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Predictive Modeling for Commercial Lines with **Schedule** Rating

Larry Seymour FCAS, MAAA

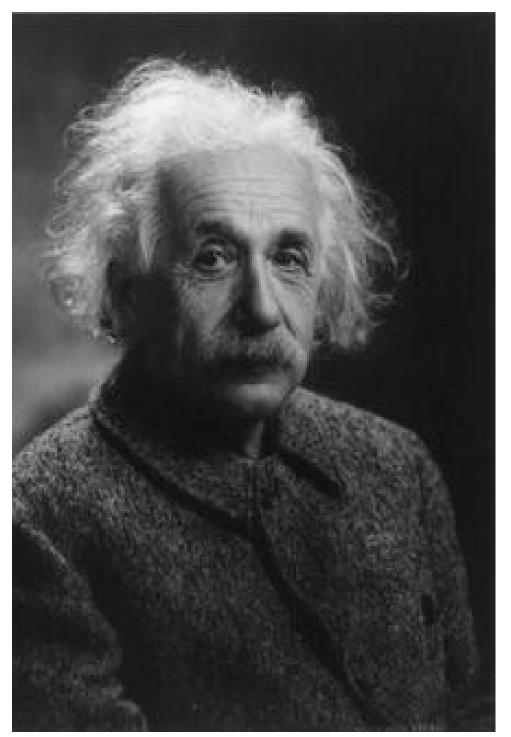


What is predictive modeling?

A process which analyzes historical experience to predict future behavior

Involves:

- Time, \$\$\$, Data, Management buy in.



What is Experience Rating?

• A method used to adjust premiums up or down based on an account's loss history.

[Ap + WAe + (1-W)Ee + B] / [Ep + WEe + (1-W)Ee + B]

ISO CGL formula

What is Schedule Rating?

A provision to adjust premiums or
 to reflect characteristics of a risk which are not otherwise reflected in the premium.

ISO GL Schedule Rating Modifications:

		Mod Range	
Risk			
Characteristic	Description	Credit	Debit
Location	Exposure inside premises	5%	5%
Location	Exposure outside premises	5%	5%
Premises	Condition and care of premises	10%	10%
Equipment	Type, condition and care of equipment	10%	10%
Classification	Peculiarities of classification	10%	10%
	Selection, training, supervision,		
Employees	experience	6%	6%
Cooperation	Medical Facilities	2%	2%
Cooperation	Safety Program	2%	2%
Maximum Credit or Debit (varies by state)>			25%

Do these three ideas conflict, collide or corroborate ?

Will modeling replace the rating bureaus?
Are we creating our own perfect storm?
Get a grip - it's only insurance!

Pricing Assumptions

- Rating bureaus are a decent starting point
- Experience rating is appropriate
- Your predictive model is appropriate

Two areas of thought for implementing a model:

 The model is the price (i.e. a filed u/w scoring model), now go sell it

• The model is another U/W tool in the tool box (i.e. an u/w guideline)

Case Study – Underwriter Variance

Hotel/Motel, >30 units, w/o Cooking, \$1M/\$2M General Liability, Standard Premium = \$10,000.

Underwriter:



Written Premium:

- \$6,000 (BOP, preferred pricing)
- \$7,500 (Package, preferred pricing)
- \$7,500 (Package, preferred pricing)
- \$8,100 (Package, preferred pricing)
- \$10,000 (Package, standard pricing)
- \$10,000 (Package, standard pricing)

Various Observations:

- 1. Hail risk
- 2. Crime statistics
- 3. Historical restaurant exposure verify class code
- 4. Age of building (some say old some say new)
- 5. Area specific rates higher than average
- 6. Tripadvisor.com (credibility issue)
- 7. Pool/Sauna exposure

BATES

What did a predictive underwriting model say?



- \$11,000 (10% debit)
- Ignores current market pressure
- Remembers hotel loss experience

Differences between Models and Schedule Rating



Models



- Consistent treatment
- Works 24 / 7



Models



- Dwells on the past
- Removes the "art" in U/W

Schedule Rating

- Provides flexibility
- Can follow the market

Schedule Rating

- Varies by underwriter
- Allows for the "art" in U/W

Can a predictive model capture all of the risk characteristics of a risk?



• For each individual risk, ?????



Predictive Modeling for Commercial Lines with Schedule Rating

Benefits and Methods

Alex Laurie, FCAS, MAAA

March 11, 2009



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Predictive Modeling for Commercial Lines with Schedule Rating

- Benefits of predictive modeling
- Leveraging underwriting knowledge
- Handling heterogeneity
- Model result formats
- Implementation methods

Consistency of Process

- Underwriting Judgment can vary:
 - From person to person
 - From policy to policy



- Consistency of Process
 - Predictive model:
 - Looks at same characteristics each time
 - Facilitates consideration of interactions among factors
 - Gives the same result for two risks with the same characteristics

Efficiently considers <u>more</u> factors than manual process



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- More precise and consistent evaluation of loss drivers can yield:
 - Increased profit
 - Expanded underwriting appetite



Benefits for Commercial Lines?

- Does predictive modeling offer more lift for personal lines or commercial lines?
 - This depends on:
 - Sophistication of current pricing/underwriting
 - Additional data sources available (both internal and external)
 - Ability to model patterns in data
 - Ability to implement model into processes
 - In some cases, this is more powerful for commercial lines, owing to greater variation in individual risk quality within a rating class.

For First Generation Predictive Models:

- Talk to underwriters
- What do underwriters consider important for assessing risk?
- How can these things be sourced and quantified?
- Do proxies exist for these risk characteristics?



- What types of internal and external data I can use?
 - Data related to:
 - Geography / Demography
 - Industry of risk
 - Individual Enterprise (including financial and prior claim data)
 - Agent / Broker / Producer
 - Other products purchased by risk
 - Competitor rating plans

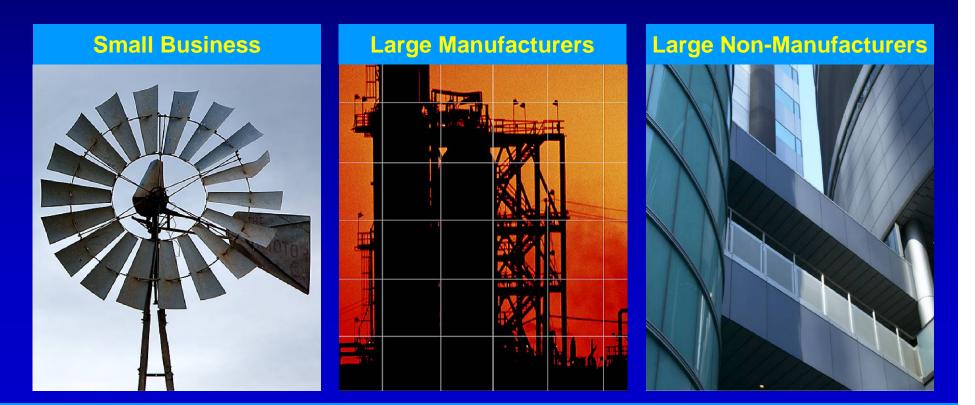
- For Subsequent Generations of Predictive Models, in addition to quantitative model monitoring:
 - Ask underwriters about when the models disagree with their judgment
 - Use system to record underwriters' observations & concerns (e.g. "help line" or "e-mail box")
 - Anecdotes may improve models via:
 - New variables or
 - Different treatment of existing variables

Modeling and Implementation Considerations

- Handling heterogeneity
- Model result formats
- Implementation methods

How do I deal with heterogeneous risks?

Segmented Models – To the extent data is credible, separate models can be built for specific types of business.



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How do I deal with heterogeneous risks?

- Interaction Variables Allow the effects of one variable to differ based on another variable.
 - Example: If state annual snowfall > 5 inches, use variable based on type of snow/ice removal equipment.
 - Example: Interact number of employees with employee turnover.

Relative Risk						
# of	Employee Turnover					
Employees	Low	Medium	High			
Under 50	0.8	1.0	1.2			
50 – 500	0.7	1.1	1.4			
Over 500	0.7	1.2	1.6			
Illustrative Data						

What does model result look like?

- Model indication of risk:
 - One of 10 tiers
 - A three-digit score
 - A suggested schedule modifier
 - A separate indicated rating factor

One of 10 Tiers

- Pros:
 - Simple to understand (e.g., tier 10 = highest expected loss costs)
 - Can choose desired number of tiers
 - Can adjust definitions of tiers over time
- Con:
 - Adjusting definitions can cause confusion (e.g., "Why does Tier 1 mean a different level of risk this year than it did last year?")

Suggested Schedule Modifier

- Pro:
 - Easy to interpret
- Con:
 - If model captures different characteristics from schedule rating plan, it may not make sense for model to affect schedule rating

Three-digit Score

Pros:

- Can create mapping of scores to rate factors (e.g., score of 400 yields 1.20 rate factor)
- Can set minimum eligibility scores for different products or companies (underwriting)
- Can adjust rate factor mappings or cut-off scores over time, without changing meaning of scores
- Con:
 - Results are more complex than an indicated tier

Separate Indicated Rating Factor

- Pro:
 - Little or no judgment required
- Cons:
 - Allows underwriter no control over how model result is used
 - May introduce competitive disadvantage

What does model result look like?

- Reason codes:
 - What variables, or classes of variables, most influenced the model's indication?

Should use of model result be mandatory?
 Implementation Methods:
 Model result...

		is used as	
is optional tool to guide schedule	is required input to	separate rating variable from	replaces
mod selection	schedule mod	schedule mod	schedule mod

Increasing Reliance on Model

Optional Tool to Guide Schedule Mod Selection

- Pros:
 - Most conservative option
 - Most similar to current processes
 - Gives underwriters maximum discretion
- Cons:
 - Underwriters may ignore model
 - May not reduce manual work

Required Input to Schedule Mod

- Pros:
 - Ensures model affects pricing and/or underwriting decision
 - Still allows underwriting judgment
 - Forces collection of data if model use required
- Cons:
 - Depending on design, model may only affect some of the risks
 - May restrict underwriting from applying judgment in areas where their judgment works

Hovsepuldencedelgoeareanuired input to schedule and?

 Makes sense when model and schedule rating seems to identify different characteristics
 Underwriter sannot deviate more than X% from the modified there doel gives
 Underwriter rational schedule gives
 Underwriter rational schedule schedule rating on three of the ioinginal six characteristics
 Underwriter cannot give schedule rating discount to a fish of the soft schedule rating discount to a fish of the soft schedule rating discount to a fish of the worst three tiers

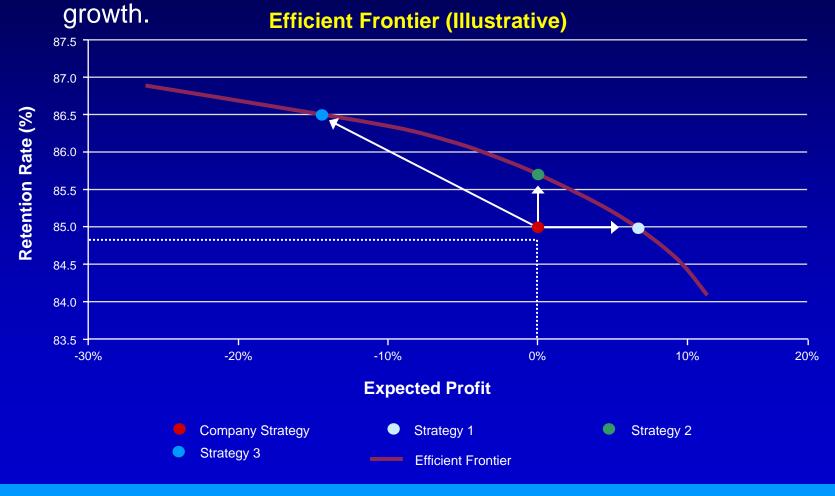
Pro:

Eliminates need for manual Key Point: Knowdwaentingodel works better than schedule rating and vice versa. Design fin process plan accordingly. (Helpful measurement techniques to Schedule rating has been follow!) ineffective, or

> Schedule rating will have little value once predictive model is implemented

Price Optimization

Price Optimization can leverage information about the competitive landscape and demand for the product to increase profit and/or



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Predictive Modeling for Commercial Lines with Schedule Rating

> Adam Sherwin March 11, 2009

Topics

General Measurement Issues
Quality of Current Underwriting
Quality of the Predictive model
Overlap Model and Underwriting
Combining
Determining Value

All tests of model should be on a hold-out

Random hold-out

- Are all observations really independent?
- Catastrophes
- Lawsuits
- Future hold-out
 - Cyclicality/Inter-period correlation
 - Unbiased
 - Worse than contemporaneous
 - Guide to degradation
 - Different risk

Interaction with experience

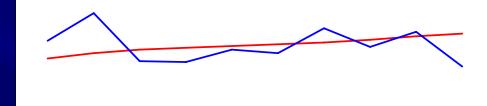
 Ignore: Use manual LR's with schedule or model—Not recommended
 Adjust each for experience
 Solve experience in model (Predictive Modeling for Commercial Lines with Schedule and Experience Rating)

How good is the current UW?

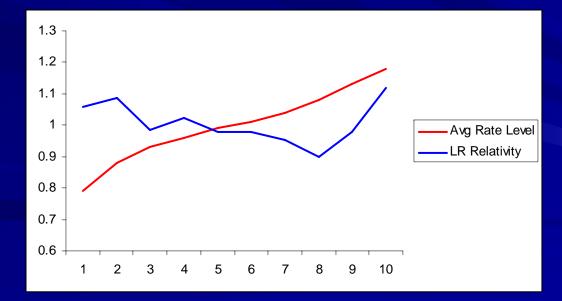
Some Variable line up well...

Others may not...

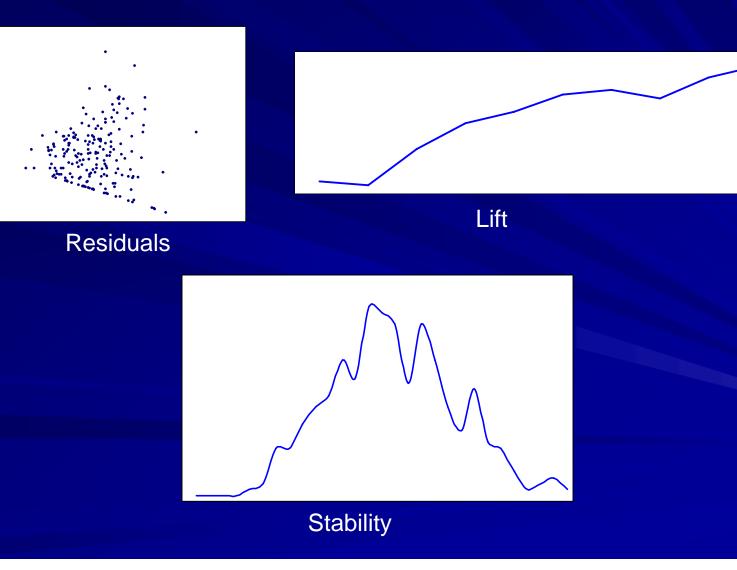




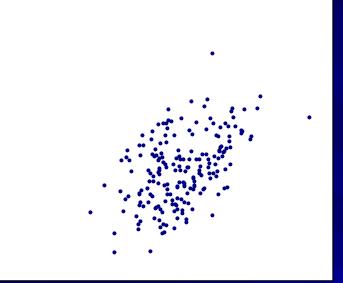
Overall results are somewhat mixed...



Does the predictive model work?

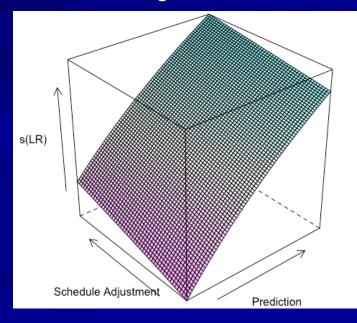


Comparison with Underwriting



Correlation

Both Predictive with predictive model stronger:



How to Combine

Blended

Simultaneous: LR=Var1+Var2+Var3+...+a*ScheduleModifier

Post-hoc: LR=a*PredictiveModel+b*ScheduleModfifier

Residual

- LR/Prediction=a*ScheduleModifier
- LR/(1-ScheduleModifier)=a*Prediction
- Full Combination
 - LR=Var1+Var2+Var3+Adj1+Adj2+Adj3
 - Change ranges or convert to score?
- Does it differ by segment?
 - Classic interaction test
 - UW characteristics can segment too

Are all Underwriters Equal?

Quality varies

- Education
- Experience
- Subjective Evaluations
- Beware of confounding variables
 - Line
 - Region
 - Business Type

Valuing the Model

Short-term:

- Maximize Profit: (Selected/Indicated-1)*Retention(Selected,Current)
- Value:

Sum(Premium*Change/IndicatedChange*Retention(Selected,Current)/R etention(Current,Current))

Long-term:

- Adverse selection on competitors
- Avoidance of adverse selection
- Assessed by Simulation

Underwriting Optimization

Value of UW Input > Cost of UW

- Value may vary by segment
- Value may vary by underwriter
- E.g. UW if f(Vars)*f(UW)*Premium>k
- Many variations:
 - If abs(Cost-Value)<k quick evaluation
 - Manual flagging
 - Automatic exceptions