

SAS® GLOBAL FORUM 2017

April 2 – 5 | Orlando, FL

Advanced Analytics Approach for Recall in Automobile Industry

USERS PROGRAM

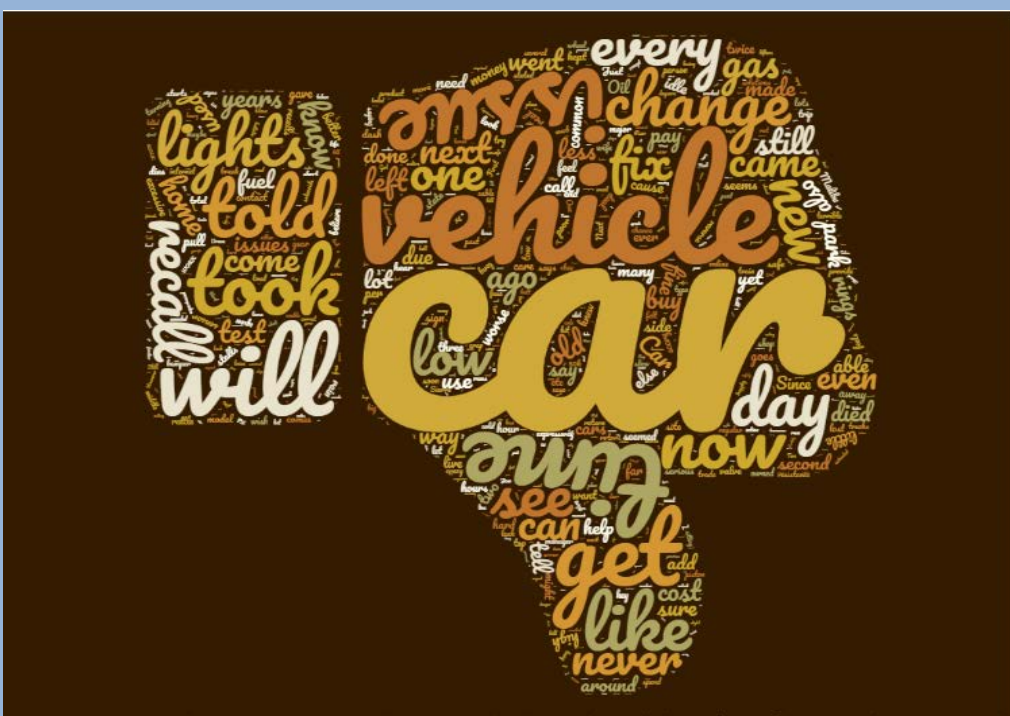
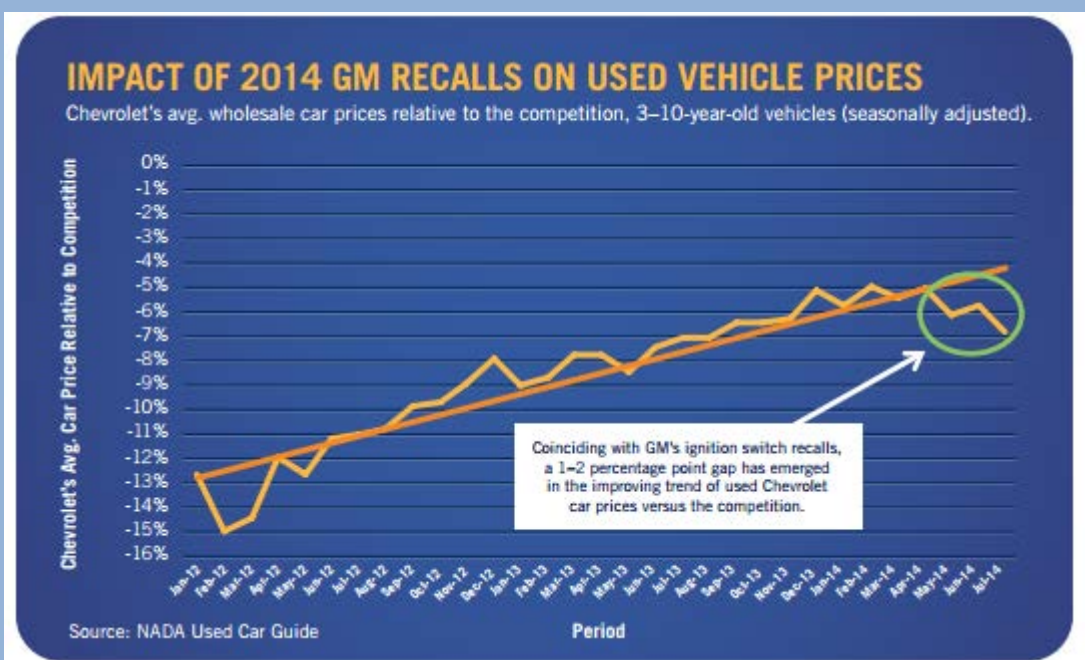
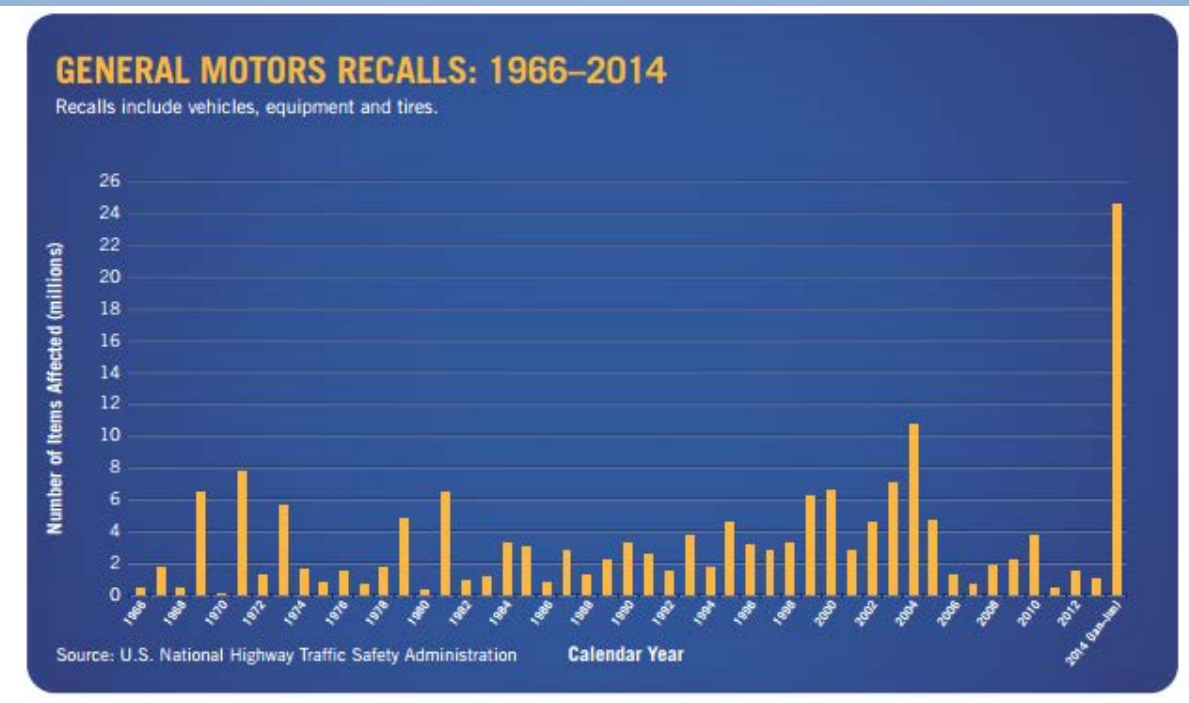


Advanced Analytics approach for recall in Automobile Industry

Prathap Maniyur, Prashant Nayak, Mansi Bhat

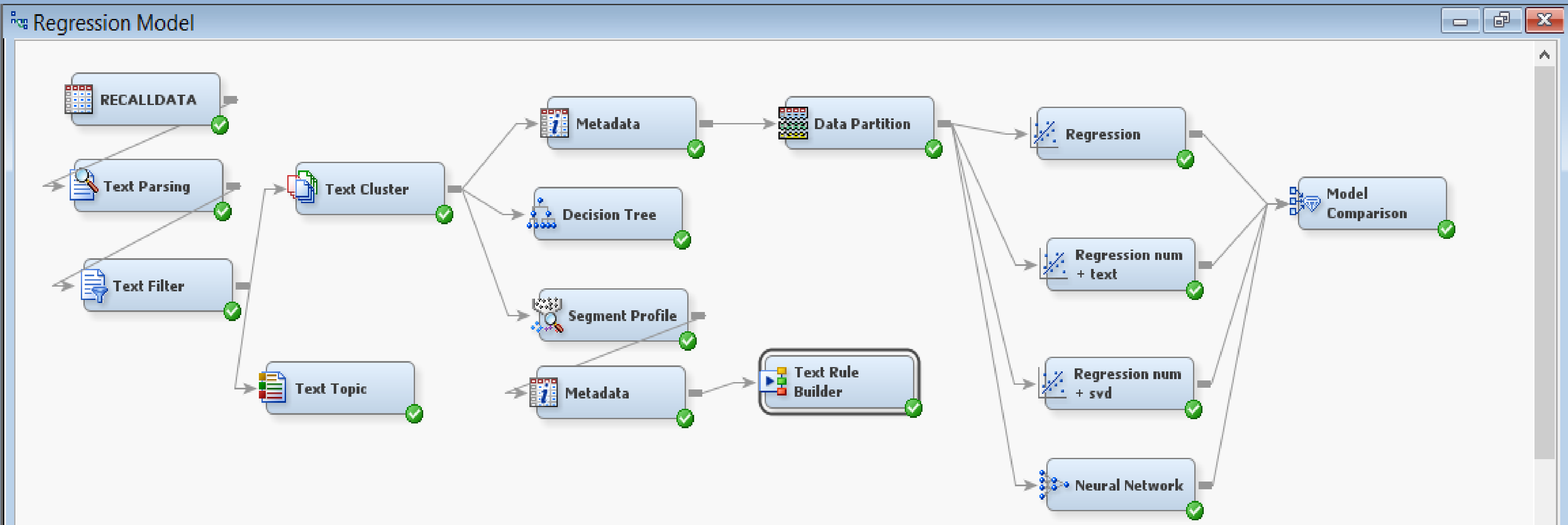
ABSTRACT

- GM had one of the biggest recalls in 2014 which involved cars such as Chevrolet Malibu, Chevrolet Impala, Monte Carlo, Cadillac CTS and Cadillac SRX³. In May 2014, GM had recalled more cars and trucks in 2014 than it sold in the last five years since it filed for bankruptcy. This paper is used to predict recall using social media analytics



METHODS

The websites and NHTSA complaints were scrapped to collect the complaints. The unstructured data such as customer comments or feedback can enhance the power of existing predictive models. The numeric data such as mileage of the cars where the issue is reported, severity issue rating and the cost incurred for the repair is captured in the numeric and the comments are captured as text. Wanted to analyse the predictive power of the text.



RESULTS

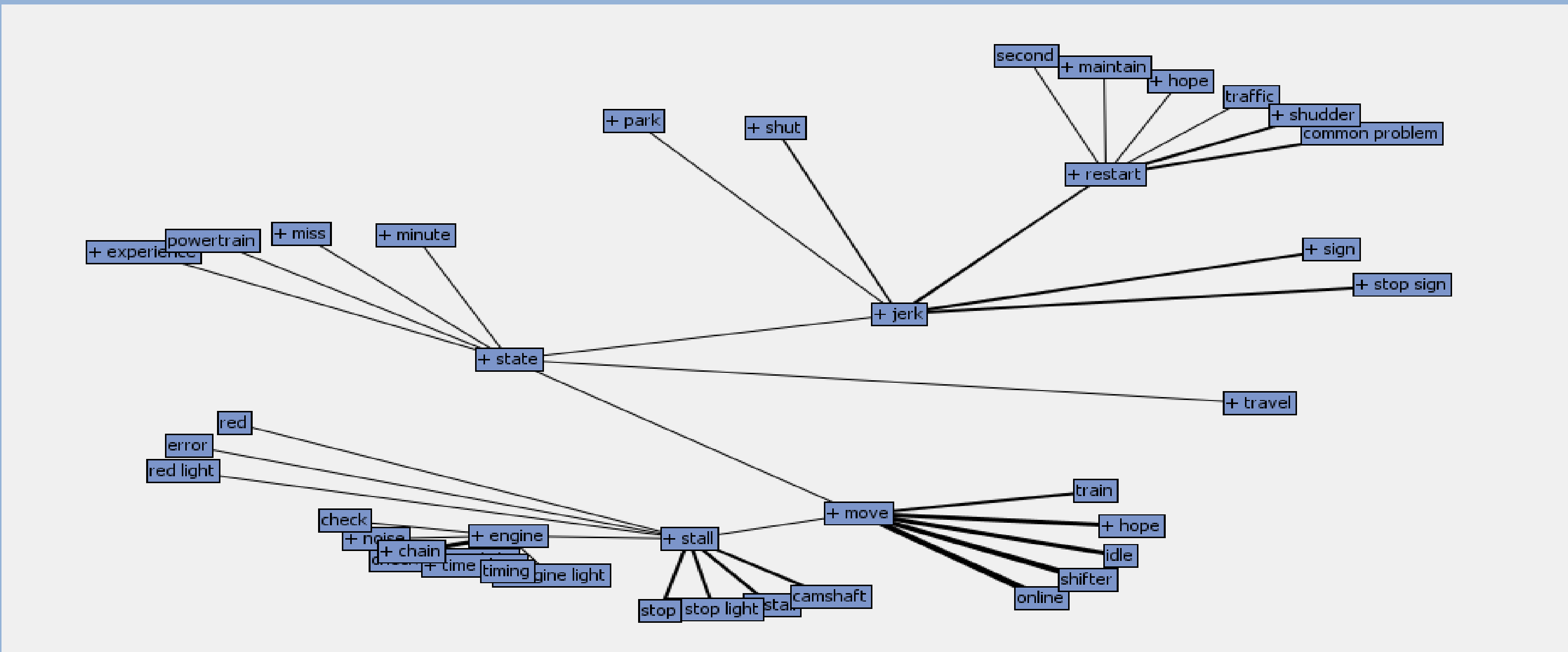
Clusters

Cluster analysis

Cluster ID	Descriptive Terms	Frequency	Percentage	Coordinate 1	Coordinate 2	Coordinate 3	Coordinate 4	Coordinate 5	Coordinate 6	Coordinate 7	Coordinate 8
1	+stall +car +happen back +recall +start chew +drive +day dealership	151	67%	0.65983	-0.15211	0.145051	-0.1006	0.066538	-0.09216	-0.0112	-0.03226
2	+headlight night sierra usa terrible +truck +side gmc +road +light	25	11%	0.224082	0.050501	-0.19071	-0.83074	-0.10537	0.251011	-0.27004	-0.05405
3	oil +quart +leak +mile consumption +add +check 'oil change' +buy low	51	22%	0.595979	0.493896	-0.04795	-0.06387	0.23995	0.10834	0.403463	-0.03617

Text topic analysis

Category	Topic ID	Document Cutoff	Term Cutoff ▼	Topic	Number of Terms	# Docs
Multiple	11	0.150	0.059 +tow, +warranty, +power, +die, home		54	14
Multiple	15	0.147	0.059 fuel, daughter, +intersection, traffic, +happen		60	19
Multiple	17	0.165	0.059 +stall, +restart, +cause, +power, +stall		51	16
Multiple	20	0.156	0.059 +shut, gmc, +steer, +back, +month		48	22
Multiple	21	0.150	0.059 +back, +park, +reverse, +hour, park		41	17
Multiple	23	0.155	0.059 +noise, +hour, +late, dept, +leave		49	30



Advanced Analytics approach for recall in Automobile Industry

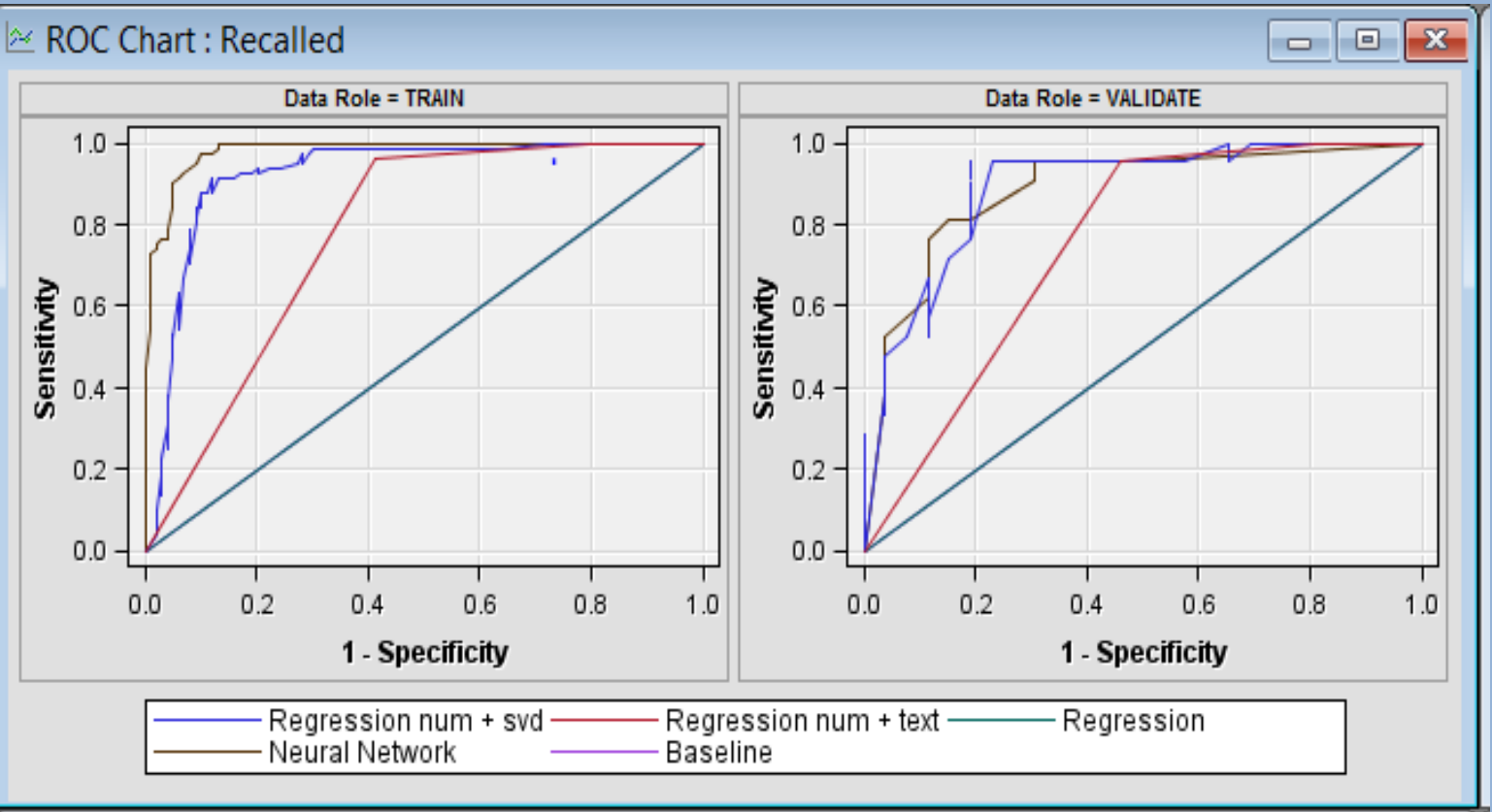
Prathap Maniyur, Prashant Nayak, Mansi Bhat

[RESULTS CONTINUED \(CLICK TO EDIT\)](#)

Text rule builder

Rules Obtained						
Target Value	Rule #	Rule	Precision	Recall	F1 score	True Positive/Total
1		1push		100.0%	8.82%	16.22% 9/9
1		2replace & ~usa & month		100.0%	22.55%	36.80% 15/15
1		3cable		100.0%	27.45%	43.08% 6/6
1		4daughter		97.22%	34.31%	50.72% 8/9
1		5car & ~usa & ~oil	86.81%	77.45%	81.87%	64/81
1		6stall	84.62%	86.27%	85.44%	30/42
0		7usa	100.0%	25.60%	40.76%	32/32
0		8quart & ~replace	100.0%	41.60%	58.76%	20/20
0		9oil & ~stall & ~leave & ~replace	98.65%	58.40%	73.37%	39/40
0		10night	96.51%	66.40%	78.67%	16/18
0		11night	86.67%	69.60%	80.93%	4/4

ROC analysis

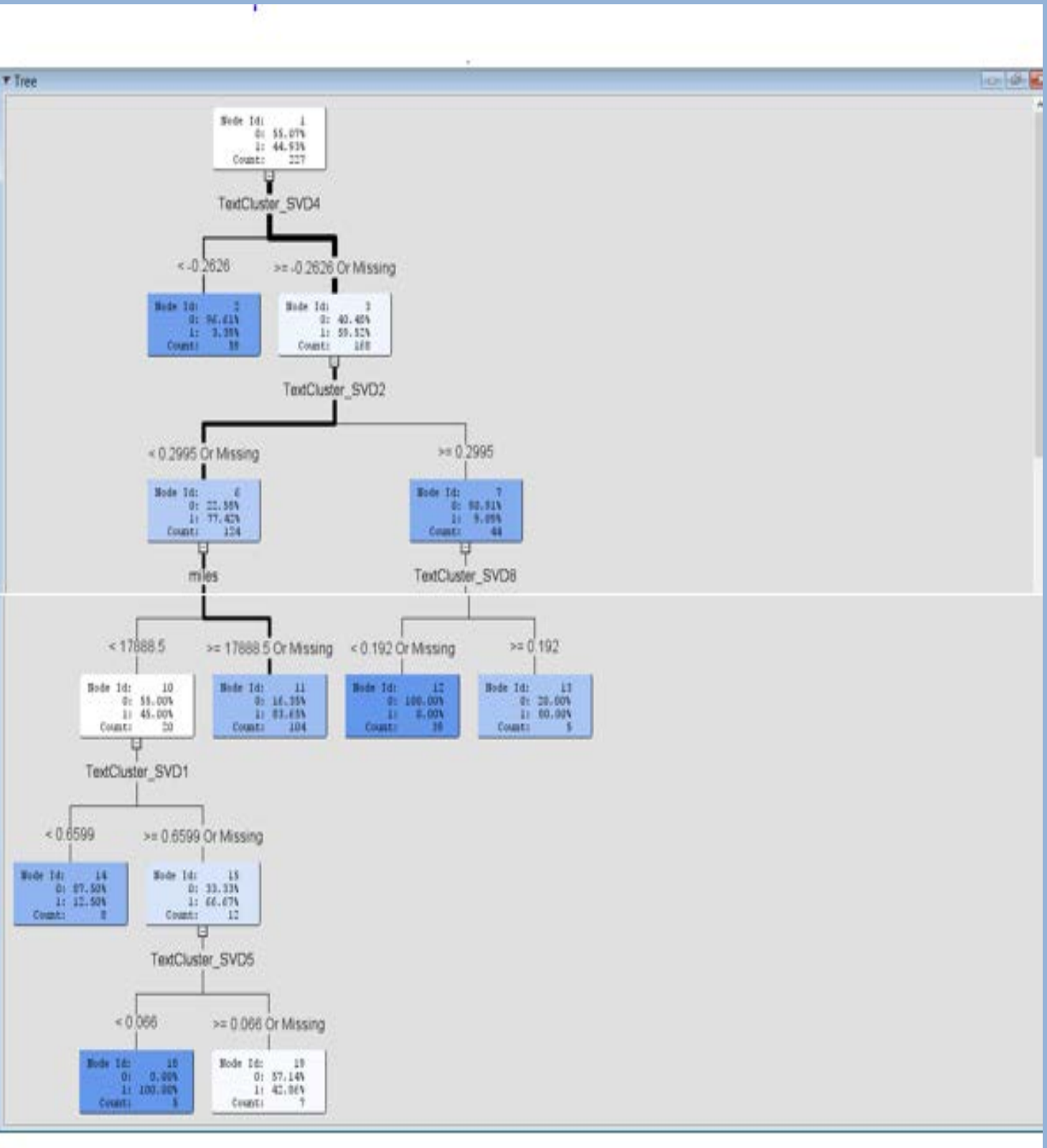


Regression Model Results

Event Classification Table								
Model Selection based on Valid: Misclassification Rate (_YMI3C_)								
Model	Data	Target	False	True	False	True	False	True
Node	Model Description	Role	Target	Label	Negative	Negative	Positive	Positive
Reg3	Regression num + svd	TRAIN	Recalled	Recalled	7	87	12	74
Reg3	Regression num + svd	VALIDATE	Recalled	Recalled	2	21	5	19
Reg2	Regression num + text	TRAIN	Recalled	Recalled	3	58	41	78
Reg2	Regression num + text	VALIDATE	Recalled	Recalled	1	14	12	20
Reg	Regression	TRAIN	Recalled	Recalled	81	99	0	0
Reg	Regression	VALIDATE	Recalled	Recalled	21	26	0	0
Neural	Neural Network	TRAIN	Recalled	Recalled	5	91	8	76
Neural	Neural Network	VALIDATE	Recalled	Recalled	4	22	4	17

[CONCLUSIONS](#)

The SVDs which are statistically significant are then sent to text builder to understand the key factors, decision tree is used to get the most important factors, text cluster and text topic which in turn can provide a final set of classification rules for both recall and no recall predictions in the total data for the GM cars.



Rules Obtained						
Target Value	Rule #	Rule	Precision	Recall	F1 score	True Positive/Total
1		1car & ~oil		100.0%	60.26%	75.21% 92/93
1		2start & ~oil		100.0%	68.21%	81.10% 47/47
1		3hour		100.0%	74.83%	85.61% 17/17
1		4stop		100.0%	78.81%	88.15% 17/17
1		5month		100.0%	83.44%	90.97% 34/36
1		6cold		100.0%	85.43%	92.14% 6/6
1		7hum		100.0%	87.42%	93.29% 11/11
1		8occur		100.0%	88.74%	94.04% 6/6
2		9headlight		100.0%	64.00%	78.05% 16/16
2		10usa & ~car & ~start		100.0%	80.00%	88.89% 15/15
3		11quant	88.66%	50.86%	50.86%	65.00% 26/29
3		12piston	91.89%	66.67%	66.67%	77.27% 10/12
3		13leak	92.86%	76.47%	76.47%	83.87% 8/8
3		14oil & oil	93.33%	82.25%	82.25%	87.50% 9/10
3		15oil & add	91.84%	88.24%	88.24%	90.00% 13/16
3		16consume	92.00%	90.20%	90.20%	91.09% 5/5

[REFERENCES](#)

1. National Highway Traffic Safety Administration(NHTSA) - Established by the Highway Safety Act of 1970 to achieve highest standards of excellence in motor vehicle and highway safety.
2. carcomplaints.com - It was started in year 2000 and was active in social media to get the buzz
3. kbb.com was launched in 1995 and is pioneer in automobile sector



SAS[®] GLOBAL FORUM 2017

April 2 – 5 | Orlando, FL