## Case Study for Insurance Modeling

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### 1 Introduction

This case study will focus on pricing strategies for a commercial auto line of business. The goal of the study is to incorporate Machine Learning to improve pricing accuracy, while maintain explainability.

Throughout the life cycle of a insurance pricing project, there are in general six steps, and we will discuss in more detail how to use ML in each step:

- Scoping (Set up a goal for the project, success criteria and/or metrics, resource, cost, etc.)
- Data Preparation (Data sourcing, EDA, data cleaning)
- Feature Engineering (Transformation, variable selection)
- Modeling (Benchmarking, feature importance ranking, model, validation)
- Implementation (Deploy model for business end-users, testing)
- Monitoring (and prepare for model refreshing)

This case study will be focusing on Data Preparation, Modeling and Implementation. I will discuss about potential risks and how to manage them in the end.

### 2 Data Preparation

Put together both internal and external data sources, and prepare for modeling data.

#### Internal data source:

- Policy data: account/policy information, exposure, location, primary usage of vehicle, business industry, etc.
- Loss data: (Target, loss history). Loss linkage may be needed.
- Vehicle information: vehicle weight, vehicle age, vehicle cost, etc. May use external VIN decoding service if needed.

#### External data source:

- Driver information (credit history, police driving record, etc.)
- Credit history (for the business)

Things to do: - Explore missing values. - Check distribution

### 3 Feature Engineering

A lot of explorations could be done here. Include but not limited to:

- Encoding catergorical variables
- Explore feature interaction
- Bin numerical variables, and/or explore polinomial trend

## 4 Model Training and Validation

- \* Split Training, Test, Holdout (or Cross Validation)
- \* Selection of Target variables: Frequency and Severity VS. Pure Premium

Things to think about:

- Different coverages
- Outlier in pure premium

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•	Treed to deve	geot dot	Tancinative	19 00	use DOILEY	vear as	COMBINI	variable

- Need to trend
- \* AutoML for model Benchmarking
- \* AutoML for model variable importance (Shapeley Value)

Explain what is Shapeley value

Shapeley plot here

How to use Shapeley Plot?

- Feature importance: (explain feature importance ranking)
- Outlier detection
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# 5 Potential Risks and How to Manage Them

### 6 Conclusion