# INTRODUCTION TO SAS® TEXT MINER™



CUSTOMER LOYALTY TEAM • Support You Can Count On

## **TODAY'S AGENDA**

### INTRODUCTION TO SAS® TEXT MINER™

- Define data mining
- Overview of SAS<sup>®</sup> Enterprise Miner<sup>™</sup>
- Describe text analytics and define text data mining
- Text Mining Process
- SAS<sup>®</sup> Text Miner<sup>™</sup>
- Illustrate text mining by example
- Q&A



# A QUICK INTRODUCTION TO DATA MINING



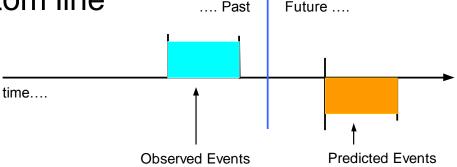
# WHAT IS DATA MINING?

Turning increasing amounts of raw data into useful information

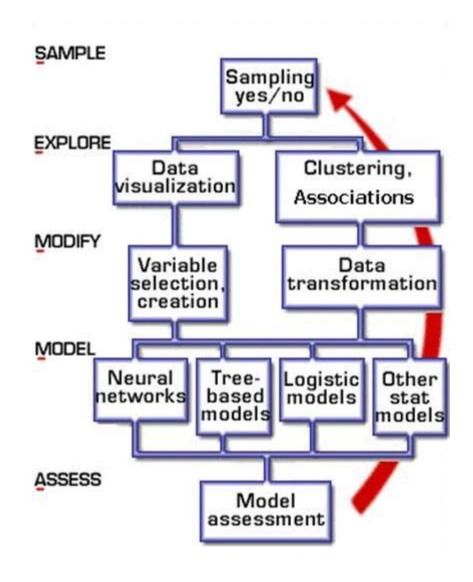


### **DATA MINING IS:**

- Discovering patterns, trends and relationships represented in data
- Developing models to understand and describe characteristics and activity based on these patterns
- Using insights to help evaluate future options and take fact-based decisions
- Deploying scores and results for timely, appropriate action that affects the bottom line



## **SEMMA** DATA MINING WITH SAS® ENTERPRISE MINER™



# SAS® ENTERPRISE MINER™ MODEL DEVELOPMENT PROCESS

Model Transform Variables Association Decision Tree DMDB Input Data Neural Network Comparison Cluster SOM/Kohonen Impute A utoNeural SVM Score File Import Dmine Regression Segment Profile Variable Partial Least Sample Replacement Graph Explore Selection Squares Interactive Binning Data Partition Mark et Bask et ⇒ DMNeural Regression necisions Ensemble Rules Builder + Merge StatExplore M Cutoff Rule Induction Variable Clustering Gradient Boosting Gradient X Drop Filter TwoStage Principal Components LARS MultiPlot Model Import Append MBR Path Analysis M Time Series

**Text** 







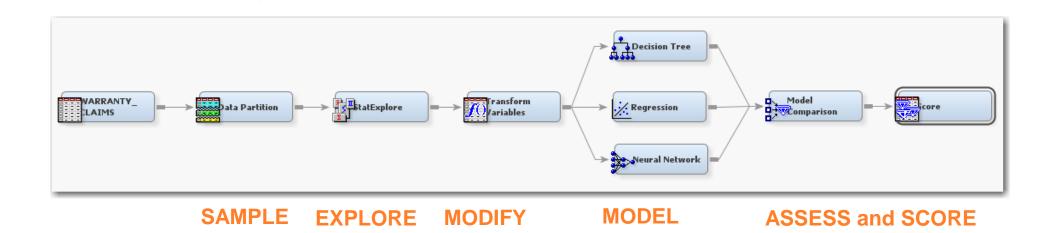






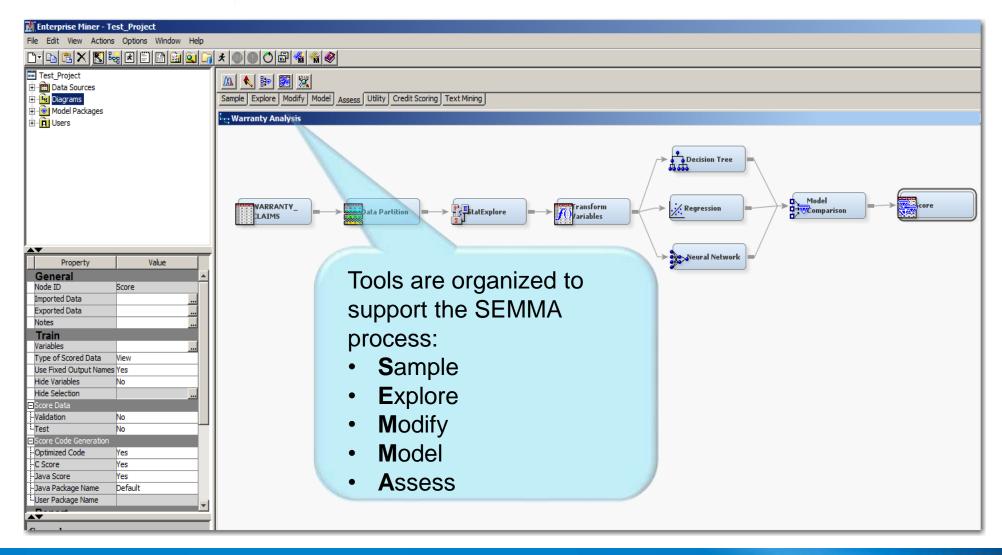


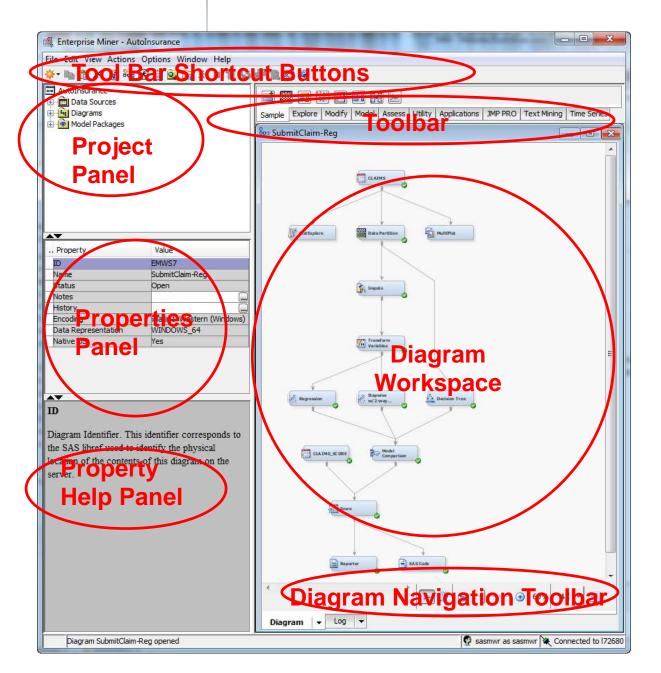
## **SEMMA** A GLIMPSE OF SAS® ENTERPRISE MINER™





## **SEMMA** A GLIMPSE OF SAS® ENTERPRISE MINER™





# **SAS Enterprise Miner GUI**



# SAS® TEXT MINER™



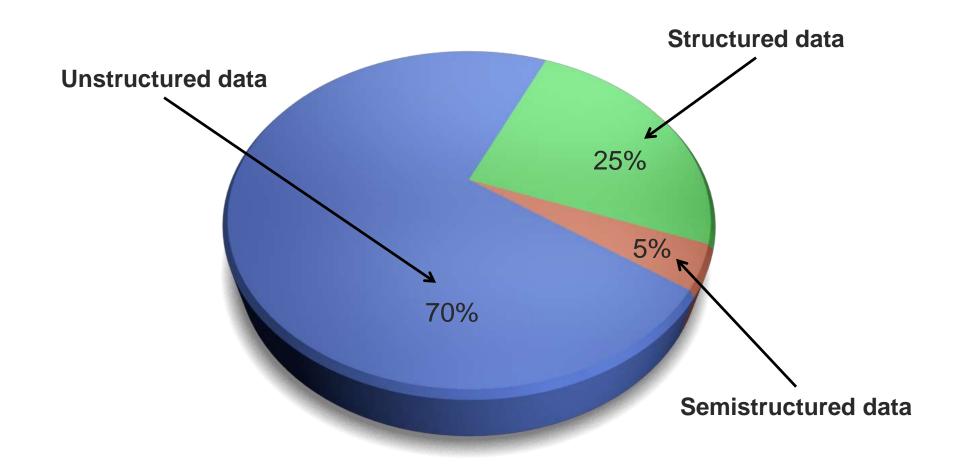
## **WHY MINE TEXT?**

# IS VALUABLE INFORMATION "LOCKED AWAY" IN UNSTRUCTURED DATA?





## **UBIQUITY** UNSTRUCTURED AND SEMI-STRUCTURED DATA





## **EXAMPLE** DATA

## **Structured Data**

- Age Group = 60+
- Satisfaction = Not Very
- Rewards Customer= No
- Total Hold Time = 8

### **Unstructured Data**

 they called me so i returned their call because it was cut off in the middle of the conversation. every time they call me, they're cut off.



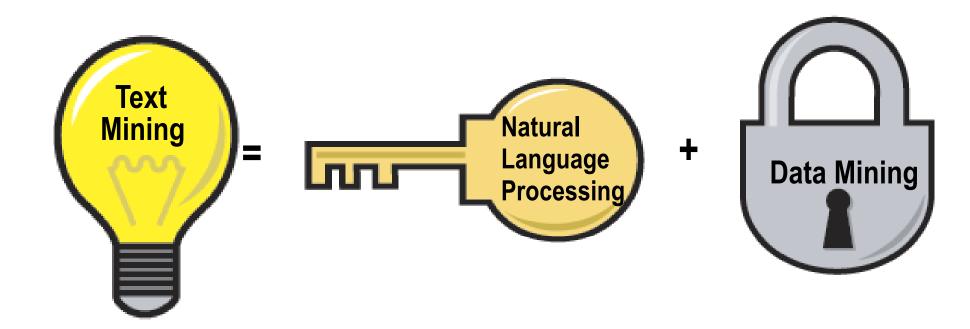
### WHY MINE TEXT? WHAT CAN BE LEARNED FROM UNSTRUCTURED DATA?

- Are any of these documents related to one another based on their contents and the characteristics of their contents?
- What are the key topics, themes or concepts being discussed?
- Are there emerging issues?
- Do the documents contain potentially valuable information that

could improve predictive models?



The process of discovering and extracting meaningful patterns and relationships from text collections



### **TEXT MINING TWO GENERAL GOALS**

- 1. Pattern Discovery (Unsupervised Learning)
- 2. Prediction (Supervised Learning)

These are the same general goals of data mining.

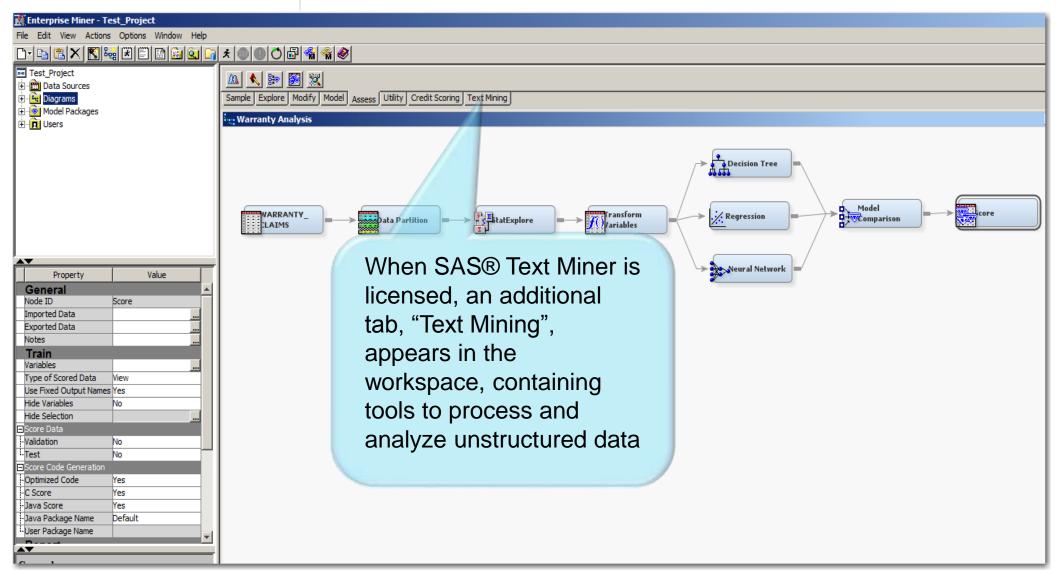
# WHAT IS THE TEXT MINING PROCESS?



# **TEXT MINING** THE PROCESS

Text Preprocessing **Text Parsing** Transformation (Dimension Reduction) **Document Analysis** 

# SAS® TEXT MINER ADD-ON



# **TEXT MINING** THE PROCESS

Text Preprocessing **Text Parsing** Transformation (Dimension Reduction) **Document Analysis** 

### **INPUT DATA** TEXT MINER

- The expected SAS data set for text mining should have the following characteristics:
  - One row per document
  - A document ID (suggested)
  - A "text" column
- The "text" column can be either:
  - The actual full text of the document, up to 32,000 characters
  - A pointer to a text file (\*.txt, \*.html) located on the file system
- The SAS data set can also have structured data and a target variable (dependent variable, response variable)

### **TEXT IMPORT NODE**



- Enables you to create data sets dynamically from files contained in a directory or from the Web.
- Takes an import directory containing text files in potentially proprietary formats such as MS Word and PDF files as input.
- Extracts the text from the files, places a copy of the text in a plain text file, and a snippet (or possibly even all) of the text in a SAS data set.
- If a URL is specified, the node will crawl Web sites and retrieve files from the Web
- The output of a **Text Import** node is a data set that can be imported into the **Text Parsing** node.

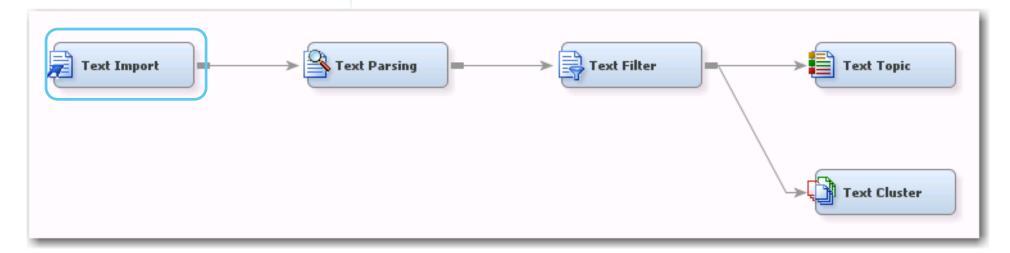
# **EXAMPLE INPUT DATA**

TOT_AGT_HOLD_DUR	TOT_AGT_TALK_DUR	CALL_REAS_1	CSAT_BRAND_RELTNSHIP	ASAT_RESPONSE	CSAT_OVERALL	v_dissat	v_call_reason
8.0	969.0	3571 - Stop Payment on an account	Very	Completely satisfied	Somewhat	i called th	just charges that were not supposed b
5.0	546.0	3071 - Statement Questions/Billing inqu	Not at all satisfied	Not at all satisfied	Not at all satisfied	i was ne	i called trying to get a hold of one repr
4.0	1162.0	3114 - Dispute the Validity of a Fee	Somewhat	Very	Not very	i'm satisfi	there was a problem with a charge.
13.0	283.0	3101 - Dispute A Merchant Charge	Somewhat	Not at all satisfied	Not at all satisfied	i'm still ca	i made a charge on my credit card prof
75.0	573.0	3157 - Questions about Account Securi	Somewhat	Somewhat	Not at all satisfied	i keep try	trying to get a hold of the fraud depart
51.0	686.0	3075 - Request Hardship	Not at all satisfied	Not at all satisfied	Not at all satisfied	because i	to receive financial assistance.
106.0	415.0	3034 - Advanced Payment/Set Up Pay	Somewhat	Not very	Not very	i was calli	i had received several messages from
66.0	810.0	3085 - Notify Of Late Payment	Somewhat	Very	Somewhat	just i alw	i called them because i told them i cou
163.0	693.0	3031 - Make a Payment	Not very	Not at all satisfied	Not at all satisfied	they wer	i tried to pay my monthly statements v.
69.0	1537.0	3075 - Request Hardship	Not at all satisfied	Not at all satisfied	Not very	it was ac	the reason was to let them know what
164.0	1132.0	3072 - Haven't Received Statement	Somewhat	Completely satisfied	Somewhat	i didn't g	they changed my account number and
101.0	880.0	3037 - Change or Inquiry regarding Pa	Somewhat	Somewhat	Not very	essentiall	kind of a lengthy reason, bottom line,
76.0	454.0	3031 - Make a Payment	Somewhat	Not very	Somewhat	i didn't m	to payoff my credit card.
274.0	941.0	3114 - Dispute the Validity of a Fee	Not at all satisfied	Completely satisfied	Not very	just the	about the \$25 fee that kept popping u.
35.0	597.0	3034 - Advanced Payment/Set Up Pay	Very	Completely satisfied	Somewhat	there wa	to get some kind of payment arrangem
69.0	629.0	3075 - Request Hardship	Not very	Not at all satisfied	Not very	my wife	my wife and i are both unemployed, and
50.0	376.0	3037 - Change or Inquiry regarding Pa	Somewhat	Not at all satisfied	Not at all satisfied	i was tryi	just a lost of job and things were getti-
52.0	941.0	3034 - Advanced Payment/Set Up Pay	Not very	Not very	Not at all satisfied	i felt reall	just to communicate about why i was r
53.0	504.0	3031 - Make a Payment	Somewhat	Somewhat	Somewhat	the repre	to get caught up on my payments.
38.0	1407.0	3034 - Advanced Payment/Set Up Pay	Somewhat	Completely satisfied	Somewhat	some thi	they had called and i wanted to try to
58.0	660.0	3021 - Change Name/Address on Acco	Not very	Not at all satisfied	Not at all satisfied	i guess it	change of address.
53.0	1032.0	3075 - Request Hardship	Somewhat	Very	Somewhat	i don't lik	the first time i called, someone from

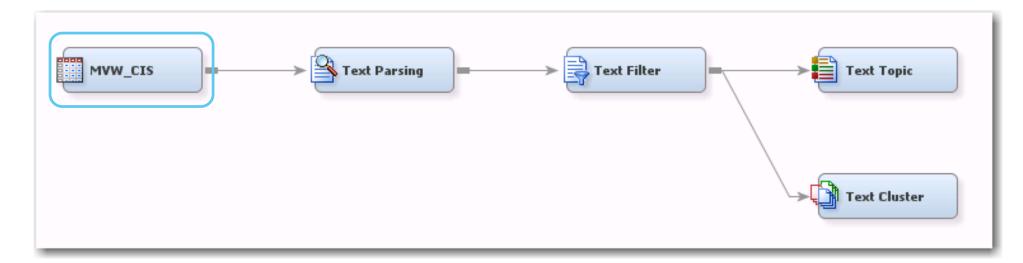


### **EXAMPLE**

### **TEXT MINING PROCESS FLOWS**



External Documents



Text in
Column or
Document
location in
column

# TEXT MINING PROCESS

Text Preprocessing

Text Parsing

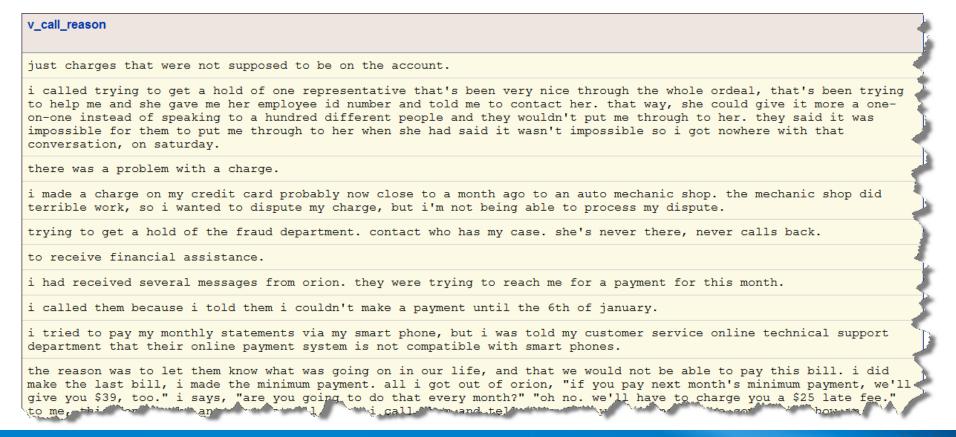
Transformation (Dimension Reduction)

**Document Analysis** 





- Text parsing decomposes textual data and generates a quantitative representation suitable for data mining purposes.
- It transforms this:





 Text parsing decomposes textual data and generates a quantitative representation suitable for data mining purposes.

• ... into this:

	Terms						
	TERM	FREQ	# DOCS	KEEP ▼	WEIGHT	ROLE	ATTRIBUTE
+	credit card	5213	2874	<b>V</b>	0.0050	Noun Group	Alpha
	credit	2076	1579	<b>V</b>	0.044	Noun	Alpha
+	payment	2103	1387	<b>V</b>	0.0040	Noun	Alpha
+	account	1644	1160	<b>V</b>	0.065	Noun	Alpha
+	want	1406	1114	<b>V</b>	0.062	Verb	Alpha
+	pay	1475	927	<b>V</b>	0.093	Verb	Alpha
+	orion	1210	857	<b>V</b>	0.121	Noun	Alpha
+	know	948	737	<b>V</b>	0.069	Verb	Alpha
+	activate	838	730	<b>V</b>	0.058	Verb	Alpha
+	bill	926	684	<b>V</b>	0.075	Noun	Alpha
+	contact	763	682	<b>V</b>	0.118	Verb	Alpha
+	charge	842	651	<b>V</b>	0.014	Noun	Alpha
	interest	743	596	<b>V</b>	0.098	Noun	Alpha
+	charge	705	541	<b>V</b>	0.085	Verb	Alpha
+	receive	612	503	<b>V</b>	0.018	Verb	Alpha
+	rate	621	498	<b>V</b>	0.097	Noun	Alpha
+	halance	614	496	J	0.01	Noun	Alpha





 Documents are represented internally in SAS® Text Miner by a vector that contains the frequency of how many times each term occurs in each

document.

Term	Role	Attribute	Freq	# Docs Keep
	Noun	Alpha	17881	4760N
- be	Verb	Alpha	11609	4099N
- card	Noun	Alpha	4043	2842Y
- not	Adv	Alpha	5016	2446N
have	Verb	Alpha	3604	2073N
- get	Verb	Alpha	2658	1803N
+ do	Verb	Alpha	3541	1792N
- credit	Noun	Alpha	2078	1580Y
+ call	Verb	Alpha	2177	1467N
payment	Noun	Alpha	2103	1387Y
- make	Verb	Alpha	1609	1214N
+ account	Noun	Alpha	1644	1160Y
+ want	Verb	Alpha	1404	1112Y
credit card	Noun Group	Alpha	1168	990 Y
on	Adv	Alpha	1119	935N
+ pay	Verb	Alpha	1475	927 Y
+ say	Verb	Alpha	1478	872N
ust	Adv	Alpha	1082	871N
+ go	Verb	Alpha	1244	860N
orion	Noun	Alpha	1209	856Y
+ know	Verb	Alpha	941	732Y
- activate	Verb	Alpha	838	730 Y
⊦ try	Verb	Alpha	859	715N
+ bill	Noun	Alpha	926	684Y
+ tell	Verb	Alpha	990	678N
- contact	Verb	Alpha	749	675Y
hen	Adv	Alpha	923	652N
- charge	Noun	Alpha	842	651Y
vhat	Adv	Alpha	771	609N

# Text Parsing

# STEMMING PART OF SPEECH

- Determines if the word is a common noun, verb, adjective, proper noun, adverb, etc.
- Disambiguate parts of speech when a word is used in a different context,
  - I wish that my bank did not have a service charge for using other vendor ATM's.
  - You can bank on either Germany or England winning the world cup next year.

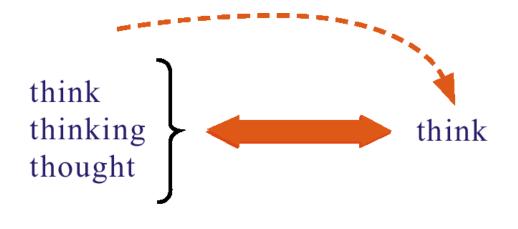
### **ENTITY EXTRACTION**

**Places**White House



People's Names
James H. Goodnight





**Dates** 



# PARTS OF SPEECH IN SAS® TEXT MINER



- Abbr (abbreviation)
- Adj (adjective)
- Adv (adverb)
- Aux (auxuliary or modal)
- Conj (conjunction)
- Det (determiner)
- Interj (interjection)
- Noun (noun)
- Num (number or numeric expression)

- Part (infinitive marker, negative participle, or possessive marker)
- Pref (prefix)
- Prep (preposition)
- Pron (pronoun)
- Prop (proper noun)
- Punct (punctuation)
- Verb (verb)
- VerbAdj (verb adjective)

# ENTITIES IN SAS® TEXT MINER

# STANDARD ENTITIES (IDENTIFIED OUT-OF-THE-BOX)



- Address
- Company
- Currency
- Date
- Internet
- Location
- Measure
- Organization
- Percent

- Person
- Phone
- Prop\_Misc (proper noun ambiguous classification)
- SSN (U. S. Social Security Number)
- Time
- Time\_Period
- Title
- Vehicle (motor vehicle)



# ADDITIONAL PARSING STEPS



# Specify Start/Stop/Synonym Lists

- Filtering out low information words such as
  - articles (e.g. the, a, this)
  - prepositions (e.g. of, from, by)
  - conjunctions (e.g. and, but, or)
- Consider document subject matter as well as domain-specific language and acronymns

# Vertical dictionaries

Automatically generate synonyms appropriate to the data



# ADDITIONAL DATA PREPARATION

- Remove "boilerplate" language common to most or all documents
  - Headers and footers
  - Common qualifiers
  - Disclaimers
- Parse created data
  - Convert abbreviations
  - Correct misspellings
- Use term frequency filtering to assist with the creation of a stop list

# ADDITIONAL DATA PREPARATION

- Recommendation: create subsets of documents by language. For example, all English documents in one corpus, all German documents in another corpus, etc.
- SAS includes extremely robust and sophisticated data manipulation capabilities, including character functions and regular expressions.



# TEXT MINING PROCESS

Text Preprocessing **Text Parsing** Transformation (Dimension Reduction) **Document Analysis** 

# TEXT TRANSFORMATION



- Also referred to as "Dimension Reduction"
- Transforms the quantitative representation into a compact and informative format
- Can also be used to further refine the data to be analyzed. For example, you
  can reduce the total number of parsed terms or documents that are analyzed.
- Eliminates extraneous information so that only the most valuable information or information that relates to a particular area of interest is considered.

### **TEXT FILTER NODE**



- Spell checking
- Concept Linking
- Full text search
- Define additional synonyms
- Sub-setting management of terms and documents that are passed to subsequent nodes

# DIMENSION REDUCTION TECHNIQUES

- Singular value decomposition (SVD)
- Roll up terms
- Combination of both approaches

# TEXT MINING PROCESS

**Text Preprocessing Text Parsing Transformation (Dimension** Reduction) **Document Analysis** 

# DOCUMENT CLUSTERING



- Expectation Maximization Clustering
  - Generates groups of similar documents from output of SVD
  - Fast clustering of many documents
- Hierarchical Clustering
  - Great for creating document taxonomies



Note: each document is assigned to a single cluster

 Optionally, use unsupervised data mining methods like self organizing maps or clustering after building text mining clusters, using the text mining cluster segment identifiers as inputs in the subsequent analysis

# DOCUMENT TOPICS

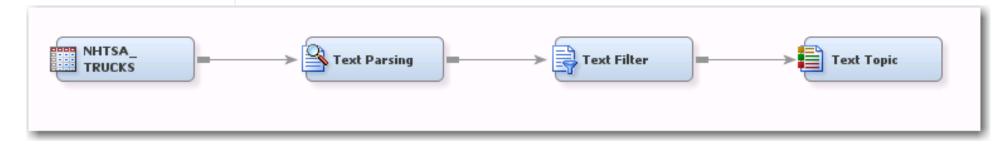


- Discovers topics in document collection
- Allows automatic creation of single and multi-word topics
- User defined topics and editing of automatic topics
- Multiple topics per document
  - Soft clustering using rotated SVD (PROC SVD followed by PROC FACTOR)

# SAS® TEXT MINER PROCESS



### **EXAMPLE** TEXT MINING PROCESS FLOWS

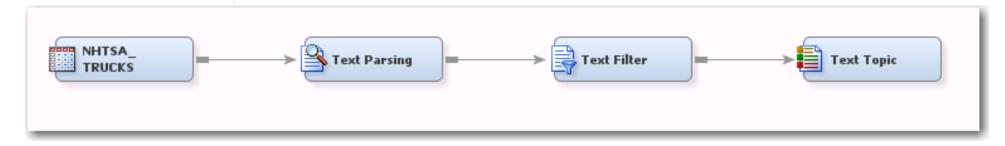




Start with a table that contains either:

- Documents saved as a variable (column)
- A column that points to physical text files

## **EXAMPLE** TEXT MINING PROCESS FLOWS



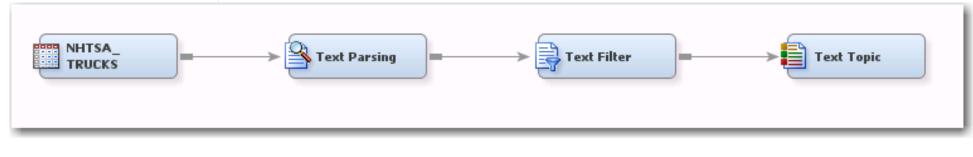


Apply natural language processing algorithms to **parse the documents** and **quantify information** about the terms in the corpus.

- Determine parts of speech (noun, verb, etc.)
- Perform stemming (run, runs, running, ran, etc.)
- Identify entities (names, places, etc.)



# EXAMPLE TEXT MINING PROCESS FLOW



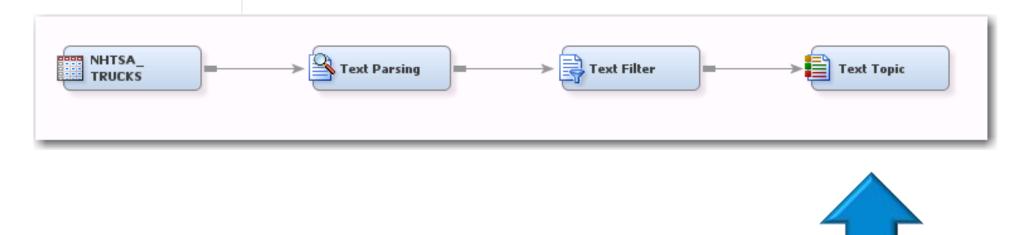


Optionally, filter the terms or documents that will be analyzed.

Can also perform spell-checking, full text searches, and analyze and view with Concept Linking

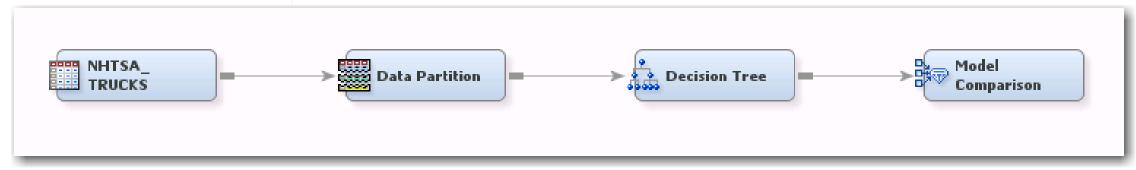


## **EXAMPLE** TEXT MINING PROCESS FLOWS



Analyze the documents to **create topics** and assign each document to one or more topics. In addition to derived topics, users can add their own topic definitions.

### **EXAMPLE** DATA AND TEXT MINING PROCESS FLOW



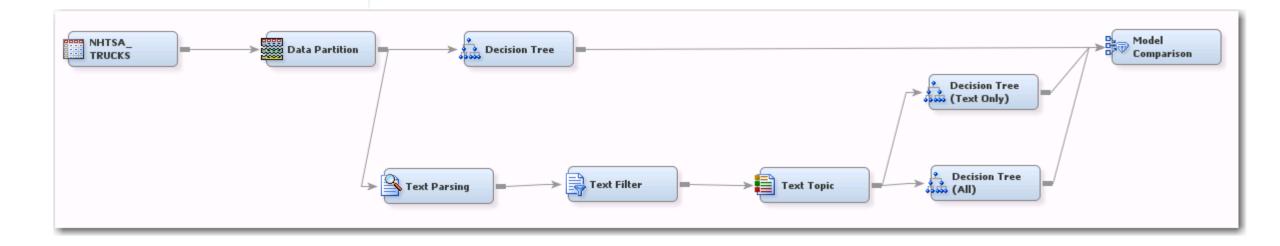
# **Mining Structured Data**



# Mining Unstructured (Text) Data



### **EXAMPLE** DATA AND TEXT MINING PROCESS FLOW



# Mining \*ALL\* Data: either Structured, or Unstructured or Both

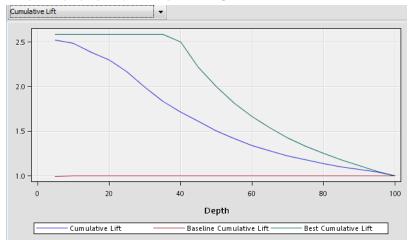


### **TEXT RULE BUILDER**



- The Text Rule Builder node generates an ordered set of rules that together are useful in describing and predicting a target variable.
- Each rule in the set is associated with a specific target category, consisting of a conjunction that indicates the presence or absence of one or a small subset of terms (for example, "term1" AND "term2" AND (NOT "term3")).
- A particular document matches this rule if and only if it contains at least one occurrence of term1 and of term2 but no occurrences of term3.

Target Value	True Positive/Total	Remaining Positive/Total	Rule	Estimated Precision	Sample Precision
1	130/133	1,137/2,946	accident	0.943884	0.97744
1	121/136	1,007/2,813	vehicle	0.860166	0.93308
1	32/33	886/2,677	mva	0.845067	0.93708
1	52/59	854/2,644	neck	0.814686	0.92797
1	29/33	802/2,585	neck	0.767854	0.92385
1	45/58	773/2,552	injury	0.718533	0.90486
1	68/91	728/2,494	shoulder & ~lift	0.680968	0.87845
1	37/48	660/2,403	car & ~door	0.649462	0.86971
1	10/11	623/2,355	drive	0.637703	0.87043
1	44/68	613/2,344	employee & fall	0.604525	0.84776



### **NEW NODE** TEXT PROFILE NODE

A tool providing a supervised approach to discovering and reporting the terms that best **profile** a set of documents associated with each level of a target variable.



- Uses a "new" procedure, Proc TMBelief, to determine the descriptive terms.
- Useful for binary, nominal, ordinal and date target variables.
- Internally we bin date variables to day, month, year etc. and map to ordinal.
- Note: User can bin interval target variables and then analyze as nominal or ordinal.

### **NEW NODE** TEXT PROFILE NODE

- How are men's and women's attitudes different toward my product?
- How has the answer to survey question #5 varied over the last 4 years?
- What is going on in the twitter feed over the last few months?
- Is there a difference in what people are talking about in different regions of the country?



# SAS® TEXT MINER™ DEMO



# SAS® TEXT MINER™ WHERE TO LEARN MORE



### FOR SELF-STUDY

- Visit
   http://support.sas.com/documentation/onlinedoc/txtmin er/index.html
- Download "Getting Started with SAS Text Miner" (How to Guide) (Available for multiple versions)
- Download "Getting Started Examples (Zip)"
- Work to complete the examples.



# SAS® TEXT MINER RESOURCES

### SAS Text Miner Product Web Site

http://www.sas.com/text-analytics/text-miner/index.html

SAS Text Miner Technical Support Web Site

http://support.sas.com/software/products/txtminer/index.html

SAS Text Miner Technical Forum (Join Today!)

**Data Mining and Text Mining Community** SAS Training

Data Miner Training Path:

http://support.sas.com/training/us/paths/dm.html

Courses for SAS® Text Miner:

https://support.sas.com/edu/prodcourses.html?code=TM&ctry=US



### YOUTUBE VIDEOS

- SASSoftware YouTube Channel
  - http://www.youtube.com/user/SASsoftware?feature=watch
- Manage All Unstructured Data with SAS® Text Analytics
  - http://www.youtube.com/watch?v=NHAq8jG4FX4&list=PL8B D07CC2C164FC40&index=4&feature=plpp\_video
- SAS® Text Analytics Software Demo
  - http://www.youtube.com/watch?v=I1rYdrRCZJ4&feature=BF a&list=PL8BD07CC2C164FC40



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Twitter: @Melodie\_Rush



QUESTIONS?
Thank you for your time and attention!



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