

Loss Cost Modeling vs. Frequency and Severity Modeling

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Purpose

- Outline an approach to compare accuracy and stability of Loss Cost Modeling vs. Frequency/Severity Modeling
- Execute this approach on a particular insurance loss dataset that includes large losses

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Procedure

- Attach a random digit (0-9) to the data
- Use the random digit to split the data into Training (0-3), Validation (4-7), and Holdout (8-9) datasets
 - Training Inclusion Criterion – Indication directionally consistent on each random digit
 - Validation Requirement – Remove indications not directionally consistent with Training
 - Holdout Dataset – Not used in developing structure, only for evaluating model results
- Fit main effects only, no interactions

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Evaluation of Results

- Accuracy
 - Offset Training indications on the Holdout dataset using a Tweedie Error Structure
 - Log-likelihood measure
 - Score the records in the Holdout dataset using Training indications
 - Compare the scored records to the actual Holdout loss experience
 - Lift Curve

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Evaluation of Results

- Stability
 - Use the Training Model structure (not the indications) to produce indications based on the Holdout dataset
 - Score the records in the Holdout dataset using the Holdout indications
 - Make a histogram of the percentage change in indication between Training indications and Holdout indications

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Models

- Loss Cost = Frequency * (Capped Severity + Propensity * Excess Severity)
- Loss Cost (1 Model)
- Frequency/Severity (4 Component Models)
 - Frequency = (Claim Count) / Exposure
 - Capped Severity = (Capped Loss) / (Claim Count)
 - Propensity = (Claim Count Exceeding Cap) / (Claim Count)
 - Excess Severity = (Total Loss - Capped Loss) / (Claim Count Exceeding Cap)

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Error Structures

- Loss Cost
 - Tweedie, p=1.6
- Frequency/Severity
 - Frequency - Overdispersed Poisson
 - Capped Severity - Gamma
 - Propensity - Binomial
 - Excess Severity - Gamma

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Extracting Indications

- Loss Cost
 - Relativities produced automatically
- Pure Frequency/Severity (2 models)
 - Multiply together relativities produced by each model
- Frequency/Severity/Propensity/Excess (4 models)
 - Create expected loss costs for each record based on indications, then fit a model to that data to produce loss cost indications by variable

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Included Variables

Variable	Loss Cost	Frequency	Clipped Severity	Propensity	Excess Severity
1					
2					
3					
4					
5					
6					
7					
8					
9					
10					
11					
12					
13					
14					
15					
16					
17					
18					
19					
20					
21					
22					
23					
24					
25					
26					
27					
28					
29					
30					
Total	10	20	10	10	5

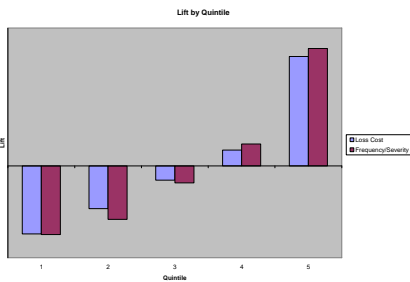
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Accuracy Results

- The Loss Cost model shows a better log-likelihood than the Frequency/Severity model

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Lift Curve



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Accuracy Results

- The Frequency/Severity model shows greater lift than the Loss Cost model

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Model Stability Comparison

Model Stability - Holdout Indications vs. Training Indications

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Stability Results

- The Frequency/Severity model shows greater stability than the Loss Cost model

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Loss Cost Pros

- On this particular dataset, the Loss Cost model resulted in a better log-likelihood measure of accuracy
- Requires only one model to build and maintain – may require less resource
- Simpler method to implement offsets
 - A modeled loss cost dataset can be produced for the Frequency/Severity modeling procedure to facilitate offsets, but this may require more work

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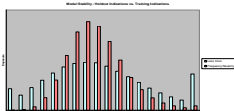
Frequency/Severity Pros

- On this particular dataset, the Frequency/Severity model showed greater lift and stability
- Potential for more focused treatment of large losses
- Potential for greater understanding of business

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Frequency/Severity Pros

- Potential reason for improved stability
 - More Flexibility
 - The Frequency/Severity approach gives the option of including some significant components (e.g. Frequency) of a certain variable in the model while excluding other insignificant components (e.g. Excess Severity) for that variable
 - The Loss Cost approach allows only a binary choice for the inclusion of a variable



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