Motor Third-Party Liability Claims Analysis and Prediction

Yi-Pei Chan

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Motor Third-Party Liability Claims Analysis and Prediction

Yi-Pei Chan

Project Concept

Data Exploration The Dataset

Data Visualization

Model & Prediction

Model &

Model

Final Validation

Link to complete code and analysis: https://yipeichan.github.io/claims.html

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Data Exploration

The Dataset Data Visualization

Model & Prediction

Model &

Model

Final Validation

Project Concept

▶ Problem to solve :

How can we predict the number of claims a policyholder would file, given his age, his car brand, and so on?

- My approach to solve the problem :
 - 1. Explore the structure and properties of the dataset
 - 2. Choose the proper models to answer the question
- Methodology :

After exploring the data with visualizations,

- 1. Generalized Poisson Linear Model
- 2. Poisson Lasso Regression, Poisson Ridge Regression
- 3. Gradient Boosting Model
- Goals achieved by this project :
 - Explored relationships between the risk factors and ranked the influences of risk factors on claim numbers
 - 2. Investigated the efficacy of using modern machine learning algorithms to do P&C ratemaking
 - 3. Make your hiring decision easier!

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Project Concept

Data Exploration

Data Visualization

Model & Prediction

Poisson GLM

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Poisson GLM
Poisson Lasso & F
Gradient Boosting

Model

Final Validation



Data Exploration- The Dataset

- CASdatasets Package : Proposed by Christophe Dutang ¹ on OpenML
- ▶ Used in this study is freMTPL2freq dataset :
 - 1. Risk features were collected from motor third-part liability policies in France
 - 2. 678,013 samples, 12 explanatory variables

Variable Name	Description	Key
IDpol	Policy ID	(link with the claims dataset)
ClaimNb	Number of claims during the exposure period	
Exposure	Period of exposure (in years)	
VehPower	Power of the car	
VehAge	Vehicle age (in years)	
DrivAge	Driver age (in years)	
BonusMalus	Bonus/malus, between 50 and 350	<100: bonus; >100: malus in France
VehBrand	Car brand	Unknown categories
VehGas	Car gas	Diesel or regular
Area	Density value of the city where the car driver lives in	"A" for rural to "F" for urban centre
Density	Density of inhabitants of the city where the car driver lives in	Number of inhabitants per square-kilometer
Region	Policy region in France	

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Project Concept

Data Exploration

The Dataset

Data Visualization

Model & Prediction

Poisson GLM

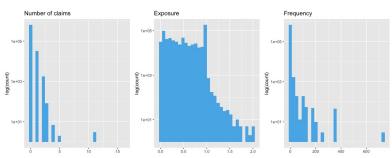
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Poisson Lasso & Ridge Gradient Boosting Model

Final Validation

- Among the 678,013 policies, there were 34,060 filed claims, i.e. 5.02% notified claims.
- Potential Problems :
 - Mean should equal to Variance in Poisson distribution
 ⇒ Use Negative binomial if Overdispersed
 - More 0s than are expected in Poisson regression?
 ⇒ Incorporate the logit model for predicting excess 0s
 - 3. Varied exposure periods (observations not comparable)
 - \Rightarrow Add offset of Exposure term to the model



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Project Concept

Data Exploration

The Dataset

Data Visualization

Model & Prediction

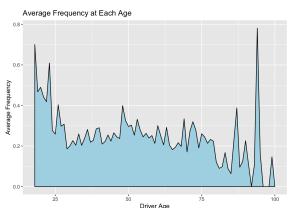
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Model & Prediction

Poisson GLM
Poisson Lasso & Ridge
Gradient Boosting

Final Validati

- Exposure : the duration of the insurance coverage
- ► Claim frequency : claim count per unit of exposure
- Did driver age influence frequency?
 - 1. The highest mean frequency happens at age 94
 - 2. Drivers between age 18 to 23 tends to have higher mean frequency



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Project Concept

Data Exploration

Data Visualization

Model & Prediction

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Model & Prediction

Poisson GLM
Poisson Lasso & Ridg
Gradient Boosting

Final Validation



Did vehicle brand and age influence frequency?



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Yi-Pei Chan

Project Concept

Data Exploration

The Dataset

Data Visualization

Model &

Prediction

Model &

Prediction

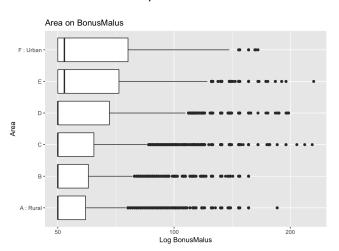
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Final Validation

Q & A

Model

▶ What is the relationship between area and bonus-malus?



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Project Concept

Data Exploration

The Dataset

Data Visualization

Model & Prediction

Poisson GI

Model &

Model & Prediction

Poisson GLM Poisson Lasso

Poisson Lasso & Ridge Gradient Boosting Model

Final Validation

Data Visualization

Model & Prediction

Poisson GLM

Prediction

Poisson GLM

Poisson Lasso & Ridge

Gradient Boosting

Model

O & A

Before training the models, randomly select 30% of the data and set aside as testing set to find the best fitting model

► Poisson GLM Model

```
glm(formula = ClaimNb ~ VehPower + VehAge + DrivAge + BonusMalus +
   VehBrand + VehGas + Density + Region + Area, family = "poisson",
   data = data[(data$data == "train"), ], offset = log(Exposure))
```

- Statistically significant variables (Signif. level 1%):
 VehAge, DrivAge, BonusMalus, VehPower, Density, etc.
- Overdispersion Test
 Overdispersion test

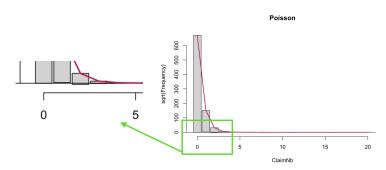
```
data: poissonglm z = 3.9191, p-value = 4.444e-05 alternative hypothesis: true alpha is greater than 0 sample estimates: alpha 0.243314
```

- 1. Small p-value :The test confirms the overdispersion
- 2. The alpha value very close to zero: Overdispersion may not be a serious concern here



Model and Prediction - Poisson GLM

► Hanging rootogram : Only 2 count is a little under predicted



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Data Exploration

The Dataset

Data Visualization

Model & Prediction

Poissor

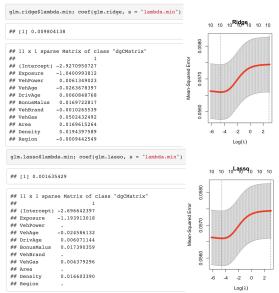
Model & Prediction

Poisson GLM

Poisson Lasso & Ridge Gradient Boosting Model

Final Validati

Model & Prediction - Poisson Lasso & Ridge Regression



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Project Concept

Data Exploration

Data Visualization

Model & Prediction

Poisson GLM

Model &

Poisson GLM

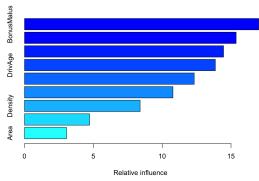
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Gradient Boosting Model

Final Validati



Model & Prediction - Gradient Boosting Model



rel.inf var BonusMalus 17,014808 Region 15.372979 VehAge 14.459134 DrivAge 13.862481 VehBrand 12.328304 VehPower 10.782009 Density 8.396521 VehGas 4.728894 Area 3.054871

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Project Concept

Data Exploration

Data Visualization

Model & Prediction

Model &

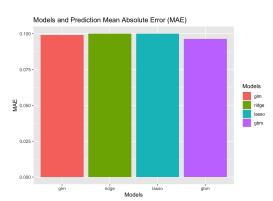
Gradient Boosting Model



Final Validation

Use the test set to find the best fitting model

- The claim number prediction MAE for test set with
 - 1. Poisson GLM: 0.09905573
 - 2. Poisson Ridge GLM: 0.09988506
 - 3. Poisson Lasso GLM: 0.09996999
 - 4. Gradient Boosting Model: 0.09630762



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Project Concept

Data Exploration

The Dataset

Data Visualization

Model & Prediction

Poisson GLM

Model &

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Poisson Lasso & Ridg Gradient Boosting Model

Final Validation



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Data Exploration

The Dataset

Data Visualization

Model & Prediction

Poisson GLM

Model &

Prediction

Poisson GLM

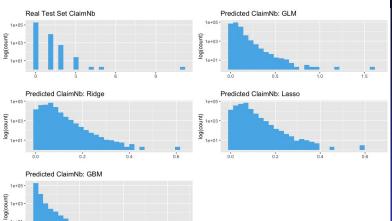
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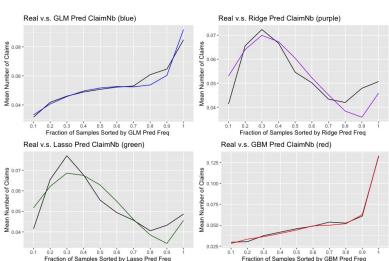
Final Validation

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08.4

Evaluation of the Predicted Number of Claims in the Test Set.





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Model &

Prediction

Model & Prediction

Poisson Lasso & Ridge Model

Final Validation





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Project Concept

Data Exploration

The Dataset

Data Visualization

Model & Prediction

Poisson

Model &

Prediction

Poisson GLM
Poisson Lasso & Ridge

Model

Final Validation