

GLMS, MACHINE LEARNING, & MORE, OH MY!

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About the Presenters

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Agenda

- Introduction
- Residual Analysis Feature Engineering
- Machine Learning
- Understanding through Competing Models



Introduction

Netflix Prize







Netflix Prize

- Prize for best collaborative filtering algorithm to predict user ratings for films
- Competition began 10/2/2006 for Netflix, an online DVD-rental and video streaming service
- \$1,000,000 awarded in 9/21/2009 to BellKor's Pragmatic Chaos which bested Netflix's own algorithm for predicting ratings by 10.06%



Modeling Process Overview





Feature Engineering

- Feature engineering, refers to the process of creating new input features.
 - Feature engineering is an effective method of improving predictive models.



Feature Engineering, creating useful features

- Calculate statistics like the minimums, maximums, averages, medians and ranges.
 - Investigating the extremes (or the lack) may help define interesting behaviors.
- Create flags and count occurrences of events, highlighting statistically interesting habitual behaviors.
 - NSF notice on renewals for retention analysis
 - Examples: Younger drivers may separate themselves more clearly in a Non-Standard book than older drivers
- Create ratios seeking to add predictive value to already meaningful variables.
 - density, population/land area
 - vehicle to driver ratio (often used in a Matrix)



Feature Engineering, creating useful features

- Develop quintiles across variables of interest seeking to create expressive segments of the population while also dealing with extreme values.
 - Creating bins to transform variates to categorical variables
- Apply dimensionality reduction techniques, ranks, clustering etc. expecting that grouping those with similar behaviors will be statistically beneficial.
 - Principal Components
 - Clustering



Feature Engineering, creating useful features

- Consider the element of time as an important interaction with any feature that has been developed.
 - Recognizing newer policies are biased toward no claim history, opportunity to tap into external data
- Use regression to identify trends in continuous variables thinking that moves up or down (whether fast or slow) will be interesting.
 - Looking at univariate and doing a fit



- Have GLM
- Create residuals from you Actual and Predicted Values
- Model on the residuals using non-linear techniques
- Practical considerations concerning algorithm from the models
- Score your original GLM dataset
- Model with new variable either as a variate or could group as a categorical
 - Modeling techniques like decision trees naturally produce bins (categorical), bins have values so could also be treated as a variate
- Alternatively could do similar exercise on the actual for example vehicle symboling



- Initial GLM has been completed
 - Validates fairly well but may have room for improvement



• Model on the residuals using non-linear techniques

• Practical considerations concerning algorithm from the models

• Use developed non-linear model to score your original GLM dataset

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EMPUBLISHSCORE -
🔏 Program
🔚 Save 🔹 🕨 Run 🔹 🔲 Stop 🛛 Selected Server: SASApp 🔹 💥 🗍 Analyze Program 👻 🛛 Export 🔹 Send To 👻 Create 👻 🔚 Properties
    ****** DECISION TREE SCORING CODE ******;
  2
    3
  4
    ****** LENGTHS OF NEW CHARACTER VARIABLES ******;
  5
  6
    LENGTH WARN $ 4;
    ****** LABELS FOR NEW VARIABLES
                                          *****;
  8
  9
    LABEL _NODE_ = 'Node';
LABEL _LEAF_ = 'Leaf';
    LABEL _NODE = 'Node';
 10
```


• Model with new variable either as a variate or could group as a categorical

Questions

Thank You for Your Time and Attention

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