

## **Using Novel Data for Vehicle Rating**

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MEASURE, MANAGE, & REDUCE RISK

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### Outline

- 1. Vehicle Characteristics vs. Series
- 2. Collecting and attaching data
- 3. Developing and Implementing Models
- 4. Some illustrative results



### Vehicle Series

**Working Definition**: A vehicle series is an collection of vehicles that shares a number of characteristics in common and is used to aggregate loss experience.

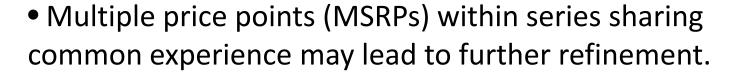
• Different companies or organizations will partition the universe of vehicles in different ways, so the specific set of series will be similar across organizations but not identical.





## Vehicle Series

- Common aggregations include:
  - Model year
  - Make
  - Model name
  - Additional attributes include:
    - Body Style &/or # of doors
    - # of drive wheels
    - Engine
    - Trim packages.







#### Vehicle Series

#### sounds simple but...:

- Model year (or range of model years).
  - ➤ When does the design change "significantly" enough to warrant a new series?
- Make (manufacturer).
  - > Chevy vs. GMC?
- Model name.
  - VW Jetta / GTI / Fox / Golf?
  - > Ford Escape vs. Mazda Tribute?
- Additional attributes, ...
  - Irrelevant alternatives?

...Credibility? ...



## **Vehicle Characteristics**

#### Alternate approach:

- Instead of defining a series, link the loss experience directly to the characteristics of the vehicle.
- Let a model *discover* the relationship between claims and the *relevant* aspects of a vehicle:

Model year	Price	Body style		
# of doors	# of cylinders	# of drive wheels		
Displacement	Horsepower	Torque		
ESC	ALB	DRL		
Curb weight	Wheelbase	etc.		



### Vehicle Characteristics

- ➤ When does the design change "significantly" enough to warrant a new series?
  - > When / as much as the characteristics do.
- > Chevy vs. GMC?
  - > The relevant differences are the characteristics, not the nameplate.
- > VW Jetta / GTI / Fox / Golf?
  - Design changes are considered, "branding" isn't.
- ➤ Ford Escape vs. Mazda Tribute?
  - > Share platform and common attributes, but some differences exist and are accounted for.
- > Irrelevant alternatives?
  - > Not significant in models.



# **Collecting Data**

In order to develop a model on vehicle characteristics, ... what data do we need?

- Exposures and Losses at the specific exposure level.
- Other relevant rating factors (covariates):
  - Other applicable elements of the rating plan (Territory, Driver, etc.)
- Some vehicle specific characteristics (e.g. price, year, body style, # of cylinders, # of doors, etc.)

#### What data do we want?

• As much detailed, relevant vehicle specific characteristic data as we can reasonably get our hands on.

#### Where does detailed vehicle data come from?

- A lot of hard work!
  - ...and multiple public and proprietary sources.



# Obtaining 3<sup>rd</sup> Party Data

#### **Outline**

- 1. Qualifying data sources
- Match keys
- 3. String matching tools
- 4. Level of aggregation
- 5. Process and QC



<sup>\*</sup> Thanks to Leila Mortazavi of ISO Innovative Analytics and the team.

# Qualifying Data Sources

- Is the data (potentially) predictive of losses?
- Is the data accurate? Can it be accurately matched?
- Completeness: does the data cover:
  - Adequate history (older model years)?
  - Adequately large proportion of insured vehicles?
- Will the data continue to be available in the future?
- Is the data allowable for use?
- Do you have (or can you obtain) appropriate rights of use?
- Does the data contain enough novel information to justify its cost (both the price and the time and effort to use it)?



# Match Keys

#### Some working definitions:

- "Base" dataset: containing exposures, losses, covariates and vehicle VIN for the specific risk.
  - The match keys should be *at least* as refined (disaggregated) as the 3<sup>rd</sup> party data.
- "3rd Party" dataset(s): Multiple sources.
  - Different match keys and levels of aggregation.
- *Ideally* (i.e. unrealistically) we would be able to match all of our 3<sup>rd</sup> party data to our base data by VIN or some common *decoded* VIN.
  - What follows is a discussion of what to do when the ideal situation doesn't hold.

## Match Key Cascade

Conceptually, the process of matching 3<sup>rd</sup> party data to the base can be thought of as hierarchical or a "cascade".

- 1. Model year
  - 2. Manufacturer (Make)
    - 3. Model Name
      - 4. Body Style
        - 5. Doors
          - 6. Drive Wheels
            - 7. Tie breakers (data source specific)
- ➤ If an exact match is found, then merge / join to base.
- ➤ If not, then roll up to next higher levels of hierarchy and resolve ambiguous cases.
- > Hierarchy may differ for various 3<sup>rd</sup> party sources.
- > Some pre-processing (clean-up) of keys helps a lot.



# Match Key Details

- 1. Model Year: matches are relatively easy
  - Some sources provide data in model year ranges (e.g. 2003-2007).
- 2. Manufacturer (Make): also relatively easy
  - Differences easily resolved (e.g. 'ACUR' ⇔ 'ACURA')
- **3. Model Name**: not easy at all a great deal of source specific detail and some idiosyncrasies.
  - Some sources have two fields (e.g. "model" and "sub model").
  - Model names in one source can be parsed to create tie breakers (or keys) with a defined field in another source e.g.:
    - Drive wheels: "4X4" vs. "4X2", "AWD"
    - Engine type: "TURBO", "HYBRID", "FLEX"
    - Engine cylinders or displacement: "(V6)", "(V8)" or "2.0", "3.2"
  - Other differences / idiosyncrasies not easily resolved.
    - Some tools to aid in matching or disambiguation of model names will be described in detail below.



# Match Key Details

- 4. Body Style ...
- 5. ...and doors: keep an eye out for differences

Base Data	3 <sup>rd</sup> Party Data					
<b>Body Style</b>	Body	Doors				
SEDAN 4D	SEDAN	4				
COUPE 2D	COUPE	2				
HCHBK 3D	HATCHBK	2				

- **6. Drive wheels**: '2' or ' 'vs. '4' (or 'AWD' or '6')
- 7. Tie Breakers:
  - Common fields that exist across the base and 3<sup>rd</sup> party source (or that can be parsed from name).
  - Will differ from source to source.
  - Sometimes measurements differ slightly among sources (rounding, definitions) – need to accommodate differences.



# **Matching Summary**

- "Cascade" approach automates the discovery of exact matches and allows efforts to focus on disambiguation.
- A lot of pre-processing of fields is required to align them.
- String matching tools can aid in the process:
  - Each function has different aspects (costs, features and options).
  - Use multiple functions, and resolve disagreement (special cases).
- There is still a large manual effort.
  - EDA (Exploratory Data Analysis), data queries (group by, unique, ...).
- Every different source requires unique solution details.
- The process needs to be replicable, in order to accommodate the introduction of new model years.



# Using 3<sup>rd</sup> Party Data Process and Quality Control

- Initial matching process is very large:
  - > 25 model years.
  - > 100K distinct vehicles.
- Annual updates need to be executed quickly.
  - About 4,000 distinct vehicle make / model / trims per year.
  - Some percentage are new model introductions, some models are significantly redesigned, and some features are added / introduced or made standard equipment.
- A robust process with built in QC is required for the production process.



# **Developing Models**

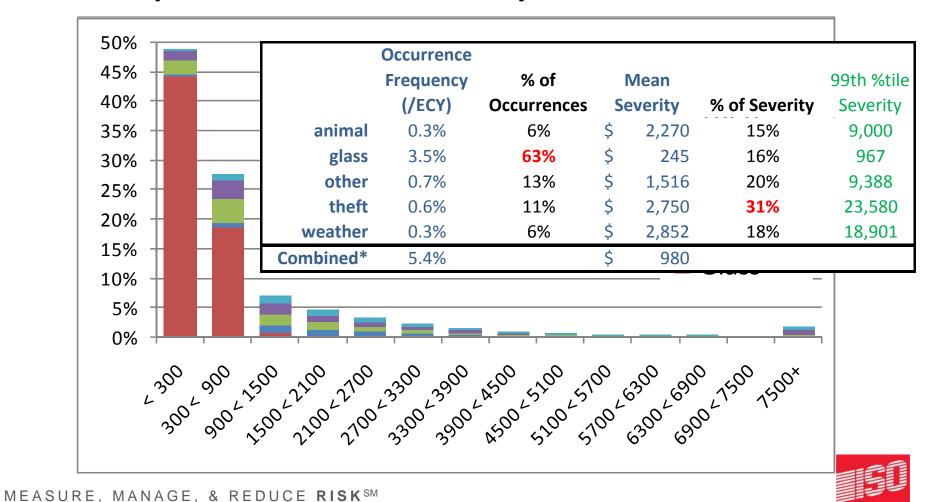
- When developing models from characteristics:
  - The variable selection task becomes challenging.
  - Need to adequately control for covariates (other elements of risk) like garaging address, driver, policy, etc.
  - Different characteristics may be associated with the likelihood (frequency) and the magnitude (severity) of losses, including antagonistic relationships (+/-).
  - Within a multi-peril coverage like comprehensive, different vehicle characteristics may be related to different perils.
    - The aspects of a vehicle that make it attractive to a thief may not matter to a deer.



# Comprehensive Perils

By Peril - Frequency and Severity Distributions

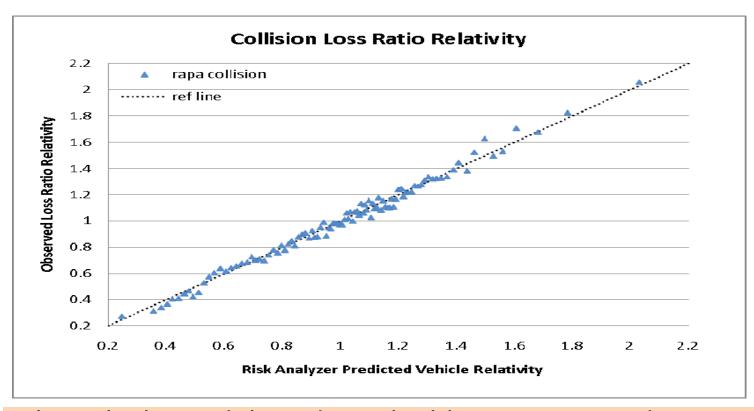
#### **Comprehensive Losses – Severity Distribution**



## Some illustrative results



# Collision Model Validation Predicted Vs Actual

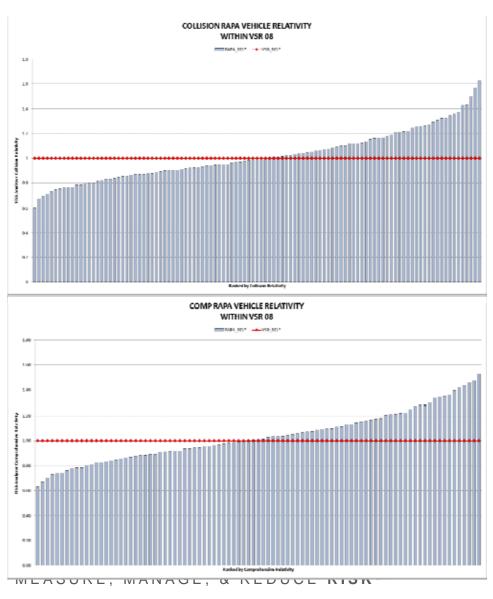


The vehicle model produces highly accurate predictions in line with the observed losses

Note: results against a holdout test dataset



## Segmentation within VSR SYMBOL 08



Collision Coverage

Comprehensive Coverage Predictive
Modeling using
Vehicle
Characteristics
provides
significant
segmentation
within VSR
Symbols



## Example 1:Differentiation within series

#### **2007 Ford Explorer Limited**

Selected Attributes			Syn	RA nbols/F		VSR Symbols/Relativitie s				
Cyl	Horse- power	Curb- weight	Price New	COL SYM	COL REL	COM SYM	COM REL	SYM	COL REL	COM REL
6	210	4615	\$34,070	LN	-	LJ	-	'12'	-	-
8	292	4615	\$35,365	LP	+2%	LT	+19%	'12'	Same	Same

RAPA

RAPA Vehicle Module is able to pick up differences among several different styles of a common line, and differentiate the risks.

The VSR Symbol Set sometimes groups different model trims within a series together under a common VSR symbol.



# Example 2: Performance Matters

#### 2007 Honda Accord

Selected Attributes			RAPA Symbols/Relativities				VSR Symbols/Relativities			
Model Trim	Horse power	Engine Size	Cyl	COL SYM	COL REL	COM SYM	COM REL	SYM	COL REL	COM REL
EX	166	2.4L	4	HU	-	HT	-	'13'	-	-
SE	244	3.0L	6	HV	+5%	HV	+7%	'13'	Same	Same

СОМР	➤The relativity for the EX model in RAPA is about 7% higher, compared to a 0% differential in VSR.
COLL	➤ The relativity increase for the EX model in RAPA is about 5%, compared to a 0% differential in VSR.



## Example 3: Redesigned Vehicle Series

#### **Toyota Camry 4-Door SE**

Selected Attributes			Syr	RA nbols/R	PA elativit	VSR Symbols/Relativities			
Model Year	Accel Rate	Price New	COL COL COM COM SYM REL SYM REL			SYM	COL REL	COM REL	
2006	Х	\$19,925	FR	-	FM	-	'11'	-	-
2007	1.6X	\$18,270	EW	+15%	ER	+8%	'10'	-5%	-9%

СОМР	➤ The 2007 redesign produces an 8% <i>increase</i> in relativity over the prior version in RAPA.  ➤ Contrast with a 9% <i>decrease</i> in relativity in VSR
COLL	<ul> <li>➤ The 2007 redesign produces an 15% <i>increase</i> in relativity over the prior version in RAPA.</li> <li>➤ Contrast with a 5% <i>decrease</i> in relativity in VSR</li> </ul>



## Summary

- Vehicle series rating and vehicle characteristic driven modeling
- Techniques and challenges: vehicle data for modeling and results
- Questions?

