

## Comparing Machine Learning and Conventional Statistical Techniques in Claims Models

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## Agenda

- Claims Triage Models
- Beyond GLM: Regression Models
- Beyond GLM: Machine Learning models
- Case Study
- Conclusions

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## Claims Triage Model

Objectives:

- Identify claims that have the potentials to blow up
- Allocate claims resources efficiently
- Manage claims to mitigate the loss

Challenges for model development:

- Time stamps of variables
- Empirical loss distributions are different from exponential family
- Many possible interactions
- Unobserved heterogeneity
- Model validation and benefit quantification

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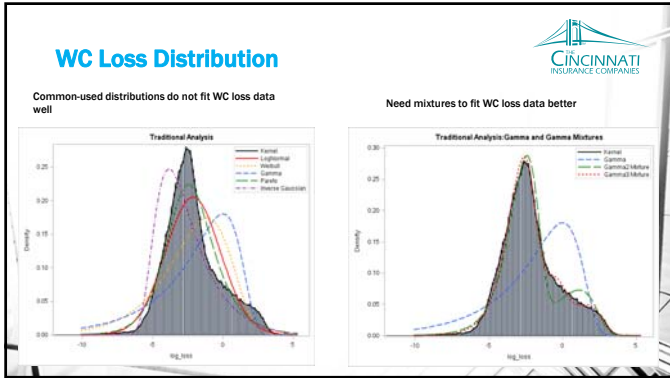
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- ### Claims Triage Models: Beyond GLM
- Regression Models
    - Finite mixture models, double GLM, quantile regressions
    - Generalized mixed model (random and fixed effects)
    - Elastic nets, LASSO, Ridge
  - Machine Learning Models
    - Support Vector Machine
    - Neural Networks
    - Random Forests
    - Gradient Boosted Trees (GBM)

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### Claims Triage Models: Beyond GLM

Case study to appear at the presentation.

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**Questions?**

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**Mixed Effects Model Resources**

General Sources

- Predictive Modeling Application in Actuarial Science Vol. 1 (Chapters 7, 8, 9, and 16):  
<http://instruction.bus.wisc.edu/jfrees/jfreesbooks/PredictiveModelingVol1/index.htm>
- [http://www.ats.ucla.edu/stat/mult\\_pkg/gimm.htm](http://www.ats.ucla.edu/stat/mult_pkg/gimm.htm)

Mixed Models in R

- <https://cran.r-project.org/web/packages/lme4/vignettes/lmer.pdf>
- <http://lme4.r-forge.r-project.org/IMMwR/lrgprt.pdf>
- <http://rpsychologist.com/r-guide-longitudinal-lme-lmer>

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## Elastic Net Resources

General Resources

- An Introduction to Statistical Learning (Ch. 6): <http://www-bcf.usc.edu/~gareth/ISL/>
- The Elements of Statistical Learning (Ch. 3): <http://statweb.stanford.edu/~tibs/ElemStatLearn/>

Fitting Elastic Nets in R

- [http://web.stanford.edu/~hastie/glmnet/glmnet\\_alpha.html](http://web.stanford.edu/~hastie/glmnet/glmnet_alpha.html)

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## Neural Network Resources

General Resources:

- The Elements of Statistical Learning (Ch. 3): <http://statweb.stanford.edu/~tibs/ElemStatLearn/>
- [http://ml.eng.cam.ac.uk/~mfg/local/110/110\\_hand4.pdf](http://ml.eng.cam.ac.uk/~mfg/local/110/110_hand4.pdf)
- <http://neuralnetworksanddeeplearning.com/chap4.html>

Neural Networks in R

- [https://journal.r-project.org/archive/2010-1/RJournal\\_2010-1\\_Guenter+Fritsch.pdf](https://journal.r-project.org/archive/2010-1/RJournal_2010-1_Guenter+Fritsch.pdf)
- <http://datascienceplus.com/fitting-neural-network-in-r/>
- <http://bloomberg.com/2015/05/13/neural-networks-using/>

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## Decision Tree Resources

- Gradient Boosting:
  - Greedy Function Approximation: A Gradient Boosting Machine: <https://statweb.stanford.edu/~jhf/ftp/trebst.pdf>
  - <https://cran.r-project.org/web/packages/gbm/gbm.pdf>
- XGBoost (extreme gradient boosting)
  - <https://xgboost.readthedocs.io/en/latest/>
  - <https://cran.r-project.org/web/packages/xgboost/xgboost.pdf>
- Random Forest
  - Random Forests: <https://www.stat.berkeley.edu/~breiman/randomforest2001.pdf>
  - <https://cran.r-project.org/web/packages/randomForest/randomForest.pdf>

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