

# IBM SPSS Predictive Analytics Workshop with IBM SPSS Modeler

Add Date and Location and Presenter Name



### Agenda

- Welcome and Introductions
- Overview of Predictive Analytics
- IBM SPSS Modeler Overview
- Exercise 1: Predictive in 20 Minutes
- IBM SPSS Modeler Basics
- Predictive Analytics Methodology and Applications
- Break
- Exercise 2: Find Patterns and Groups
- Exercise 3: Understand the Past, Predict the Future
- Exercise 4: Deploy Insights
- Integrating Predictive Analytics with Planning and BI
- Wrap up and Next Steps





### Introductions

- 1. Name and Organization you represent
- 2. Have you done any predictive analytics projects before today?
- 3. Do you have any experience with any IBM SPSS Products?
  - **%** SPSS Statistics
  - **X** SPSS Modeler (formerly Clementine)
  - SPSS Text Analytics for Surveys (TAFS)
- 4. What are your objectives for today?





### Today's presenter: Lorem Ipsem



XXXX@us.ibm.com

XXX-XXX-XXX



linkedin.com/in/xxxxxx



**Project Experience** 

- •XXX
- •YYY
- •ZZZ



### Purpose of the workshop

- Introduction to predictive analytics
- Stimulate thinking about how predictive analytics would benefit your organization
- Demonstrate ease of use of powerful technology
- Get experience in "doing" predictive analytics
- Explore multiple predictive analytics techniques
- Understand the role of predictive analytics in decision optimization, planning and business intelligence





### Advanced Analytics is For Everyone



Advanced analytics is the analysis of all kinds of data using sophisticated quantitative methods (i.e., statistics, descriptive and predictive data mining, simulation and optimization) to produce insights that traditional approaches to business intelligence (BI) — such as query and reporting — are unlikely to discover.

IBM. Ó

### Realizing Power and Insight in your Data

#### **Descriptive Analytics**





### Analytics Driven Organizations Reap Rewards

#### Front Runners outperform on







Business outcomes (69%)

Competitive advantage (53%)

Revenues (60%)

#### by using data and analytics

Source: IBM Institute for Business Value (IBV)



### Areas for Predictive Analytics





### **Predictive Analytics Use Cases**



#### What are some use cases from your organization?

### **Predictive Analytics in Action**



#### **FleetRisk Advisors**

Helps trucking operators prevent accidents by building stronger & faster risk prediction models

- > 20% reduction in minor accidents
- > 30% increase in driver retention rates



#### **U.S. Insurer**

Analyzes and links claims and medical data, to prevent fraud and fast-track legitimate claims

- USD 22 Millon ROI anticipated, and
- 100% payback anticipated from fraud reduction
- Over 85% accuracy for predicting independent medical failures



#### ASTRON

Uses streaming analytics to deliver insights from the world's largest radio telescope

- > 99% faster identification of data
- Analyzes >1 exabyte of data daily
- Integrates data from >3,000 dishes & antennas to form the largest & fastest radio telescope in the world



#### Oak Lawn Marketing

Understand customer buying patterns for targeting infomercials

- 159% boost in the average monthly rate of customers who return to shop compared to the previous year
- 400% increase in expected total revenue over a three year period



### IBM SPSS Modeler: At a glance





### Exercise: Predictive in 20 Minutes

#### Goal:

Identify who has responded to a marketing campaign

Approach:

- Use a data extract from a CRM
- Prepare data for modeling
- Define which fields to use
- Choose the modeling technique
- Automatically generate a model to identify who has responded
- Review results

Why?

 To save marketing cost and increase marketing response, identify those likely to respond and focus marketing efforts on those prospects.



Methodology



### **CRISP-DM**

- CRoss-Industry Standard
   Process for Data Mining
- Describes Components of Complete Data Mining Project Cycle
- Shows Iterative Nature of Data Mining
- Vendor and Industry Neutral



Data mining is a key discipline for applying predictive analytics



### Modeling Techniques in IBM SPSS Modeler

Technique	Usage	Algorithms
Classification (or prediction)	• Used to predict group membership (e.g., will this employee leave?) or a number (e.g., how many widgets will I sell?)	<ul> <li>Auto Classifiers, Decision Trees, Logistic, SVM, Time Series, etc.</li> </ul>
Segmentation	<ul> <li>Used to classify data points into groups that are internally homogeneous and externally heterogeneous.</li> <li>Identify cases that are unusual</li> </ul>	<ul> <li>Auto Clustering, K- means, etc.</li> <li>Anomaly detection</li> </ul>
Association	<ul> <li>Used to find events that occur together or in a sequence (e.g., market basket)</li> </ul>	<ul> <li>APRIORI, Carma, Sequence</li> </ul>



### Extend Capabilities through R





### **SPSS Modeler Editions**





### **Uncovering Patterns in Unstructured Data**

- Text Analytics
  - Natural Language
  - Sentiment Analysis
- Entity Analytics
  - Disambiguate identity
  - People, places, things
- Social Network Analysis

\$EA-ID \$EA-SRC key

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Table Annotations

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Because people communicate with words, not numbers, it has become critical to be able to mine text for its meaning and to sort, analyse, and understand it in the same way that data has been tamed. In fact, the two basic types of information complement each other, with data supplying the "what" and text supplying the "why".

Source IDC: "Text Analytics: Software's Missing Piece?"

### **Text Analytics**

Uses natural language processing heuristic rules and statistical techniques to reveal conceptual meaning in text

Extracts concepts from text and categorizes them

Makes unstructured qualitative data more quantifiable, enabling the discovery of key insights from sources such as:

- -Documents
- -Survey responses
- -Call center Notes
- -Social Media
- -Web Pages



## Entity Analytics Overview

**Identify Matching Entities** 

- Analysts Struggle to Match Entities, Especially From Diverse Sources
   Natural variability (Bob vs. Robert)
   Errors (transposed month and day)
   Fabrications (fake identities)
- Enhances Model Accuracy Model Against a Complete, Accurate Entity
- Multiple Applications
  - & Business: is this the same order the customer submitted?
  - **%** Fraud: is this the same person who already defaulted on a loan?
  - & Government: is this the same vehicle that was carrying illegal content last time?
  - R Policing: Is this the same person who called us before?



Resolved Entity			
Name	Elizabeth Lisa Johns		
	Liz Johns		
	Beth L Johns-Parker		
	BL Johns		
Addr1	123 Main Street		
	777 Park Road		
	33 Red Dr		
	33 Reed Dr		
City	New York,		
	White Plains,		
	Mamaroneck		
State	NY		
Postal	11732, 10354		
Phone	212-733-1234		
	914-698-2234		
DOB	6/21/1954		
Defaults	Yes		
Income	\$48000		
Credit Debt	\$12.722		
Other Debt	\$9.009		
Debt to Incom	e 113.5		
Prev Default?	True		
Pending Loan	True		



### **Social Network Analysis**

- Used to indentify depth and reach of social media content.
- Focuses around identifying groups, leaders and probabilities that others will churn based on influence
- Enhances existing churn predictions of Modeler
- Expressed as two new nodes in the Sources Palette
  - & Group Analysis what are the groups in my data and who are the leaders
  - X Diffusion Analysis uses existing churn information to determine who else that churner is likely to influence to leave







### Break - please return in 10 minutes







Classify customers into groups based on underlying characteristics

#### Understand the Past, Predict the Future



Model response to marketing offers using historical data

#### Deploy Insights on new projects



Leverage models to create insights on new customers



Classify customers into groups based on underlying characteristics

#### Understand the Past, Predict the Future



Model response to marketing offers using historical data

#### Deploy Insights on new projects



Leverage models to create insights on new customers

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### **Exercise 2: Find Patterns and Groups**

Goal:

**X** Create segments of customers

Approach:

**X** Merge disparate data sources

& Define which fields to use



& Automatically generate a model to group customers

& Apply business terms to new groups

**X** Export newly created groups to database

Why?

**X** Better customer understanding (demographics, socio-economic etc)

**X** Tailored messages for each group/segment

**X** Personal and more relevant for consumers



Classify customers into groups based on underlying characteristics

#### Understand the Past, Predict the Future



Model response to marketing offers using historical data

#### Deploy Insights on new projects



Leverage models to create insights on new customers

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### Exercise 3: Understand the Past, Predict the Future

Goal:

- **X** Use a data extract from a CRM
- **%** Define which fields to use
- **%** Choose the modeling technique
- & Automatically generate a model to identify who is likely to respond
- **X** Review results

Why?

- **X** Target those likely to respond to offers to increase revenue, cut costs
- $\boldsymbol{\textbf{x}}$  Using unstructured data improves modeling accuracy and provides more insight





Classify customers into groups based on underlying characteristics

#### Understand the Past, Predict the Future



Model response to marketing offers using historical data

#### Deploy Insights on new projects



Leverage models to create insights on new customers

![](_page_31_Picture_1.jpeg)

### Exercise 4: Deploy Insights on New Projects

Goal:

- X Use new customer records who have never received an offer
- & Leverage the text extraction and classification models used in exercise 3

![](_page_31_Picture_7.jpeg)

- & Automatically generate scores of who is likely to respond
- **%** Review results
- **%** Deploy results for use by marketing team

Why?

- **X** Create models on large populations for robust, representative results, and easily score smaller record batches when necessary, on-demand.
- Apply consistent and seamless methodology over many model scoring iterations, ensuring quality control.

![](_page_32_Figure_2.jpeg)

![](_page_33_Picture_1.jpeg)

### **IBM SPSS Predictive Analytics**

![](_page_33_Figure_3.jpeg)

![](_page_33_Picture_4.jpeg)

![](_page_34_Picture_1.jpeg)

### Integration with IBM Cognos Analytics

![](_page_34_Figure_3.jpeg)

![](_page_35_Picture_1.jpeg)

### Flexible deployment

![](_page_35_Picture_3.jpeg)

Flexible Deployment Options, Including Cloud, Supported By An Infrastructureagnostic Platform

![](_page_35_Picture_5.jpeg)

Integrated With IBM's High-performance Systems Built for Big Data Analytics

![](_page_35_Picture_7.jpeg)

![](_page_35_Picture_8.jpeg)

Embedded Into Operational, Mobile and/or Cloud-based Applications

![](_page_35_Figure_10.jpeg)

![](_page_36_Figure_2.jpeg)

![](_page_37_Picture_1.jpeg)

### **Predictive Analytics for All**

![](_page_37_Picture_3.jpeg)

Deep Predictive Capabilities With Indatabase, R and Python Integration, and Integrated Deployment

![](_page_37_Figure_5.jpeg)

![](_page_37_Picture_6.jpeg)

Visual Workbench for Building Models of Any Complexity With Ability to Automate and Combine Tasks

![](_page_37_Figure_8.jpeg)

![](_page_37_Picture_9.jpeg)

Simple, Easy-to-Use, Visual, Guided Analytical Discovery, Intelligent Automation and Visual Storytelling

![](_page_37_Figure_11.jpeg)

![](_page_38_Picture_1.jpeg)

### Workshop takeaways

#### Easy to use, visual interface

- Short timeframe to be productive with actionable results
- Does not require knowledge of programming language
- No proprietary data formats
- Open architecture

#### **Business results focused**

- Leverages the investments already made in technology
- Cost effective solution that delivers powerful results across organization
- Full range of algorithms for your business problems
- Big Data enabled (Hadoop, SQL Pushback)

#### **End-to-end solution**

- Data preparation through real time interactions
- Use structured, unstructured and semi-structured data
- Integrated portfolio for business analytics
- Scales from a single desktop to an enterprise deployments

![](_page_39_Picture_1.jpeg)

### **Call To Action**

![](_page_39_Picture_3.jpeg)

Demo the software download a trial version of SPSS modeler and create your own use-cases with your own data

![](_page_39_Picture_5.jpeg)

**IBM SPSS Resource Center** stay connected to this site to learn more about Modeler capabilities, latest use-case stories, and upcoming training opportunities

![](_page_39_Picture_7.jpeg)

Connect with IBMers stay connected with IBM SPSS representatives. We would love to hear your success and help with any roadblocks to enablement

![](_page_40_Picture_2.jpeg)

![](_page_41_Figure_1.jpeg)

### **Getting Started**

![](_page_41_Figure_3.jpeg)

![](_page_42_Picture_1.jpeg)

### Partner projects

![](_page_43_Picture_1.jpeg)

![](_page_43_Picture_2.jpeg)

![](_page_44_Picture_1.jpeg)

### **IBM Watson Analytics**

![](_page_44_Figure_3.jpeg)

Single Analytics Experience

![](_page_44_Figure_5.jpeg)

Fully Automated Intelligence

![](_page_44_Picture_7.jpeg)

![](_page_44_Picture_8.jpeg)

Guided Analytic Discovery

IBM. 😽

### Single Interface ... Explore > Predict > Assemble

![](_page_45_Figure_3.jpeg)

![](_page_46_Picture_1.jpeg)

Appendix

## **Suggested Books**

![](_page_47_Picture_1.jpeg)

### IBM SPSS Modeler Cookbook

From Amazon.com

**XIBM SPSS Modeler Cookbook X**by Keith McCormick, Dean Abbott, Meta S. Brown, Tom Khabaza, Scott R. Mutchler **X**Paperback - 382 pages (October 2013) (also on Kindle) **X**ISBN : 1849685460

 Written by those who teach and have been working with IBM SPSS Modeler since its beginnings. Full of practical examples that span the full gamut of capabilities.

![](_page_47_Picture_6.jpeg)

![](_page_48_Picture_1.jpeg)

### Data Mining Overview

From Amazon.com

**%Paperback:** 512 pages **%Publisher:** Wiley; 1 edition
(December 28, 1999) **%Language:** English **%ISBN-10:** 0471331236 **%ISBN-13:** 978-0471331230;

 Good introductory text on data mining for marketing from two top communicators in the field

![](_page_48_Picture_6.jpeg)

![](_page_49_Picture_1.jpeg)

# Handbook of Statistical Analysis and Data Mining Applications

![](_page_49_Picture_3.jpeg)

- Handbook of Statistical Analysis and Data Mining Applications
- Robert Nisbet, John Elder IV, and Gary Miner
- Academic Press (2009)
- ISBN-10: 0123747651
- An excellent guide to many aspects of data mining including Text mining.

![](_page_50_Picture_1.jpeg)

### **Data Mining Algorithms**

From Amazon.com

&Data Mining: Practical Machine Learning Tools and Techniques with Java Implementations
&by Eibe Frank, Ian H. Witten
&Paperback - 416 pages (October 13, 1999)
&Morgan Kaufmann Publishers;
&ISBN: 1558605525;

 Best book I've found in between highly technical and introductory books. Good coverage of topics, especially trees and rules, but no neural networks.

![](_page_50_Picture_6.jpeg)