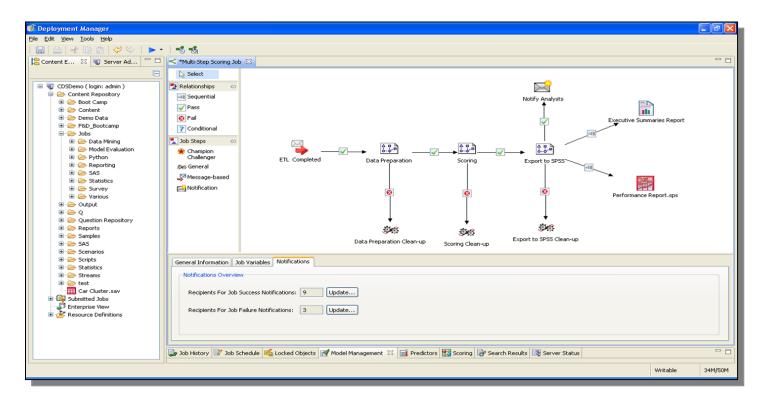
/ISION

 $s \sim$

 Batch jobs can be scheduled to run using the Windows Task Scheduler



IBM/SPSS C&DS is the next level of automation





Contents

- Background
- Levels of automation with syntax and streams
- Automating beyond syntax and streams
- Automating SPSS from the outside



Automating beyond standard syntax - Statistics

- Macros
- Visual Basic
- Python
- Java

• R

More programming power This includes:

- Creating re-usable blocks of code
- Creating our own User Interfaces
- Automating processes beyond SPS
 - e.g. controlling Excel, PowerPoint etc.



Automating beyond standard syntax - Macros

Pros:

- An extension of the SPSS syntax language
- Run inside the same files(s)

Cons:

- They have their own syntactic rules
- Functionally limited
 - Don't support some more advanced programming constructs
 - Can't control other tools



Example Macros

A simple to define a re-usable variable

Define !myfolder () "C:\train\syntax_II\" !enddefine.

Using the macro variable in syntax

GET FILE = !MYFOLDER + 'census.sav'. DATASET NAME census WINDOW=FRONT.



A macro to create a new "command"

DEFINE !CLOSEALL (DATASETS = !CHAREND ("/") /VIEWERDOCS = !CMDEND)

IF (IDATASETS = YES) ITHEN NEW FILE. DATASET CLOSE ALL. IFEND

IF (IVIEWERDOCS = YES) ITHEN OUTPUT CLOSE ALL. IFEND

ENDDEFINE.

Calling that macro

ICLOSEALL DATASETS = YES /VIEWERDOCS = YES.

Automating beyond standard syntax – VB, Python, Java, R

Pros:

- More powerful / widely used languages
- Allow us to add extended functionality
- Go beyond automating SPSS

Cons:

- They run separately so we need to integrate syntax into them
- Need to learn / have access to programming expertise



An example VB script

```
'Begin Description
'This file removes upper diagonal of correlation matrix and highlights
'correlations significant at the .01 level.
'End Description
Sub Main
Dim objPivotTable As PivotTable
Dim objItem As ISpssItem
Dim bolFoundOutputDoc As Boolean
Dim bolFoundOutputDoc As Boolean
Dim lngIndex As Long
Dim objOutputDoc As ISpssOutputDoc
Call GetFirstSelectedPivot(objPivotTable, objItem, bolFoundOutputDoc, bolPivotSelected)
```

```
Call Correlations_Table_Correlations_Create(objPivotTable, objOutputDoc, lngIndex)
'Deactivate the correlation pivot table
objItem.Deactivate
```

End Sub

This script looks inside a correlation table Identifies statistically significant correlations



Extensibility

=				(Custom	Dialog Builder - (Untitled) 🛛 🗕 🗖 🗙
<u>F</u> ile	Edit	t <u>V</u> iew	<u>W</u> inc	low	<u>H</u> elp	Tools ×
		¥ 15		注 [1 🔀	Source List
						Target List
						Check Box
						Combo Box
						List Box
						A Text control
						2 Number control
						Ab Static Text
						🗀 Item Group
						🔛 Radio Group
						E Check Box Group
						🚍 File Browser
						Sub-dialog Button
Dialo	g Prop	oerties				
Prop					Value	
Dialo	og Nar				dialog1	4
	u Loca	ition				
Title Help	File					
		yment Pr	operties			
	eless	-			True	

We can use the Custom Dialog builder in SPSS to create our own UIs and automate behind them With Syntax, Python, R, etc.



A KDA extension

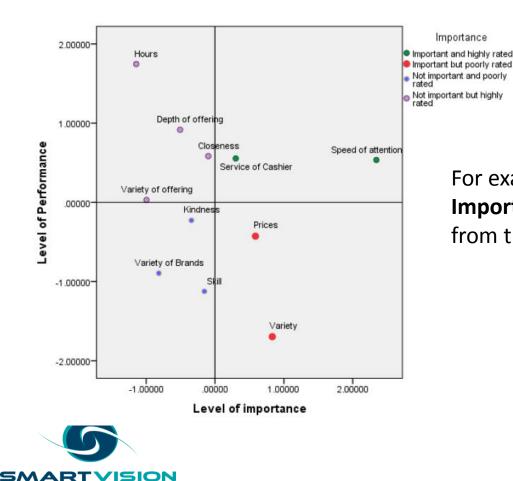
12 C	y Drivers Analysis	×	
Variables:	Characteristics Satisfaction: Sat	Attractive Shopping [sat3] thod y el Group Inc/SV Europe	
OK	Paste Reset Can	cel	

This example (available for download from our web site shortly) was developed by Channel Group in the US

It simplifies several steps beyond the KDA syntax that we ran earlier



A KDA extension



Eur

o p e

For example it automatically produces the **Importance v Performance** quadrant chart from the SPSS regression output

Extension Bundles

PSM Propensity Score Matching SPSSINC CENSOR TABLES Censor cells of a pivot table based on the values of a test statistic SPSSINC CREATE DUMMIES Create a set of dummy variables representing the values of one or more variables and interaction SPSSINC MERGE TABLES Merge the contents of one pivot table in the Viewer into another SPSSINC PROCESS FILES Apply a file of syntax to a set of data files SPSSINC PROCESS FILES_SEARCH Search the cases in a set of SPSS SAV files SPSSINC PROCESS FILES Split a dataset into separate files according to splitting variables SPSSINC TRANS Apply a Python function to case data SPSSINC TURF Perform TURF analysis SPSSINC_RAKE Calculate weights to control totals in up to ten dimensions by rim weighting, i.e. raking STATS OUTPUT ATTRS Set Viewer page attributes and setup via syntax STATS SUBGROUP PLOTS Graphically compare the distributions of a set of variables across a partition of the data	Name	Summary
PLS Partial least squares regression PSM Propensity Score Matching SPSSINC CENSOR TABLES Censor cells of a pivot table based on the values of a test statistic SPSSINC CREATE DUMMIES Create a set of dummy variables representing the values of one or more variables and interaction SPSSINC CREATE DUMMIES Merge the contents of one pivot table in the Viewer into another SPSSINC PROCESS FILES Apply a file of syntax to a set of data files SPSSINC PROCESS FILES Apply a file of syntax to a set of SPSS SAV files SPSSINC SPLIT DATASET Split a dataset into separate files according to splitting variables SPSSINC TRANS Apply a Python function to case data SPSSINC TURF Perform TURF analysis SPSSINC RAKE Calculate weights to control totals in up to ten dimensions by rim weighting, i.e. raking STATS REGRESS PLOT Plots useful in assessing regression relationships STATS SUBGROUP PLOTS Graphically compare the distributions of a set of variables across a partition of the data		
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STATS SUBGROUP PLOTS Graphically compare the distributions of a set of variables across a partition of the data		
STATS_TABLE_CALC Calculate with pivot table cells		
	STATS_TABLE_CALC	Calculate with pivot table cells

Typically written in Python (or R)

Check out the SPSS Developer Central for more resources

www.ibm.com/spss/devcentral



R

	The R Project for Statistical Computing
	Getting Started
[Home] Download CRAN	R is a free software environment for statistical computing and graphics. It compiles and runs on a wide variety of UNIX platforms, Windows and MacOS. To download R , please choose your preferred CRAN mirror.
R Project	If you have questions about R like how to download and install the software, or what the license terms are, please read our answers to frequently asked questions before you send an email.
About R Contributors	News
What's New? Mailing Lists	• R version 3.2.0 (Full of Ingredients) has been released on 2015-04-16.
Bug Tracking	R version 3.1.3 (Smooth Sidewalk) has been released on 2015-03-09.
Conferences Search	The R Journal Volume 6/2 is available.
	• useR! 2015, will take place at the University of Aalborg, Denmark, June 30 - July 3, 2015.
R Foundation Foundation Board Members Donors Donate	useR! 2014, took place at the University of California, Los Angeles, USA June 30 - July 3, 2014.

Contributed Packages

Available Packages

Currently, the CRAN package repository features 6646 available packages.

Table of available packages, sorted by date of publication

Table of available packages, sorted by name

http://www.r-project.org/



An Alternative KDA in R

* Install additional packages

BEGIN PROGRAM R.

install.packages("kappalab") install.packages("relaimpo") require(relaimpo)

testdata = spssdata.GetDataFromSPSS()

END PROGRAM

* now run the shapley regression.

BEGIN PROGRAM R.

regdata = spssdata.GetDataFromSPSS()

```
reg1 <- Im( overallrating ~ characteristic_1 +characteristic_2 +characteristic_3 +characteristic_4 +characteristic_5 +characteristic_6 +characteristic_7 +characteristic_8 +characteristic_9 +characteristic_10 +characteristic_11, data=regdata)</p>
```

summary(reg1)

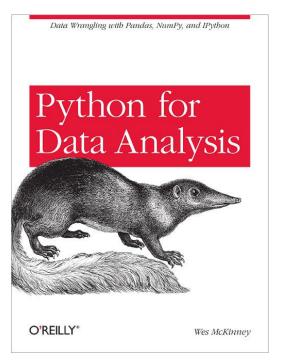
```
shap <- calc.relimp(reg1, rela=TRUE)
shap
```

END PROGRAM



Python

- In SPSS Python is in the first instance a more powerful scripting language
- It can also be used for Data Analysis





The Create Dummy Variables extension

ta Create Du	immy Variables			
Variables: Variab	Create Dummy Variables for:		Variable Creation	
	Main Effect Dummy Variables		Label	
HIGHEST YEAR SCHOOL COMPLETED, FAT HIGHEST YEAR SCHOOL COMPLETED, MOT	Create main-effect dummies	WORKING_STAT_1	wrkstat=WORKING FULLTIME	
Dummy Variable Labels	Root Names (One Per Selected Variable): WORKING STAT	WORKING_STAT_2	wrkstat=WORKING PARTTIME	
© Use <u>v</u> alue labels	Macro Name:	WORKING_STAT_3	wrkstat=TEMP NOT WORKING	
© U <u>s</u> e values		WORKING_STAT_4	wrkstat=UNEMPL, LAID OFF	
Value Order Ascending	Two-Way Interactions	WORKING_STAT_5	wrkstat=RETIRED	
© Descending	Root Name:	WORKING_STAT_6	wrkstat=SCHOOL	
Macros	Macro name:	WORKING STAT 7	wrkstat=KEEPING HOUSE	
Omit first dummy category from macro definitions		WORKING_STAT_8	wrkstat=OTHER	
Note: It is conventional to start macro names with !.	Three-Way Interactions			
Measurement Level Usage © Do not create dummies for scale variable values	Create dummies for all three-way interactions Root Name:			
© <u>C</u> reate dummies for all variables	Macro name:			
This dialog requires the Python Essentials				
OK Paste	Reset Cancel Help	1		



The Propensity Matching extension

Propensity Score Matching >						
Variables:	nsity So	Group Indicator:	Qptions			
 income rincome region sei sei This procedure runs a logistic regression on the group indicator and then uses the resulting propensity variable to select controls for the cases The procedure requires the Statistics Regression module and the Python Essentials At least version 1.3.0 of the FUZZY extension command is required 	*	Name for Propensity Variable (must not already exist): prediction Match Tolerance: 0.2 Case ID: fid Match ID Variable Name(must not already exist): matchid Output Dataset Name (must not already exist): matched				
OK Paste Reset Cancel Help						



Contents

- Background
- Levels of automation with syntax and streams
- Automating beyond syntax and streams
- Automating SPSS from the outside

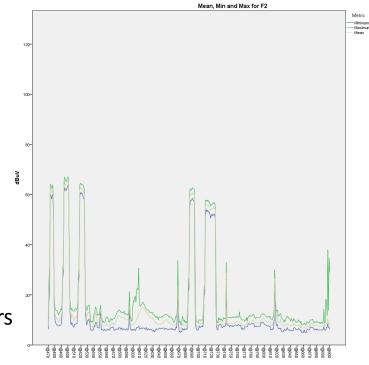


Automating from the outside

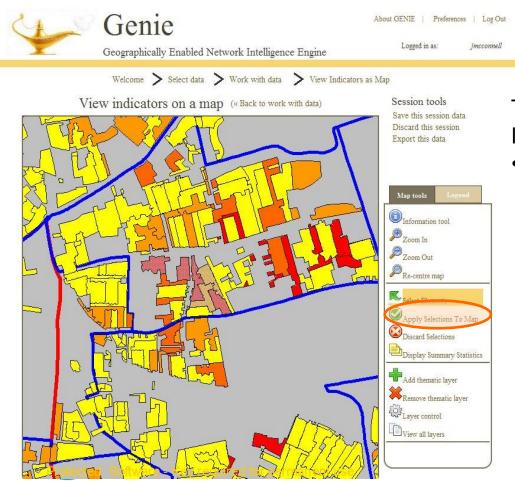
Browse
Clear
Browse
Run
Close

This UI runs a standalone app

- a) Reads and cleans data coming from sensors
- b) Produces summary graphs as jpegs for integration into reports







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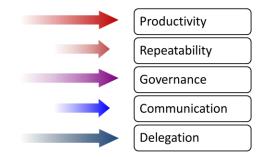
o p e

This on-line GIS app is designed for local planners

 It runs factor analysis models based on selected criteria to create indices of sustainability

In Summary

- The interface to R allows us to mix and match R analysis with SPSS Analysis inside the SPSS UI
- It is possible to automate just about anything in and around SPSS
- This can lead to significant time saving, increased productivity, higher quality and better governance
- As usual the key question is whether the build (development) time is worth investing
 - Does it save time, money etc. in the long run?





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Summary, Next Steps & Close

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