Going from Demand Modeling to Price Optimization

Demand Modeling to Price Optimization

Presented by:

Lee M. Bowron, ACAS, MAAA Kerper and Bowron LLC Birmingham, AL www.kerper-bowron.com

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What is an insurance CEO's number one concern?

From Dow Jones:

Insurance executives say insurance pricing weakness is the biggest risk their industry faces in the next few years.

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What does the customer think?



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The Renewal Question

- The more tenure, the better the loss ratio
- But switching can be hard, tenure = value
- Most companies will ignore renewals or not give the full actuarial discount – is that the optimal treatment?

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Demand Modeling

- •Given a quote, will we convert?
- Start getting the data now (the ether of the renewal offer)
- Different for new business and renewals
- A key variable is the amount of rate change as well as the tenure of the policy

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GLMs can be used to model demand

Logistic regression analyzes binomially distributed data of the form

 $Y_i \sim B(p_i, n_i), \text{ for } i = 1, \ldots, m,$

where the numbers of Bernoulli trials *ni* are known and the probabilities of success *pi* are unknown. An example of this distribution is the fraction of flowers (*pi*) that germinate after *ni* are planted.

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Demand Models (Continued)

The model is then that for each trial (value of *i*) there is a set of explanatory/independent variables that might inform the final probability. These explanatory variables can be thought of as being in a *k* vector *Xi* and the model then takes the form

$$p_i = \mathbb{E}\left(\frac{Y_i}{n_i}\middle|X_i\right).$$

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Demand Models (Continued)

The logits of the unknown binomial probabilities (*i.e.*, the logarithms of the odds) are modeled as a linear function of the Xi.*

$$logit(p_i) = ln\left(\frac{p_i}{1-p_i}\right) = \beta_1 x_{1,i} + \dots + \beta_k x_{k,i}.$$

Note: there are other ways to analyze demand, but make sure you are doing it in a statistically significant manner.

* Logistic regression is from Wikipedia

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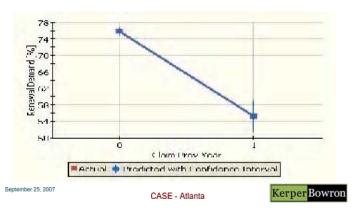
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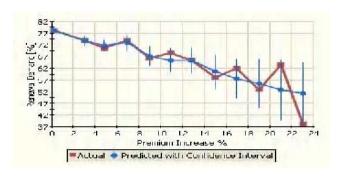
Effect of Previous Claim on Renewal Demand



EΛRNIX

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Effect of Premium Increase on Renewal Demand



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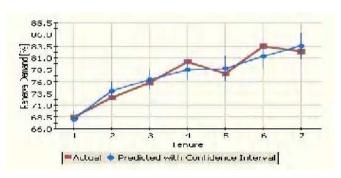
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Effect of Tenure on Renewal Demand



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Next step is optimization

The "Objective Function"

Given an objective of X subject to the condition Y what is the price I should charge?

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Xs and Ys

Optimization is not blind profit maximization!

Possible Objectives (X)

- · More profit
- · More volume
- More retention

Possible Constraints (Y)

- Volume
- Retention
- Profit
- Lifetime value

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Optimization

- Once you have defined the objective function, you must find the optimal points
- Use calculus to find the minimum/maximums
- Because of the complexity of the objective function and the constraints, you will need to use software (or Excel Solver).

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Arguments against optimization

2. "I want to maximize PIF and take no policies below the cost of capital, therefore I don't want to under price (capital destruction) and I don't want to overprice (I won't sell as many policies)."

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OJ Sidebar: Lifetime Value

- · Lifetime value is the present value of a piece of business today to the company
- · Easy to explain, but hard to implement

Example: a 25 year old single male buys a liability only

- · Will he eventually get full coverage?
- · Will he eventually get married (and stay with the company)?
- · Will he buy a homeowner's policy from us?
- · Will he buy life insurance?

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Arguments against optimization

- 1. We are getting away from expected costs.
- European companies are monitoring this issue, they haven't seen major problems.
- Hard market would likely see focus return to costs.

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Arguments against optimization

Answer: Focus on Marginal ROE

- •Determine the amount of expenses fixed over the policy term.
- Throw these OUT!!!!
- •If fixed expenses are 10% of last year's premium, and you need to price to a 4% underwriting profit, you can now price a policy to -6% underwriting profit and still make your return on capital,

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Getting optimization into the rates . . .

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		Current Factor/ Rates	Indicated Factor/ Rates	Current Margin	Earnix Factor/ Rates	Earnix Margin
Bodily Injury – Illinois Auto		250	260	6.0%	249	2.0%
Driver Class	35 Married Male	0.85	0.89	0.0%	0.90	1.5%
Points	Clean	1.00	0.99	0.0%	1.01	2.0%
Territory*	Carbondale	1.20	1.14	0.0%	1.14	0.0%
Symbol*	13	1.10	1.12	0.0%	1.12	0.0%
Model Year	2004	1.15	1.22	0.0%	1.24	1.5%
Credit Score	700	0.80	0.74	0.0%	0.75	2.0%
MultiCar	Yes	0.85	0.83	0.0%	0.84	0.5%
Homeowner*	Yes	0.95	0.93	0.0%	0.93	0.0%
Final Rate for this Risk		208	202	6.0%	210	9.4%

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Regulatory Issues

Regulation - Open Issues

- Optimization began in Europe and Israel where there is little rate regulation.
- •Easier to implement in commercial lines.
- •Might be possible to optimize a regulated line if you have related products (example: worker's comp)
- Personal lines implementation will vary by state.

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What is optimization?

Optimization is a tool to assist rating judgment to balance these factors as well as actuarial considerations, its just formalizing what we currently do.

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Efficient Frontier of Maximum Retention

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Regulatory Issues

Regulation - Open Issues

Most companies don't currently file actuarially indicated rates for every cell

- Ignored Classification Issues (Renewals versus New Business)
- Credibility
- Competitive Issues
- Stability

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Regulation

Regulation is generally done at the program level, but optimization is at the individual rate level.

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Price optimization in the travel industry

Differences

- Demand can be more elastic in travel than insurance due to ease of substitution. This will vary by consumer and (in the case of airlines) the specific route.
- High variable costs in the insurance industry means that one less policy causes significantly less costs.
- Supply is highly constrained in the short term for travel, especially hotels. (Check New York hotel rates)

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Final Thoughts

- Entrenched in Europe
- Still early in the process for the US early adapters may make a lot of money (See "Credit Scoring" circa 1990)
- · Regulatory impact unclear
- · Actuaries are very involved in the US, less so in Europe

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