



**SPE-1  
Policyholder Retention  
and its Impact on  
Pricing**

**2006 CAS Seminar on  
Ratemaking**

**Claudine Modlin, FCAS**

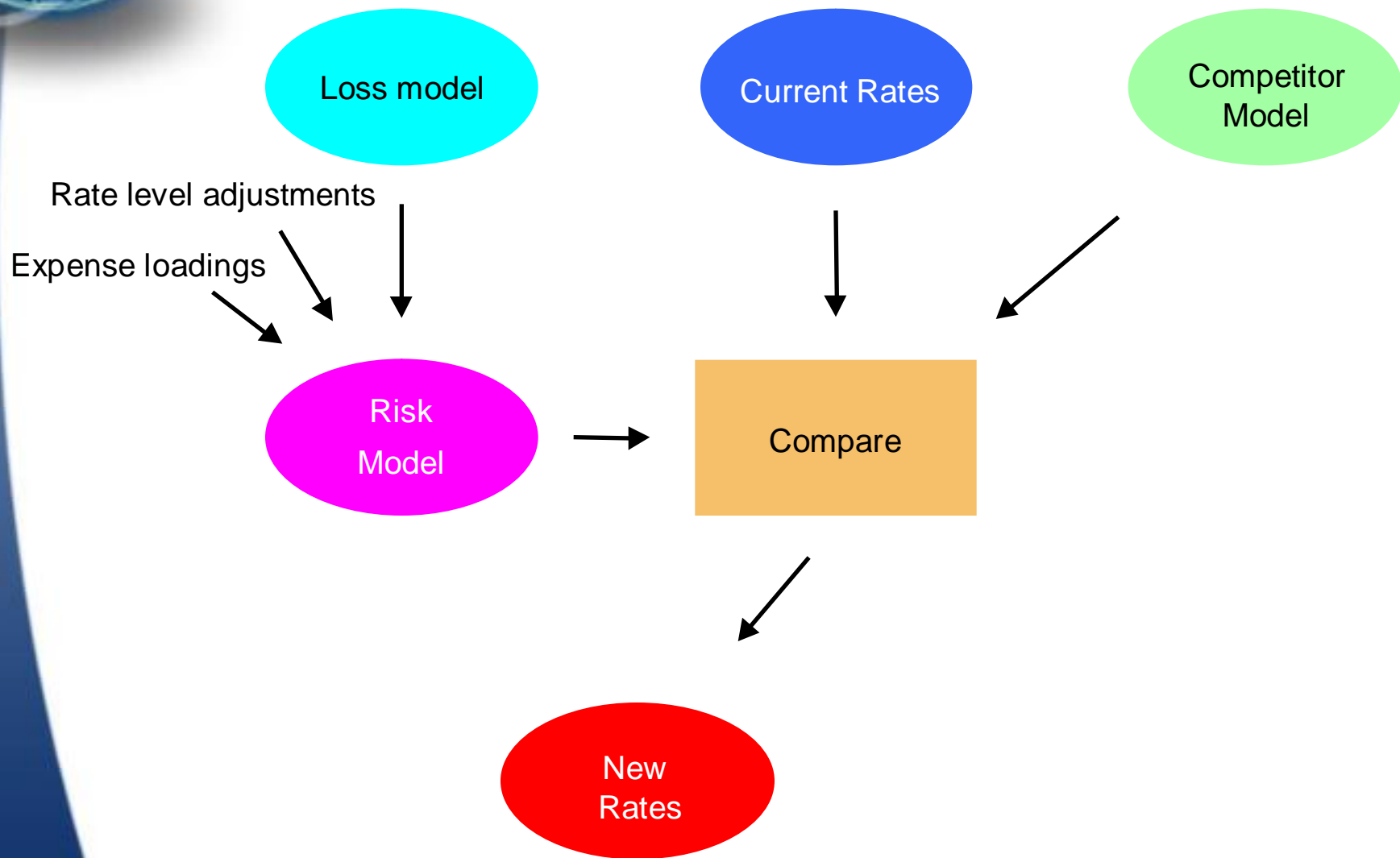
**Watson Wyatt Worldwide**



[WWW.WATSONWYATT.COM](http://WWW.WATSONWYATT.COM)

**W** **Watson Wyatt**  
*Worldwide*

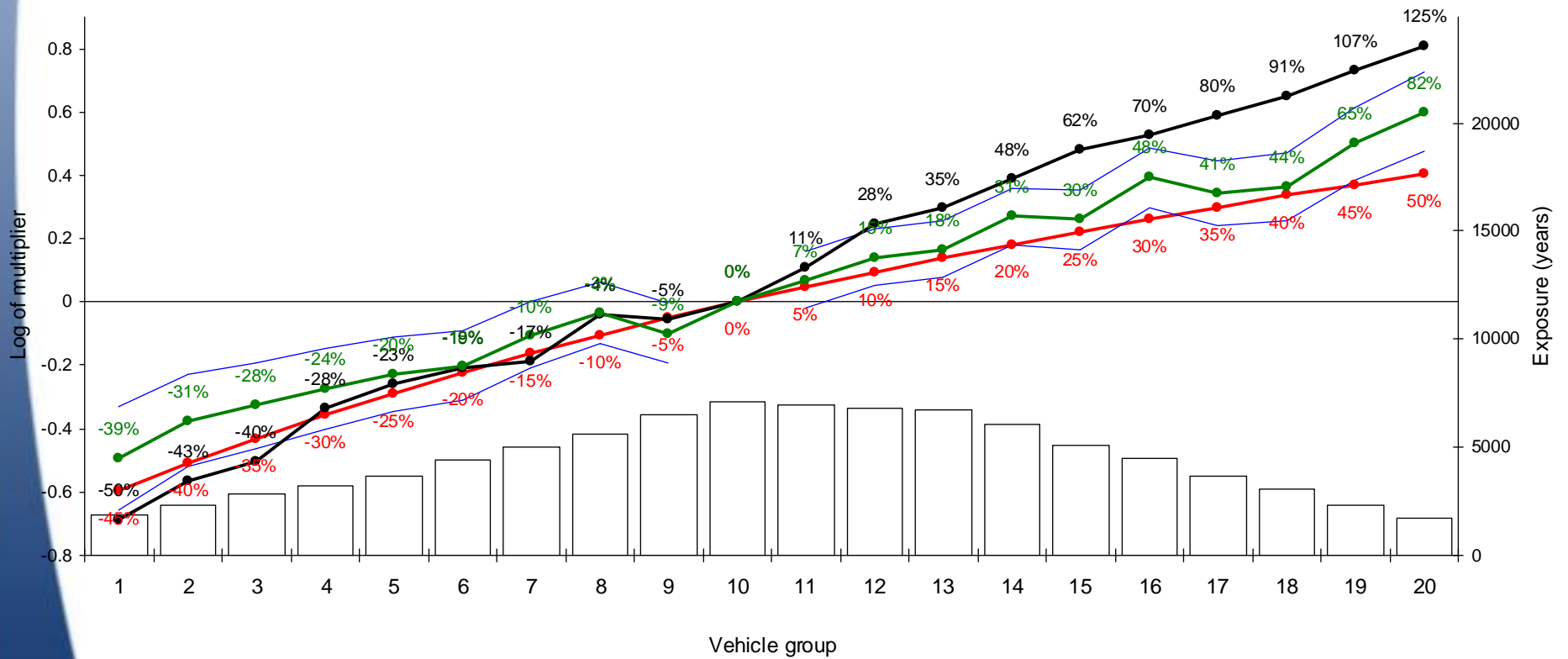
# Traditional ratemaking process



# Rate relativity indication

## Example of competitor analysis

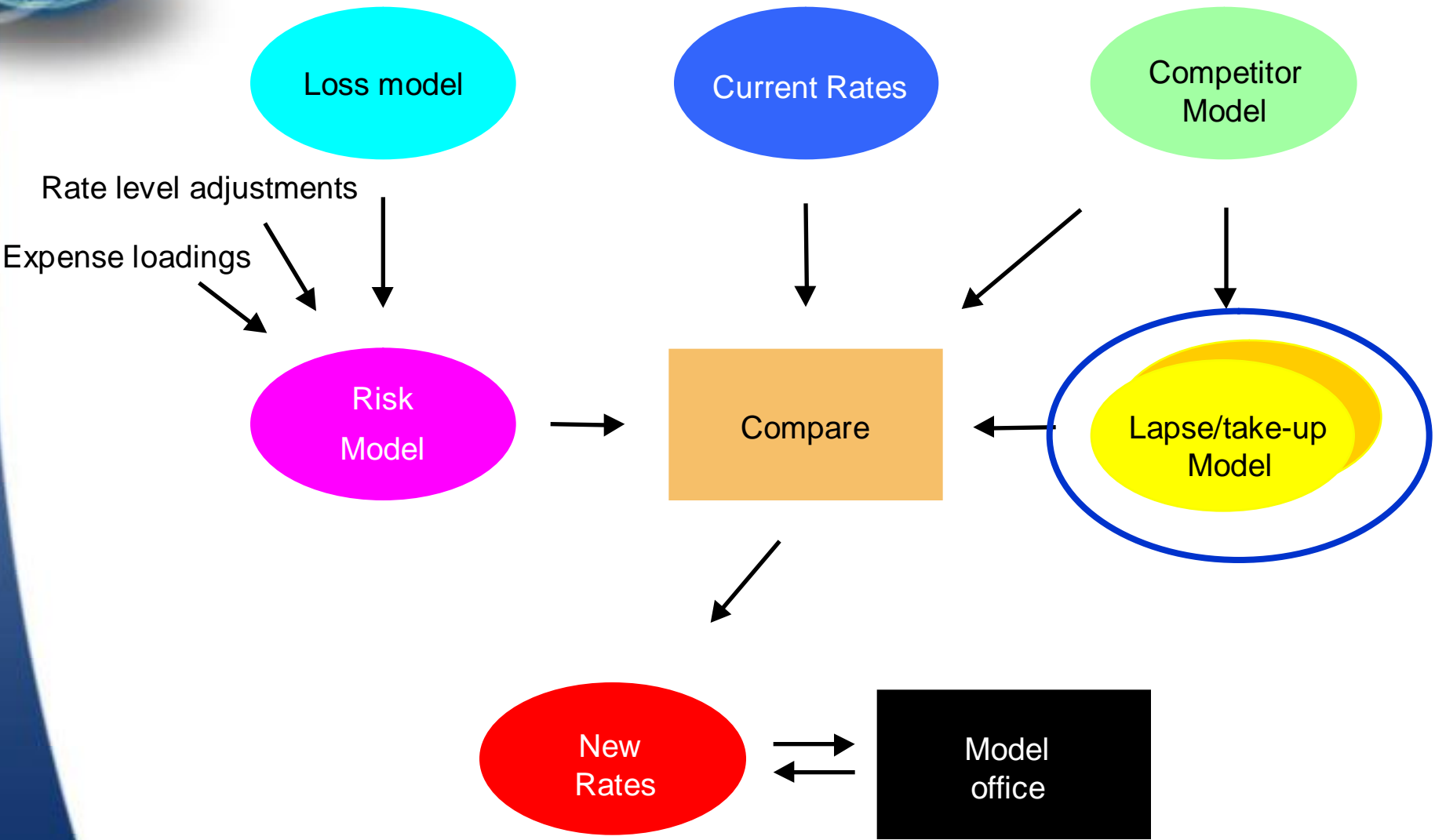
Third party cover



● Current tariff 
 — Approx 95% confidence interval 
 ● Third cheapest market quote 
 ● Smoothed estimate

P value = 0.0%  
Rank 9/11

# Full ratemaking process



# Retention / conversion analysis

- What to measure
- What to consider
- Practical tips
- Why do it





# Retention analysis

---

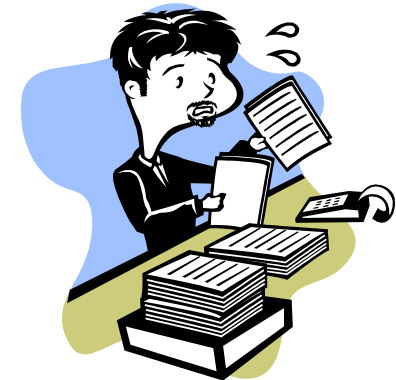
- **What to measure**
- What to consider
- Practical tips
- Why do it



# Data required

---

- Individual policy (or quote) level
- Offer & resulting accept/lapse
- Policy characteristics
- Rate change information
- Period during which rates changed





# Generalized linear models

---

$$E[\underline{Y}] = \underline{\mu} = g^{-1}(\underline{X} \cdot \underline{\beta} + \underline{\xi})$$

$$\text{Var}[\underline{Y}] = \phi \cdot V(\underline{\mu}) / \underline{\omega}$$

- Consider all factors simultaneously
- Allow for nature of random process
- Provides diagnostics
- Robust and transparent



# "A Practitioner's Guide to Generalized Linear Models"

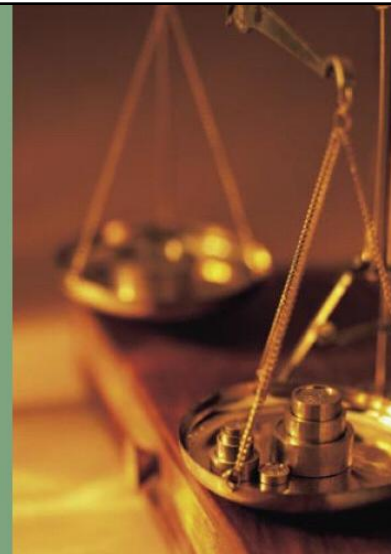
## A Practitioner's Guide to Generalized Linear Models

A foundation for theory,  
interpretation and application

Second edition - May 2005

Paper authored by:

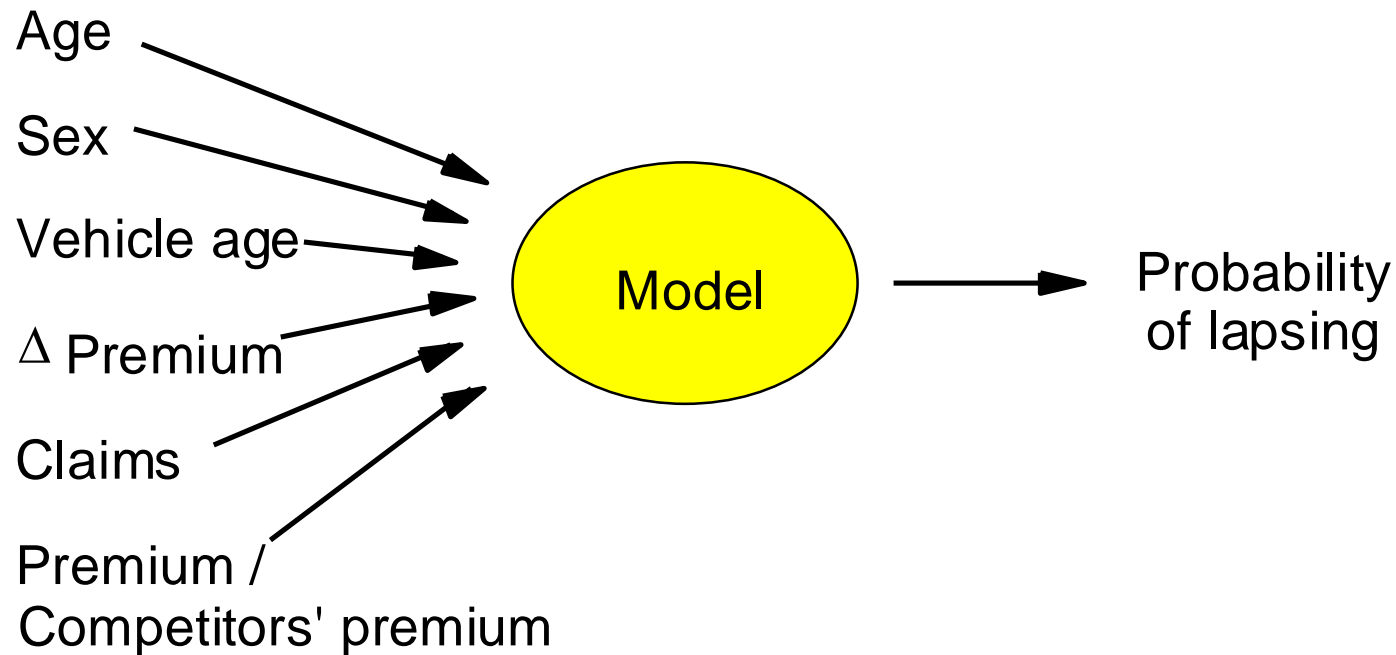
Duncan Anderson, FIA  
Sholom Feldblum, FCAS  
Claudine Modlin, FCAS  
Doris Schirmacher, FCAS  
Ernesto Schirmacher, ASA  
Neeza Thandi, FCAS



WWW.WATSONWYATT.COM

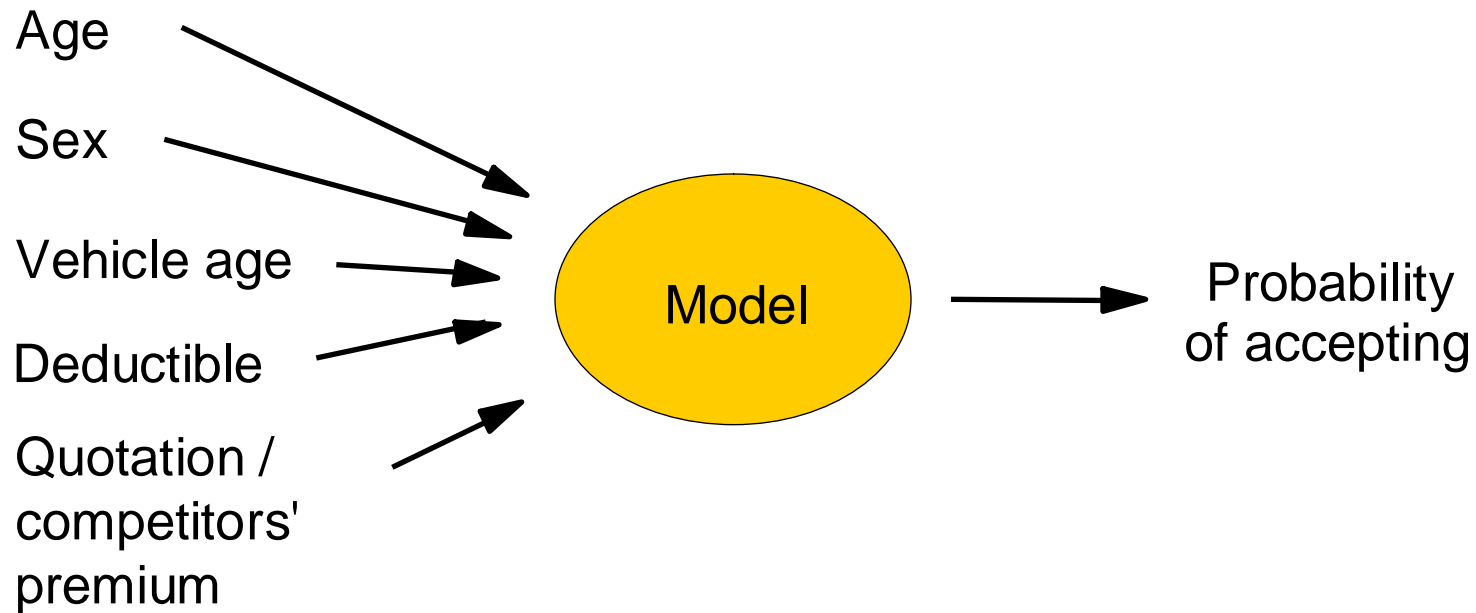
# Modeling retention

- Most companies have data on renewal offers



# Modeling new business rates

- If details of individual quotes known, can be modeled in similar way
- Otherwise much simpler analysis is all that can be undertaken





# Retention analysis

---

- What to measure
- **What to consider**
- Practical tips
- Why do it





# What to consider

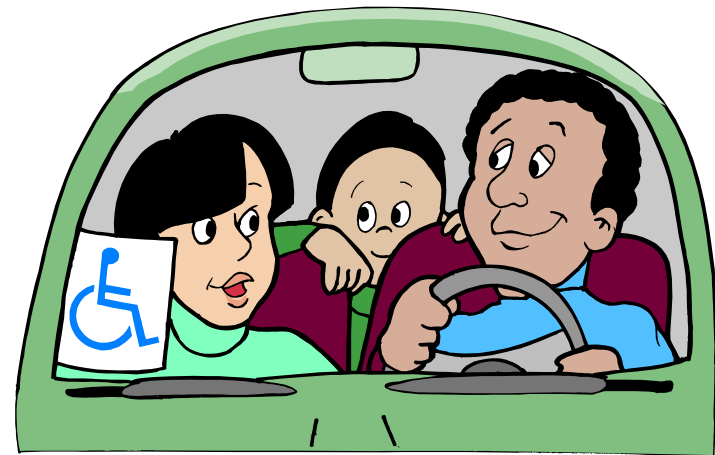
---

- Who are your customers
- How do you connect
- What have you done to them
- What have others done to them

# Who are your customers?

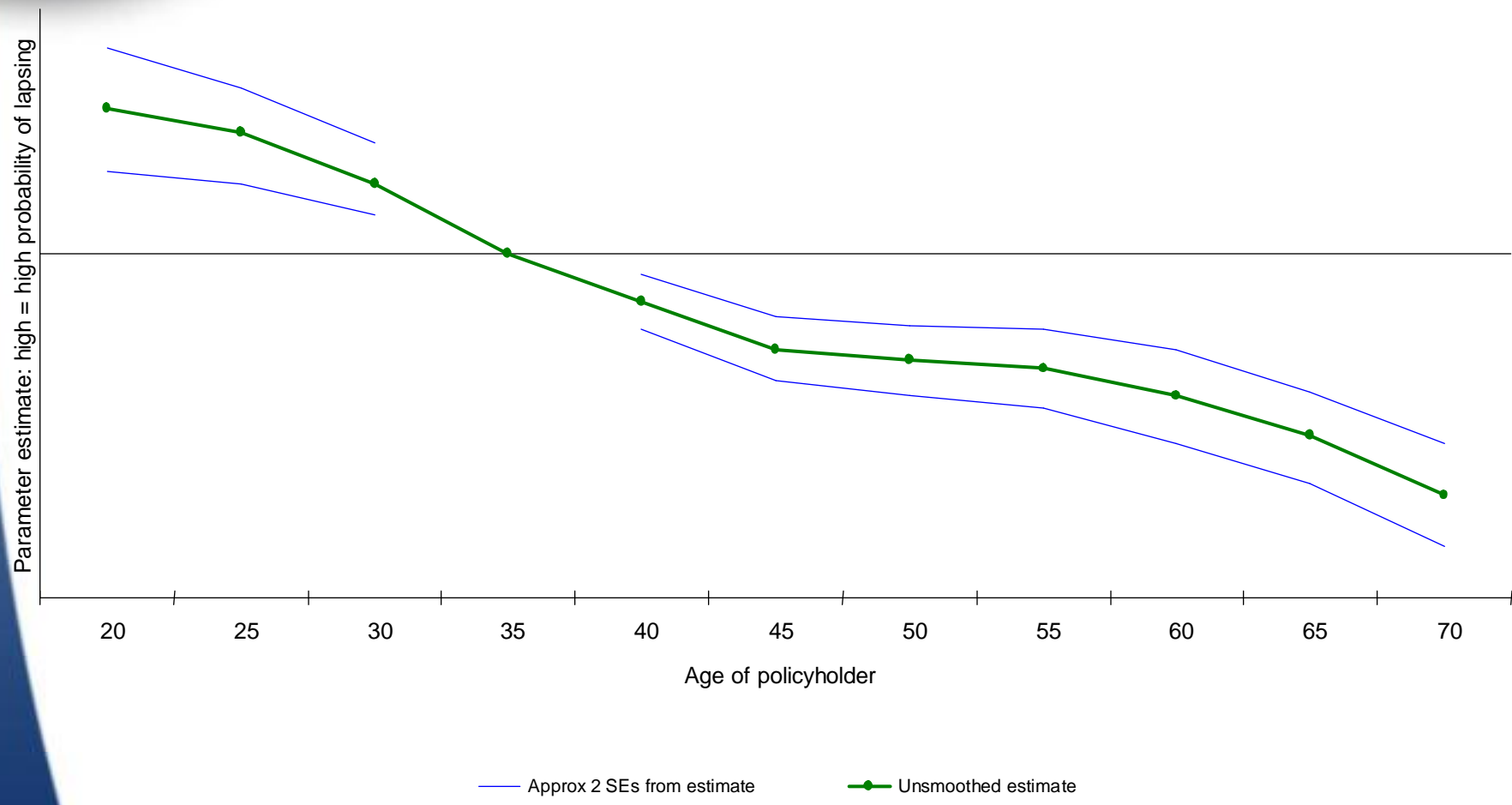
---

- Age of policyholder
- Age of car
- Claims history
- Other rating factors
- Endorsement activity





# Effect of age of policyholder on lapses





# How do you connect with them?

- Distribution channel
- Payment plan
- Affinity membership
- Other products held
- # years with company



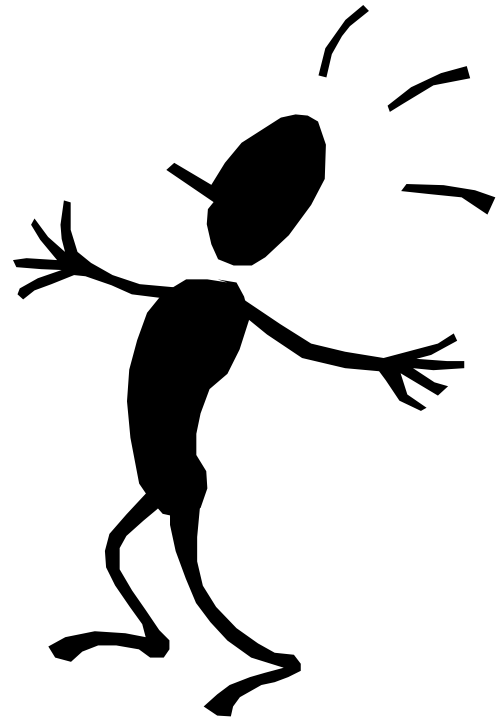




# **What have you done to them?**

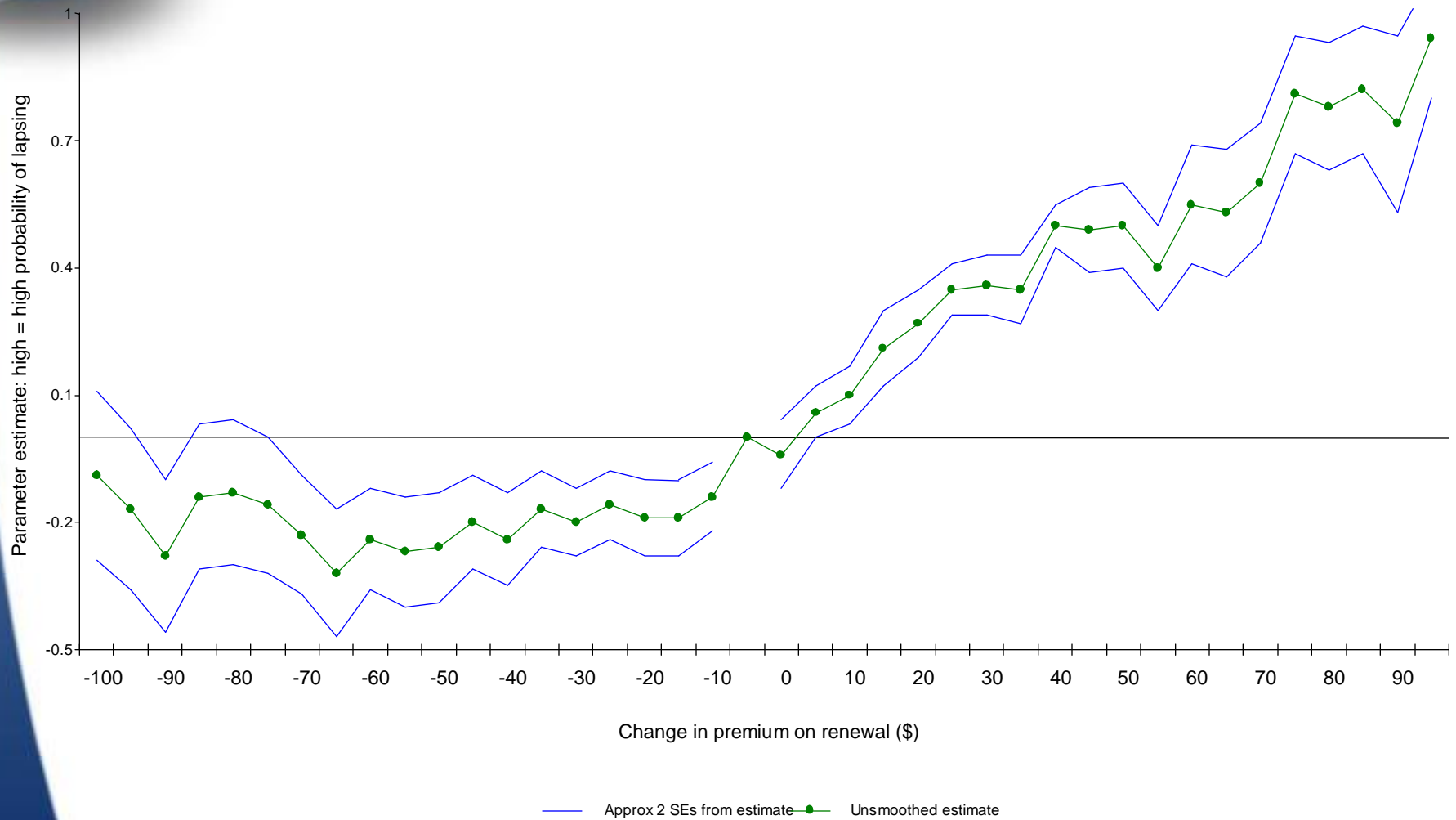
---

- Rate change
- Claims service
- Agent service





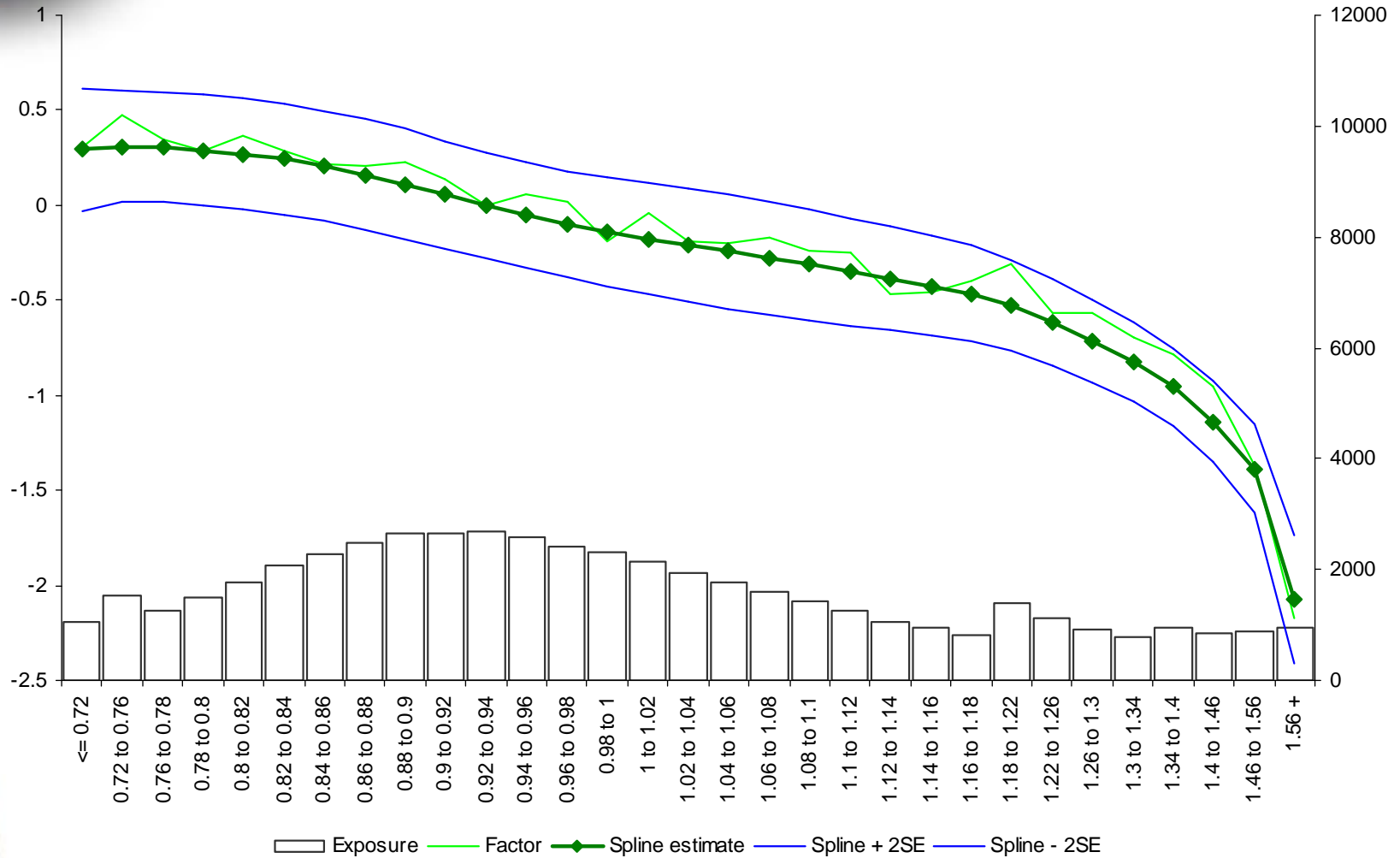
# Effect of premium change on lapses





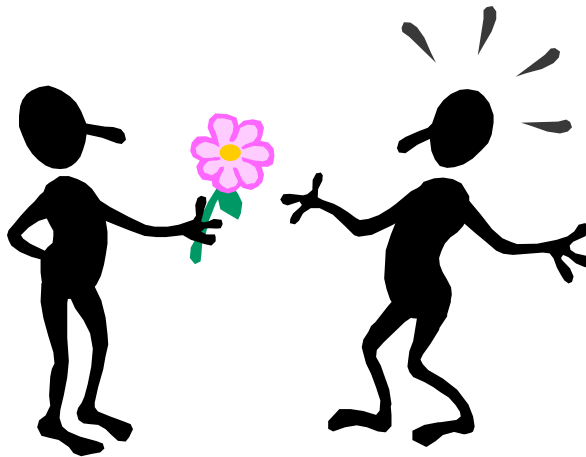
# Splines

Effect of premium change on renewal using cubic splines



# What have others done to them?

- Competitors' premium
- Product differentiation  
(may not be applicable to some products)





# Competitive indices

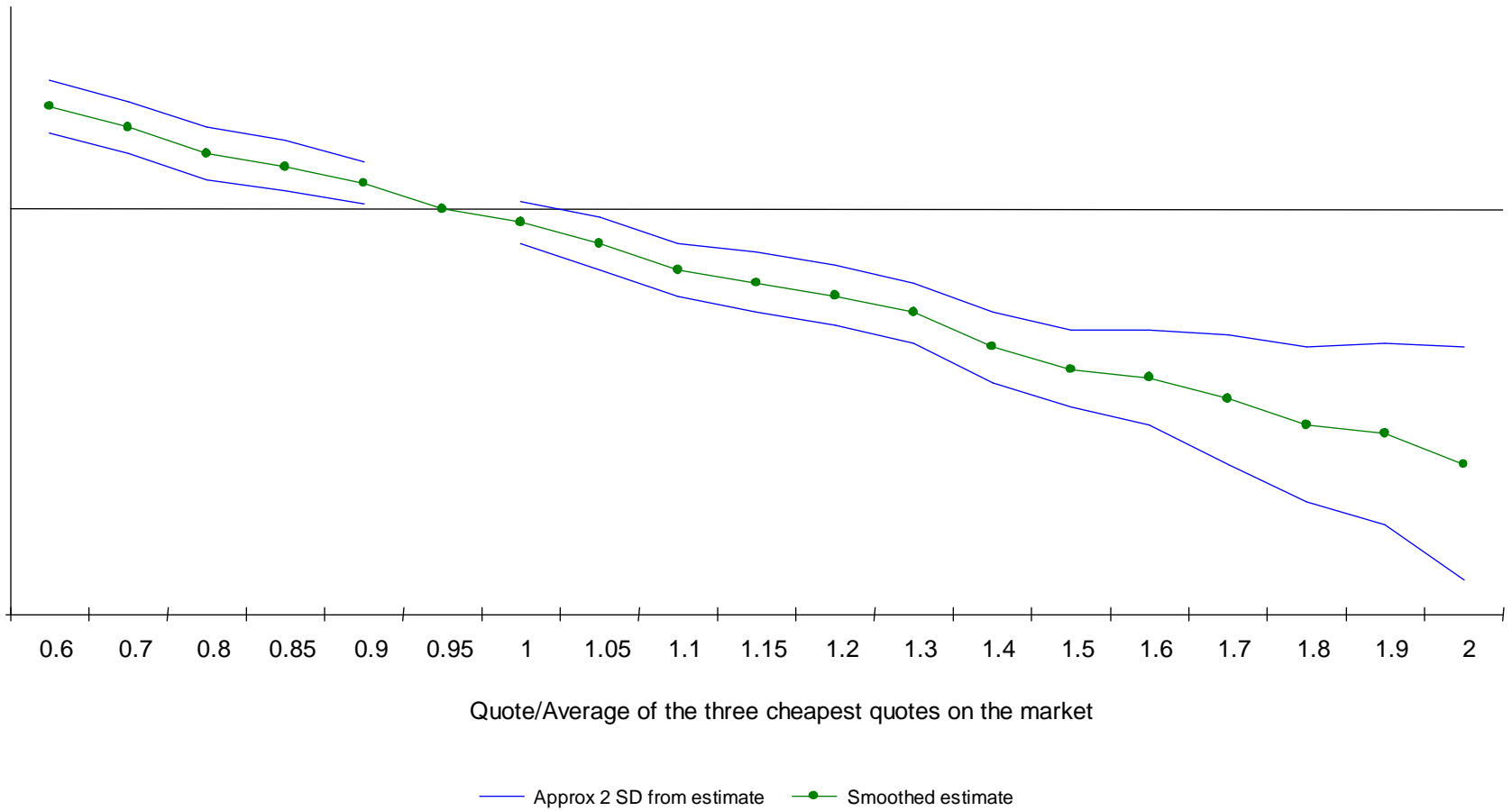
---

- For modeling, required at individual policy level
- Sources of competitor info
  - rate manuals
  - comparative rating software
- Measures
  - index (comparing to one competitor or averaged across several)
  - rank of quote relative to competitors
- Challenges
  - tier criteria
  - point in time
  - cost



# Effect of competitiveness on new business

Parameter estimate: high = high probability of quote being accepted

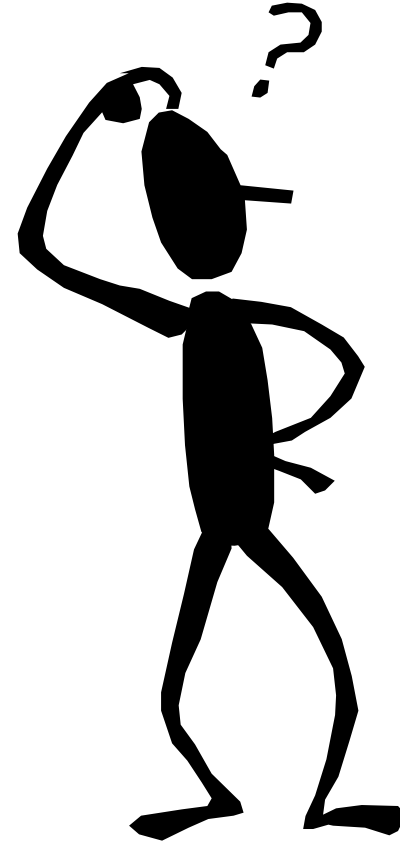




# Retention analysis

---

- What to measure
- What to consider
- **Practical tips**
- Why do it





# Statistical assumptions

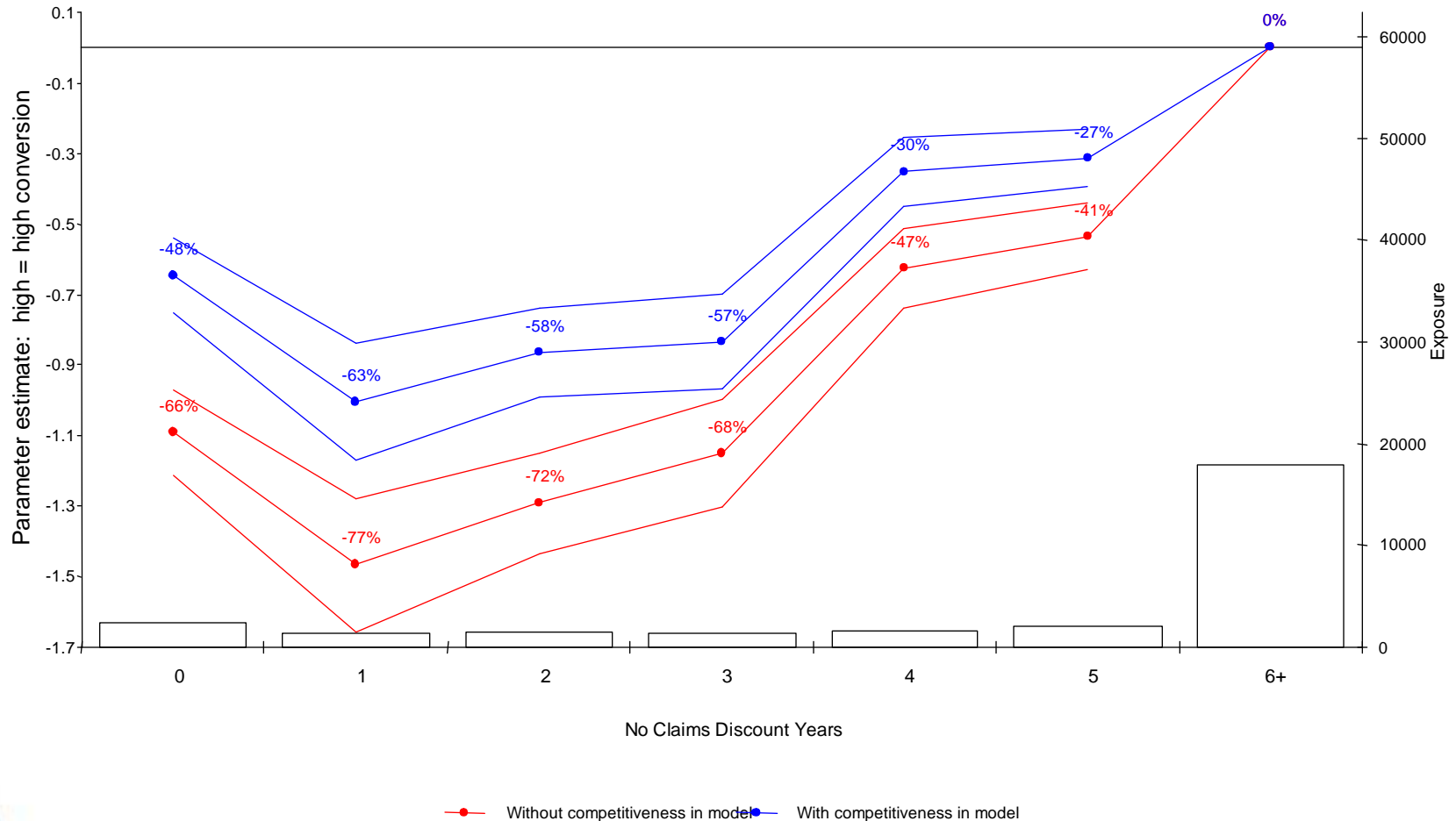
---

- A logistic model is most appropriate
  - considers  $\log( p / [1-p] )$  and binomial error
  - maps  $[0,1]$  to  $[-\infty,\infty]$
  - invariant to whether you measure lapse/renew
- If lapses are low and results not to be used directly, a Poisson multiplicative model can help
  - theoretically wrong (can predict multiple lapses), but:
  - easier to understand
  - can superimpose one-way results more easily



# Practical tip on competitiveness

- Superimposing models with and without competitiveness will show extent to which effects are simply price related





# Beware absolute premium

---

- GLM shows effect *all other factors being equal*
- For varying premium all other factors are never equal
- Results, while statistically correct, can be hard to interpret, for example adding premium size can reverse the multivariate result for age of driver
- Consider fitting separate models for different premiums bands



# Measuring rate change

---

- Best to have more than one rate change in data
- Investigate % change and \$ change
- Suggest fit rate change as a categorical factor and then model with splines if appropriate
  - some results are straight lines in logistic space, some are clearly not

# Beware expectations

---

- Customer expectations of premium change
  - try to isolate rate change from risk criteria change which affects premium
  - consider premium change adjusted for change in risk criteria (ie new rates for new risk / old rates for new risk)





# Retention analysis

---

- What to measure
- What to consider
- Practical tips
- **Why do it**



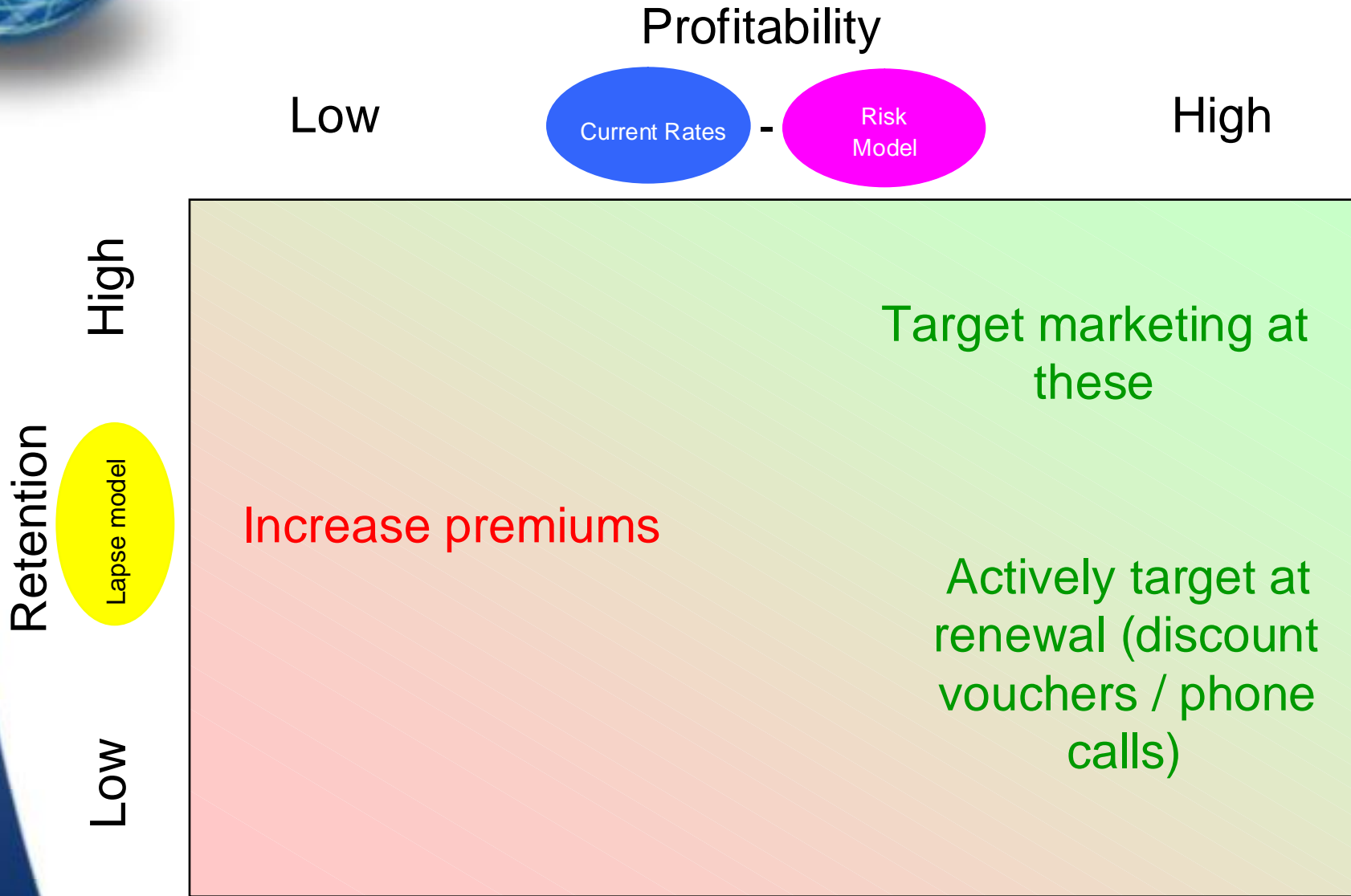


# **Why model lapses / new business?**

- Qualitative management decisions
  - marketing strategies
  - renewal campaigns
- Simple expense loadings
- Modeling
  - simple lifetime modeling
  - detailed impact modeling
  - detailed lifetime modeling
  - price optimization



# Customer value





# Lifetime expense loads

---

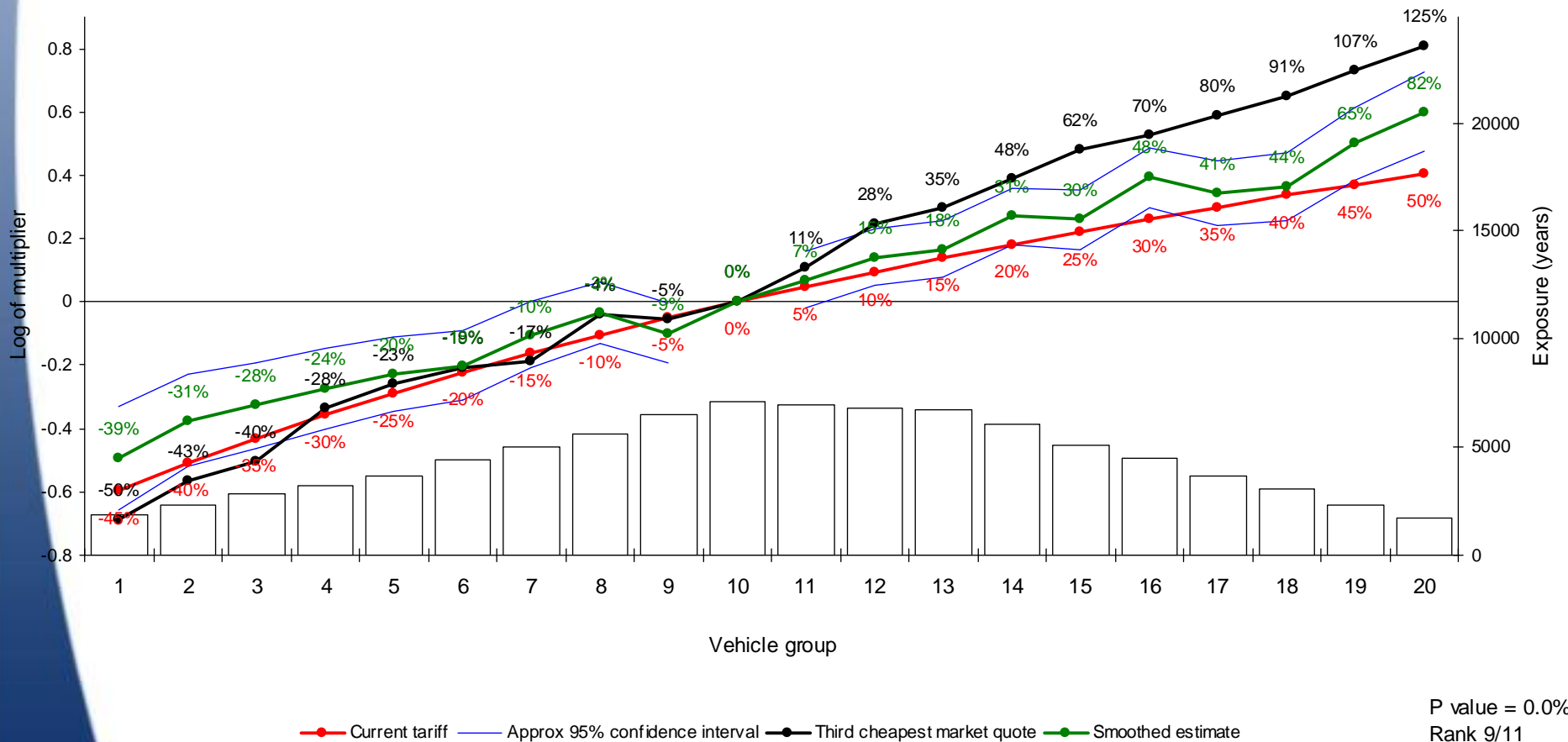
- Expenses per policy
  - acquisition 100
  - renewal 30
- Expected lifetime
  - young 2 years
  - old 5 years
- Lifetime expense loadings
  - young  $( 100 + 1 * 30 ) / 2 = 65$
  - old  $( 100 + 4 * 30 ) / 5 = 44$



# Rate relativity indication

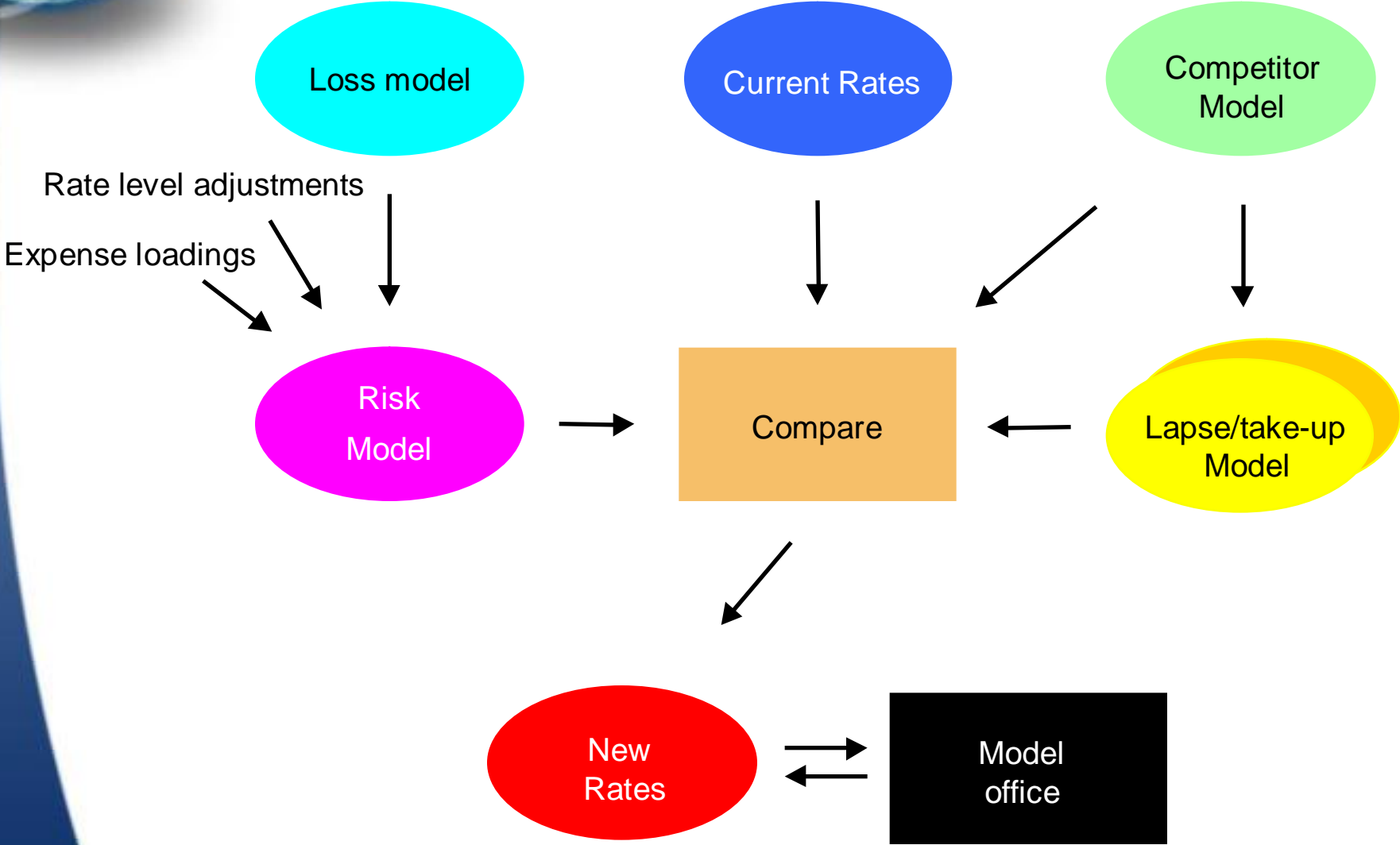
## Example of competitor analysis

Third party cover



P value = 0.0%  
Rank 9/11

# Price optimization





# Scenario testing and price optimization

---

- What will happen if I do rating action X?
- What is the "best" rating action?
  - given a form of rating structure, seek the parameters which maximize a company's strategic objectives, perhaps with defined constraints



# Price optimization in four easy steps

---

1. Assemble ingredients
2. Build a "model office" scenario test
3. Define problem and success criteria
4. Optimize



# Ingredients

---

Data

Portfolio now

Current Rates

Assumptions

Competitor Model

Expenses

GLMs

Loss model

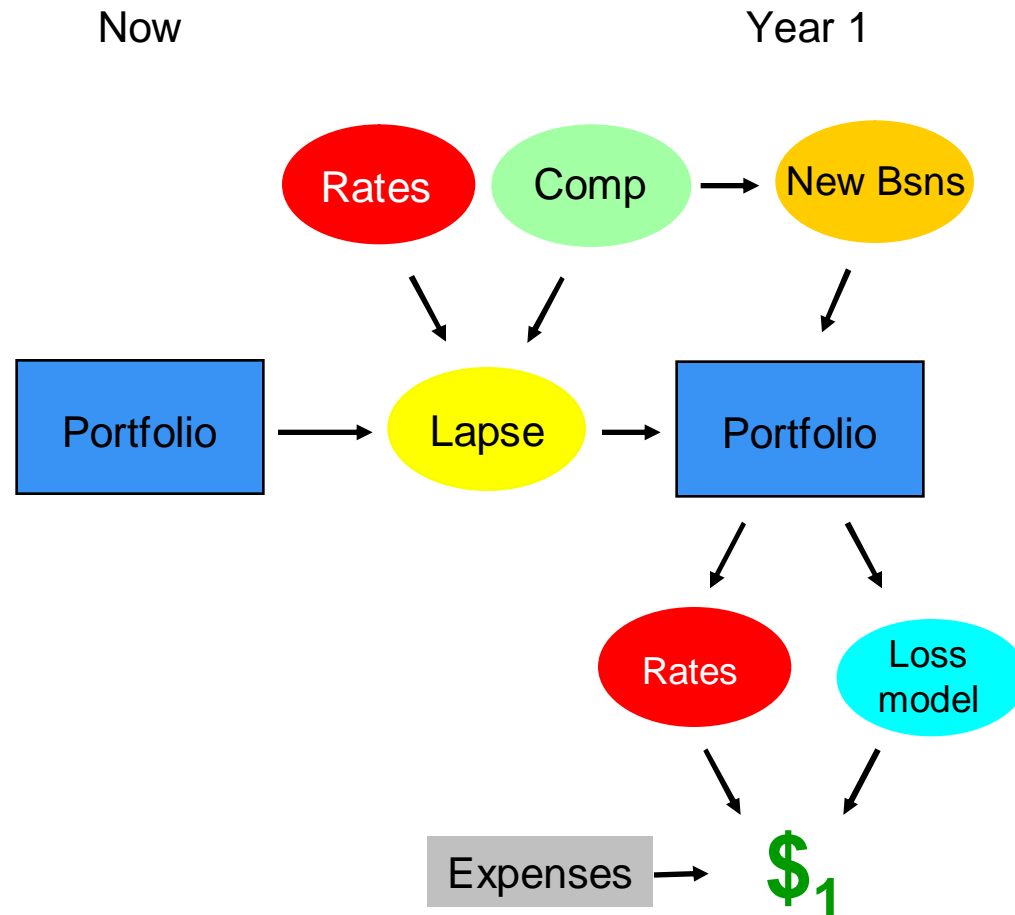
Lapse model

New business model

Test

New Rates

# Scenario testing





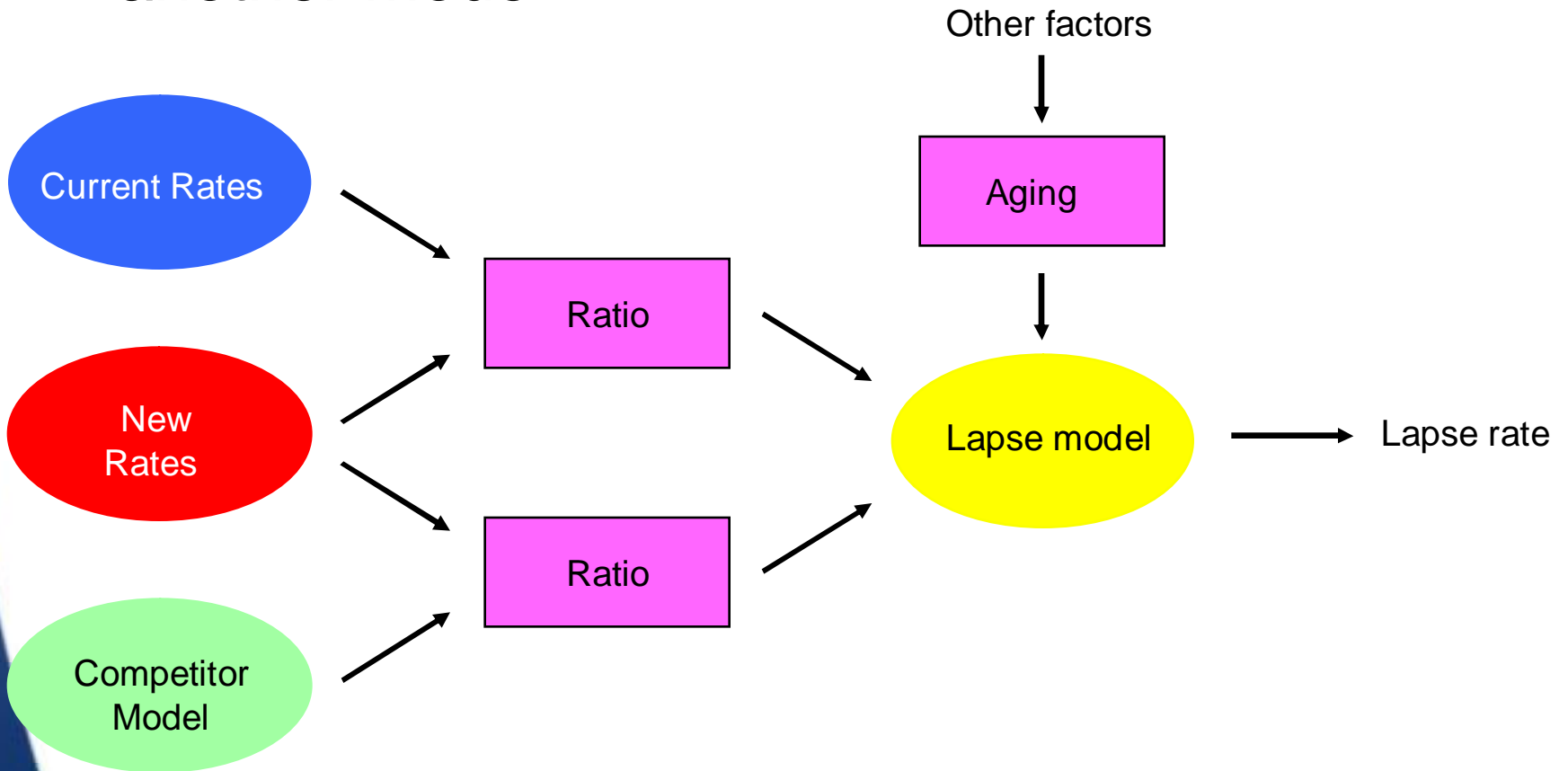
# Issues

---

- Competition
- Changes to model
  - age of insured
  - age of vehicle (home)
  - claim surcharges
  - vehicle (home)
  - address
- Programming issues
- Period of projection
- Success criteria

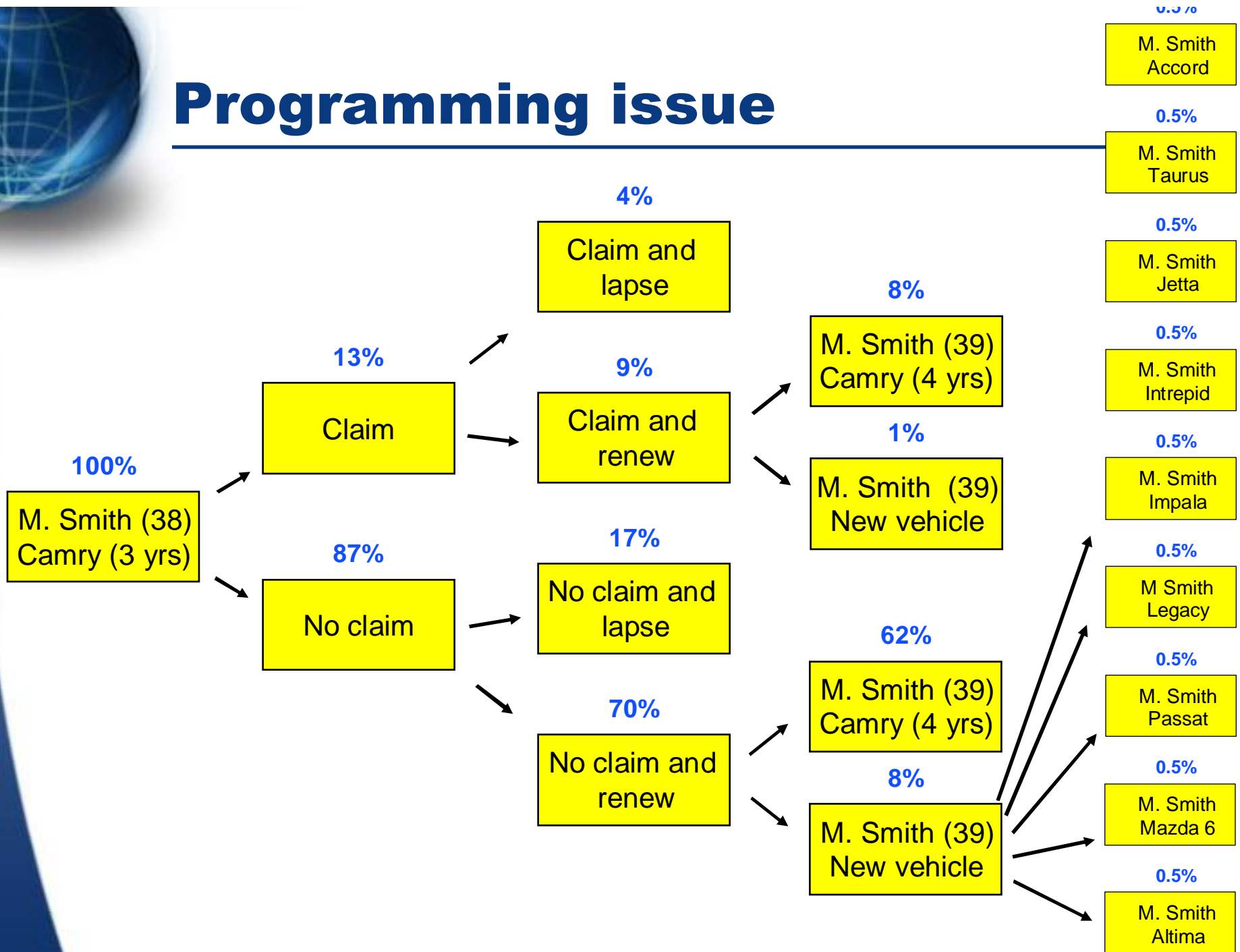
# Programming issue

- Sometimes model output needs to be processed and/or recategorized before being input to another model



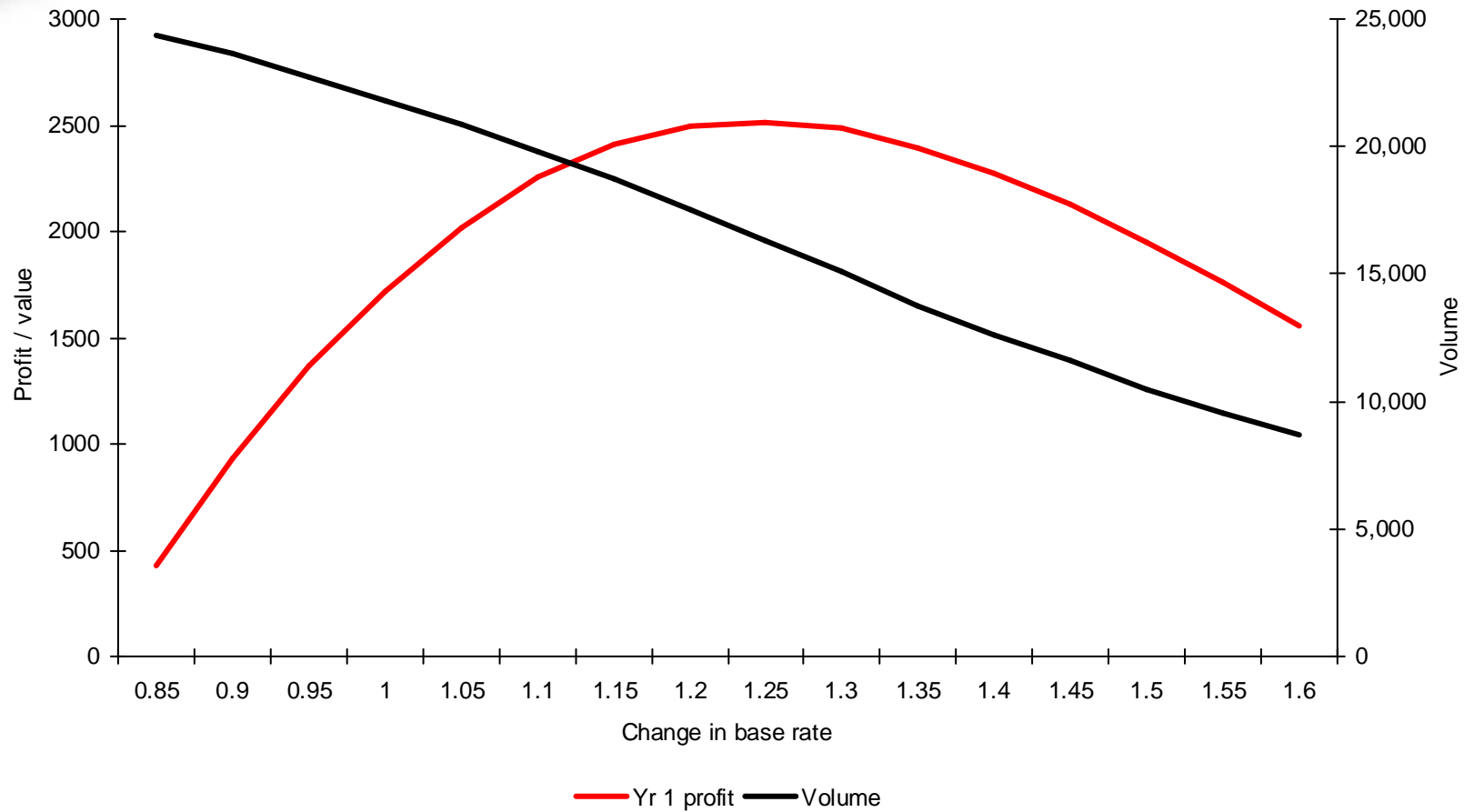


# Programming issue



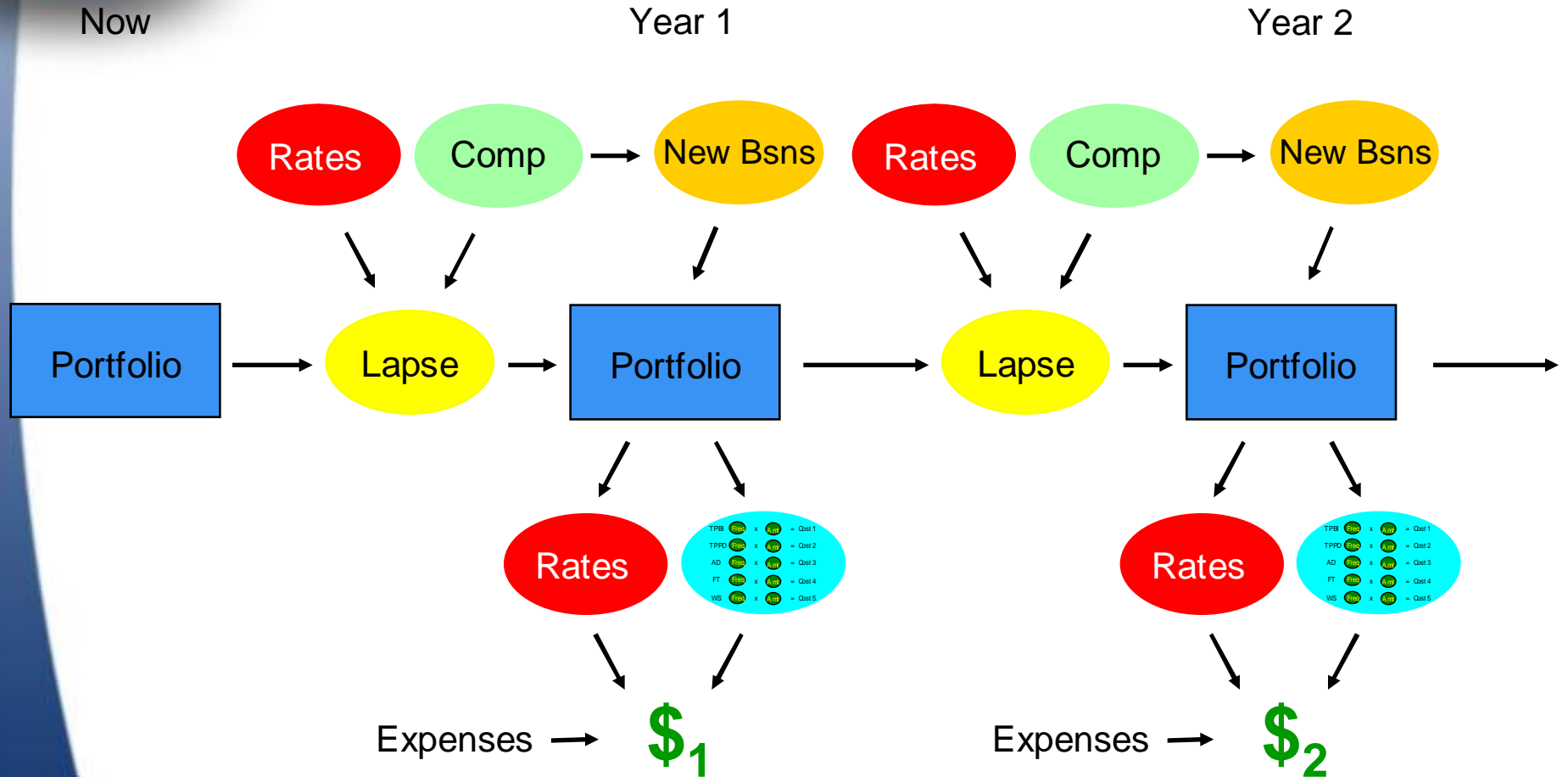


# Period of projection

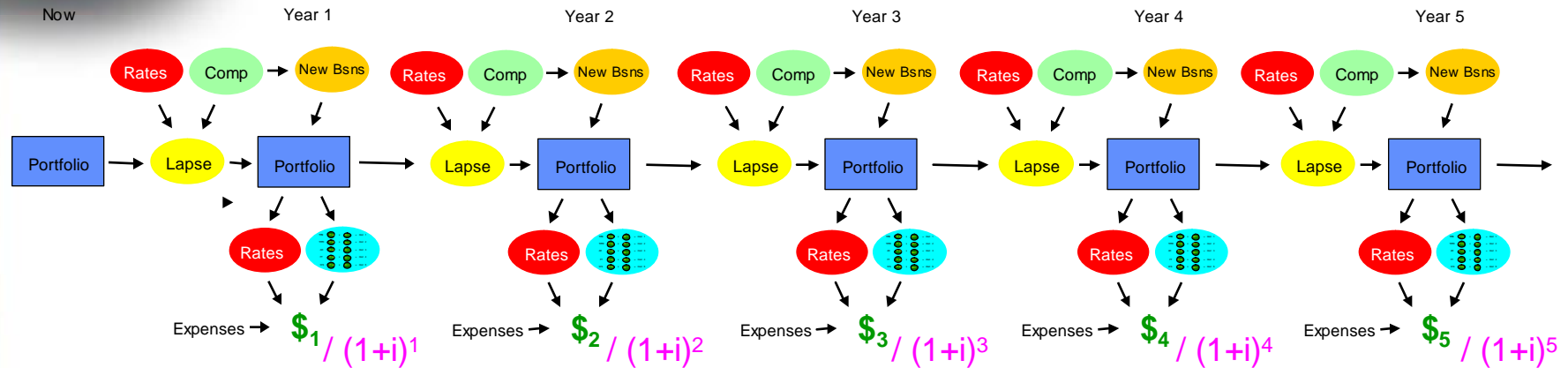




# Multiple year projections

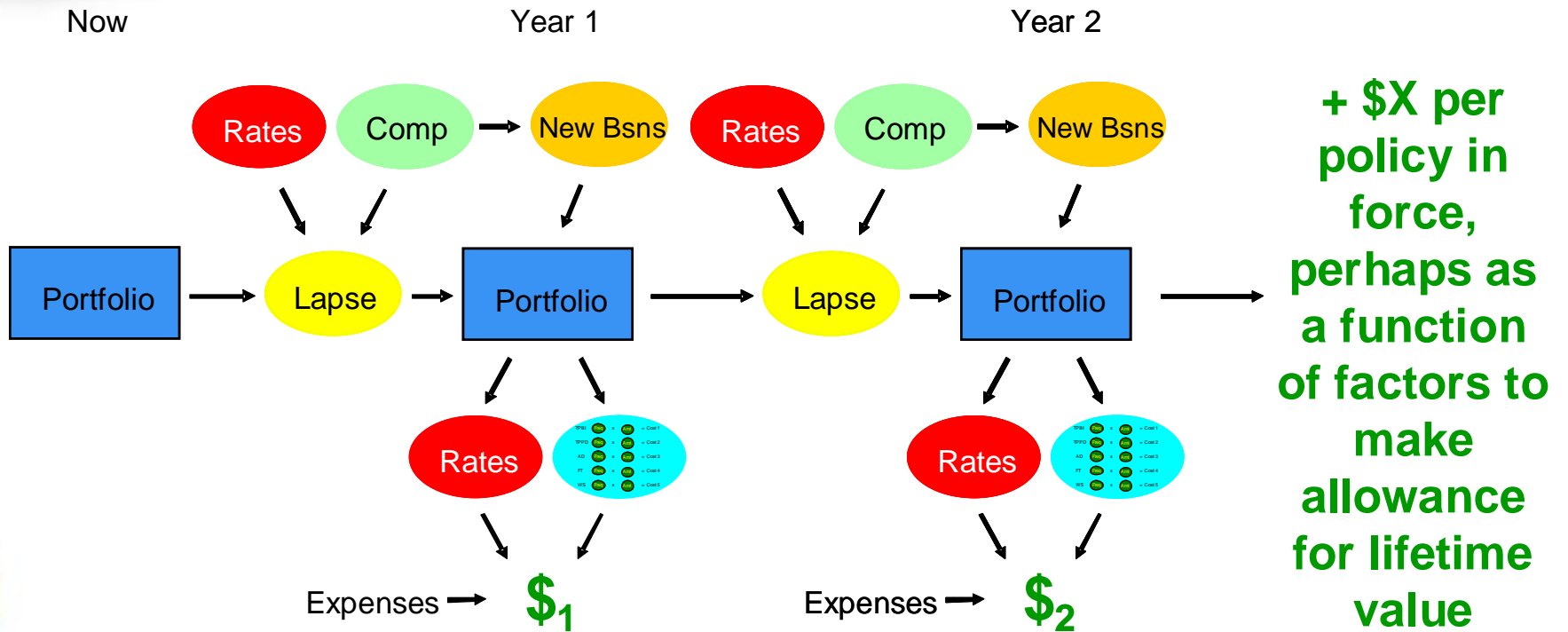


# Multiple year projections



- In theory project many years
- In practice assumptions become too uncertain and model becomes too complex

# A pragmatic compromise?





# Model office scenario tests

---

- Can be very simplistic or highly detailed
- GIGO - level of sophistication dependent upon quality of assumptions and models
- Once constructed can be used to optimize anything from a minor change, eg base rate adjustment, to a wholesale review of entire rating structure

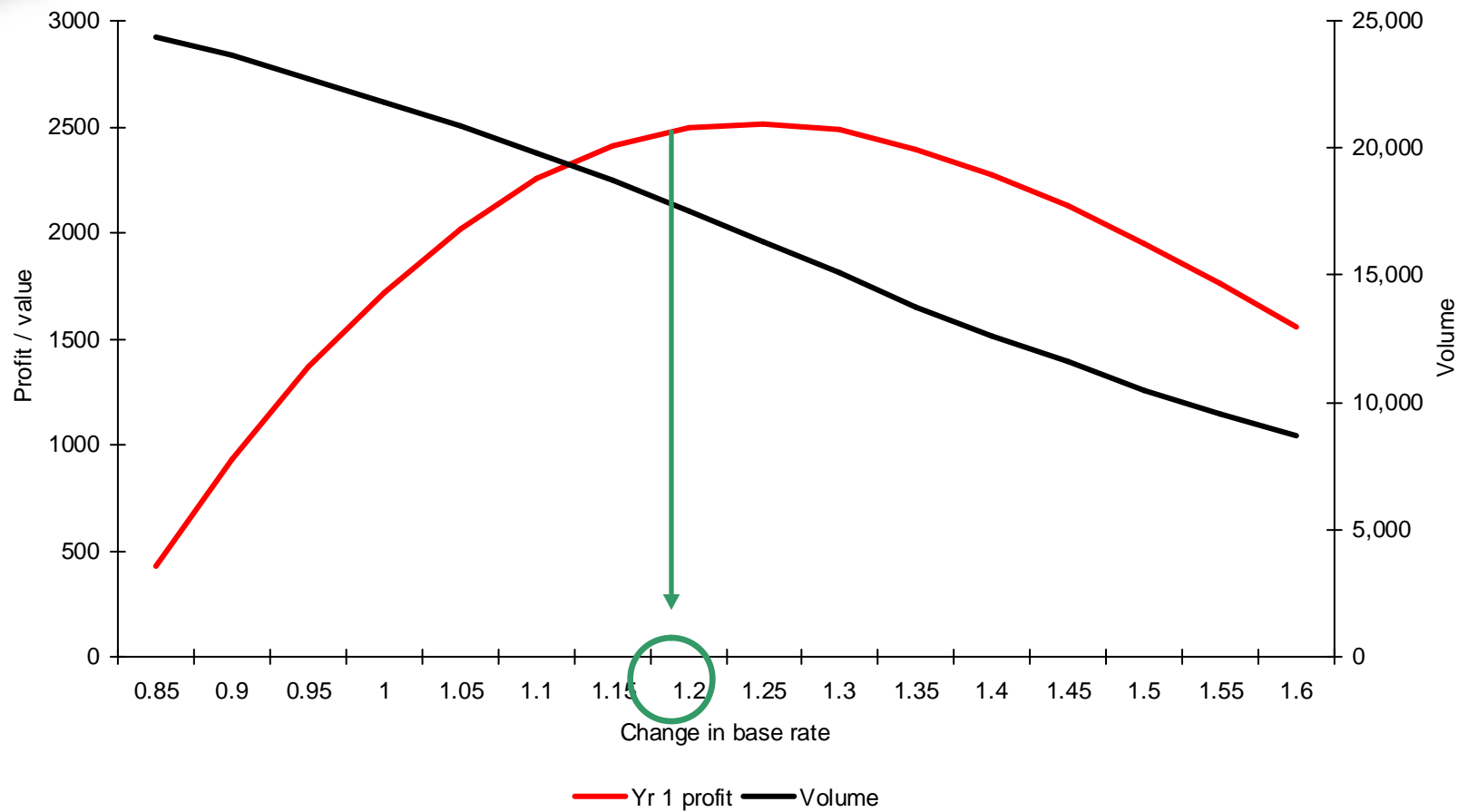


# Types of optimization

---

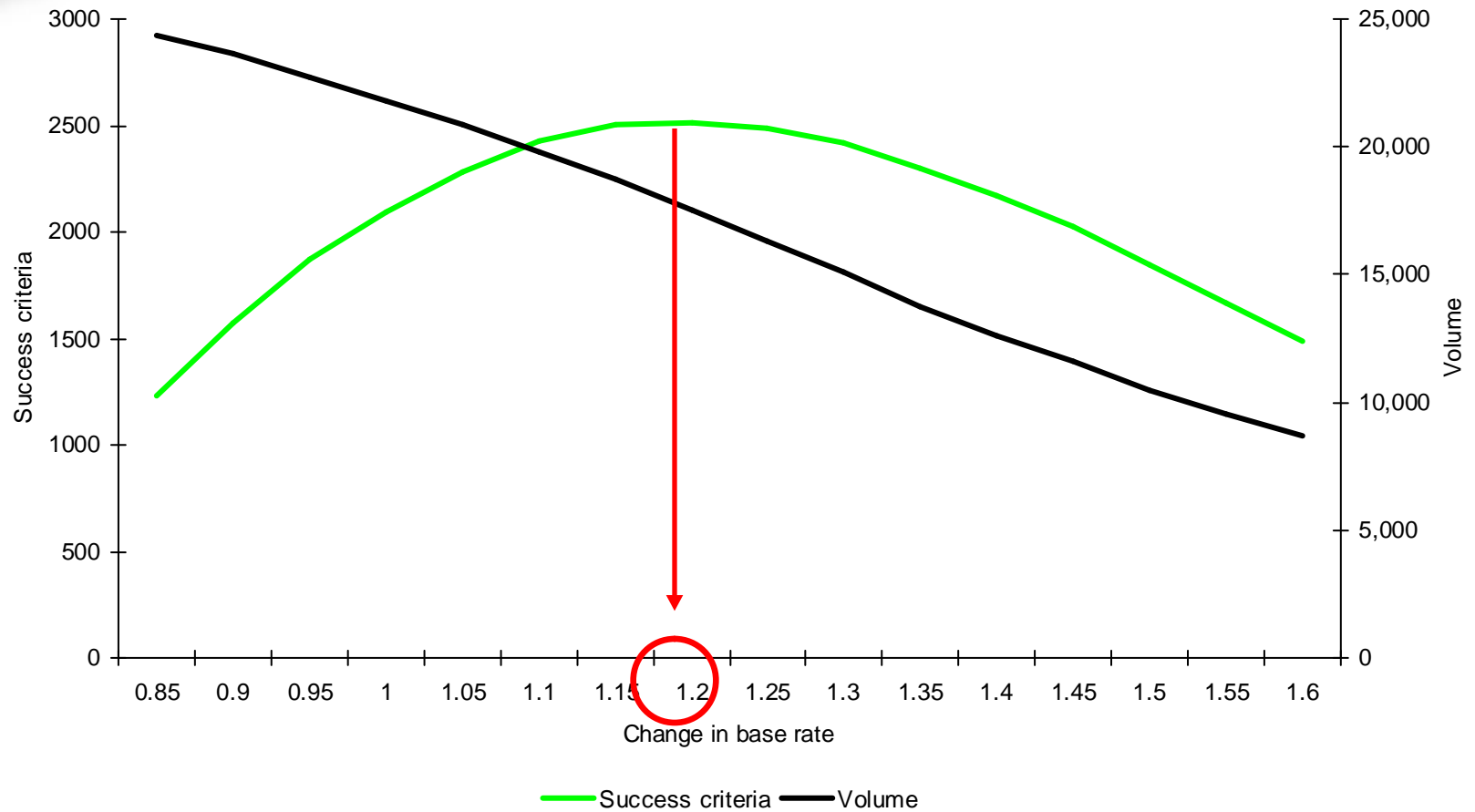
- Optimization via scenario tests, for example:
  - base rate change
  - base rate change with simple relativity change
  - moderator algorithms
- Full optimization for each individual policy at point of sale

# Base rate change - consider profit vs volume



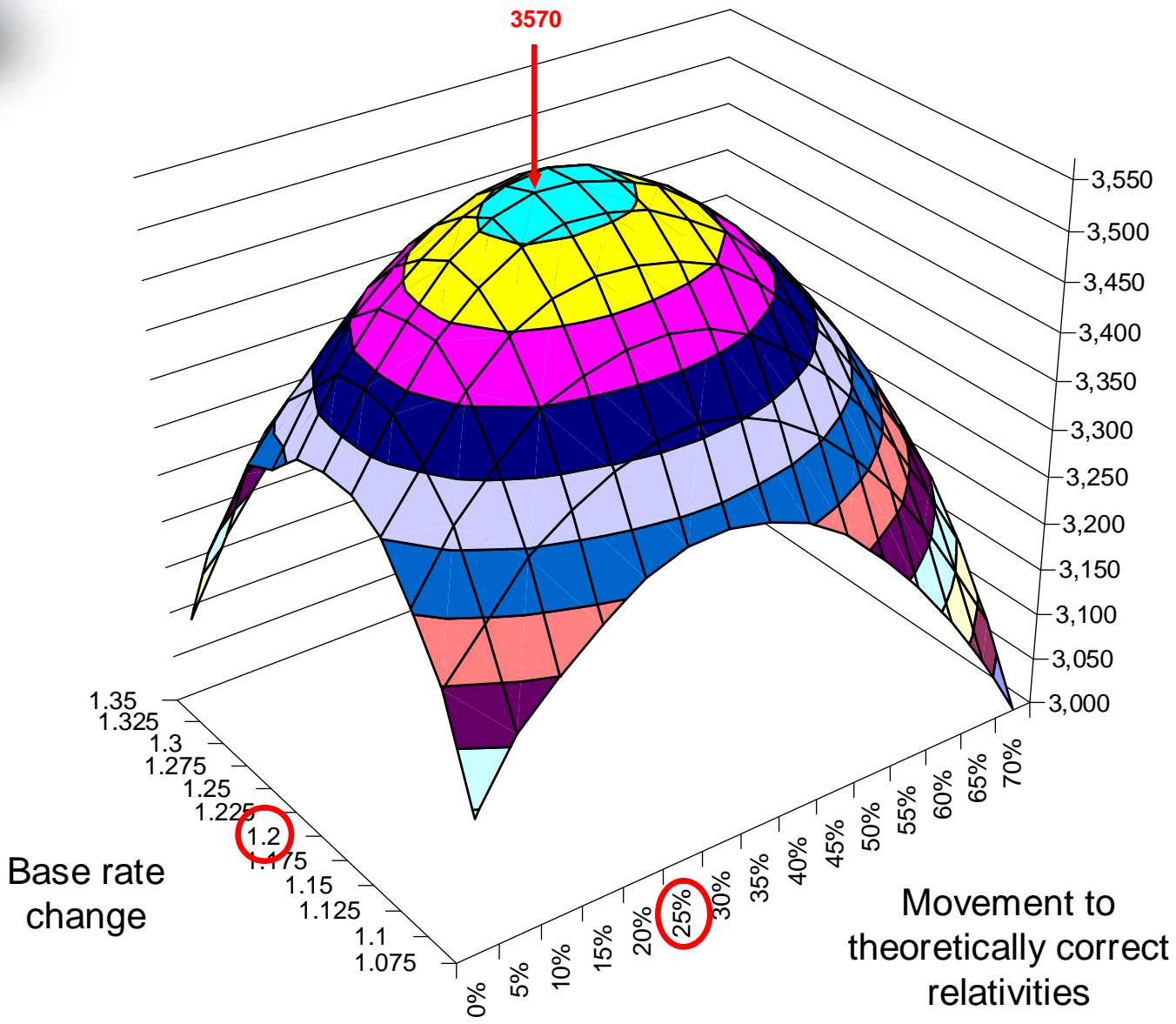


# Base rate change - single success criteria





# Base rate change with relativity change





# Moderators

## Types of rating structures - simple multiplicative

\$621.50 x

| Age   | Factor |
|-------|--------|
| 17    | 2.52   |
| 18    | 2.05   |
| 19    | 1.97   |
| 20    | 1.85   |
| 21-23 | 1.75   |
| 24-26 | 1.54   |
| 27-30 | 1.42   |
| 31-35 | 1.20   |
| 36-40 | 1.00   |
| 41-45 | 0.93   |
| 46-50 | 0.84   |
| 50-60 | 0.76   |
| 60+   | 0.78   |

| Group | Factor |
|-------|--------|
| 1     | 0.54   |
| 2     | 0.65   |
| 3     | 0.73   |
| 4     | 0.85   |
| 5     | 0.92   |
| 6     | 0.96   |
| 7     | 1.00   |
| 8     | 1.08   |
| 9     | 1.19   |
| 10    | 1.26   |
| 11    | 1.36   |
| 12    | 1.43   |
| 13    | 1.56   |

| Sex    | Factor |
|--------|--------|
| Male   | 1.00   |
| Female | 1.25   |

| Area | Factor |
|------|--------|
| A    | 0.95   |
| B    | 1.00   |
| C    | 1.09   |
| D    | 1.15   |
| E    | 1.18   |
| F    | 1.27   |
| G    | 1.36   |
| H    | 1.44   |



# Moderators

## Types of rating structures - multiplicative with moderator

\$621.50 x

| Age   | Factor |
|-------|--------|
| 17    | 2.52   |
| 18    | 2.05   |
| 19    | 1.97   |
| 20    | 1.85   |
| 21-23 | 1.75   |
| 24-26 | 1.54   |
| 27-30 | 1.42   |
| 31-35 | 1.20   |
| 36-40 | 1.00   |
| 41-45 | 0.93   |
| 46-50 | 0.84   |
| 50-60 | 0.76   |
| 60+   | 0.78   |

| Group | Factor |
|-------|--------|
| 1     | 0.54   |
| 2     | 0.65   |
| 3     | 0.73   |
| 4     | 0.85   |
| 5     | 0.92   |
| 6     | 0.96   |
| 7     | 1.00   |
| 8     | 1.08   |
| 9     | 1.19   |
| 10    | 1.26   |
| 11    | 1.36   |
| 12    | 1.43   |
| 13    | 1.56   |

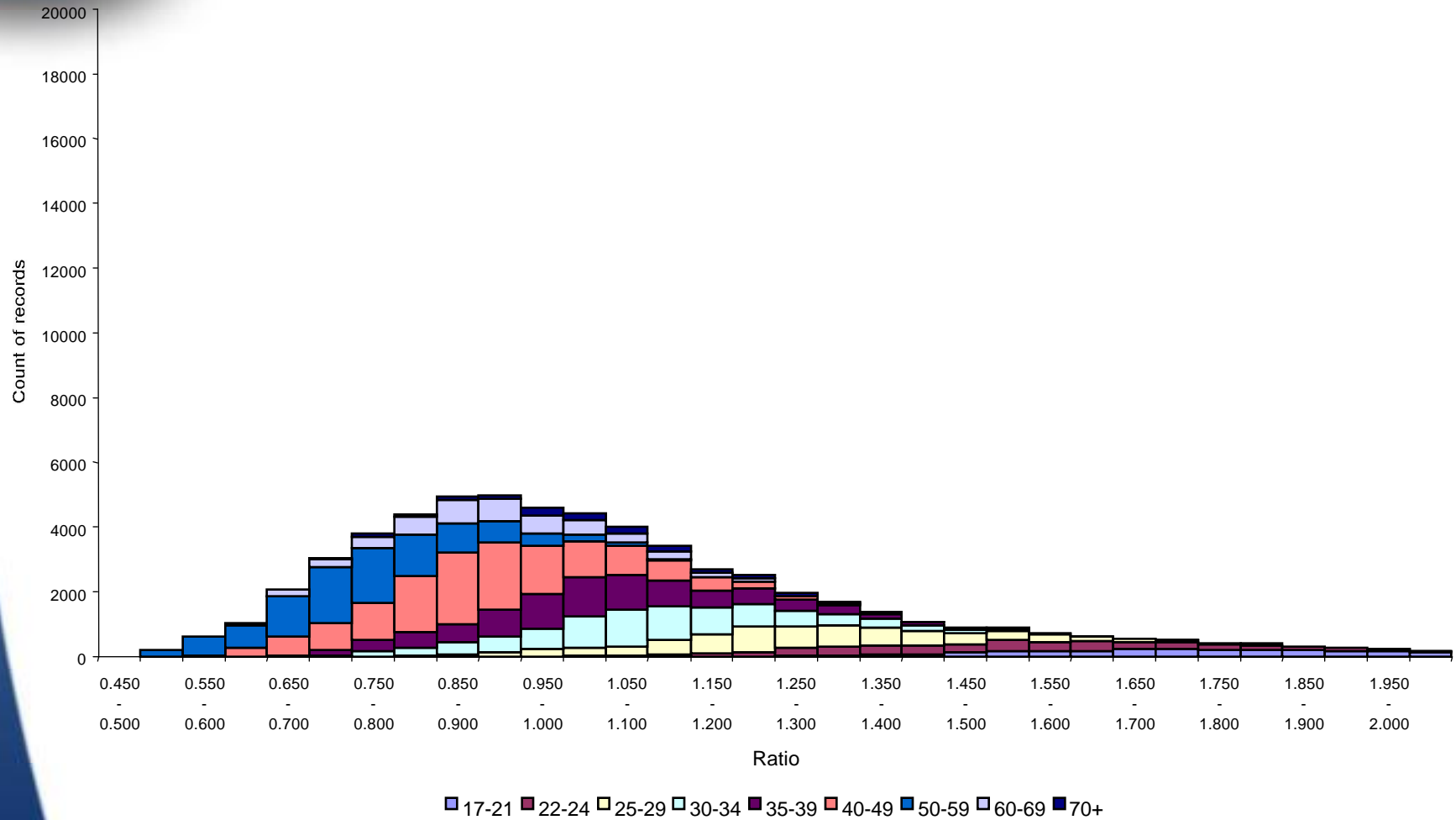
| Sex    | Factor |
|--------|--------|
| Male   | 1.00   |
| Female | 1.25   |

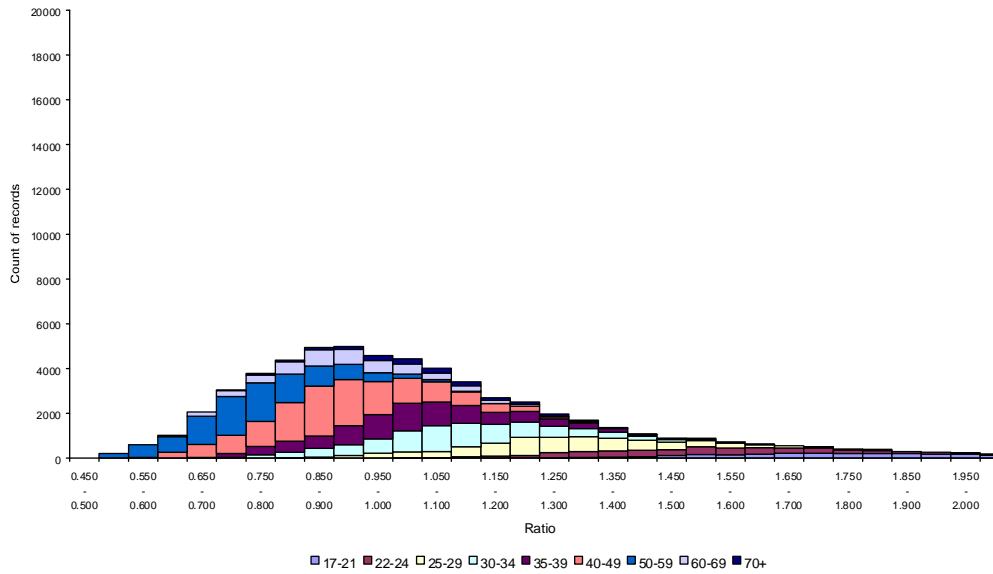
| Area | Factor |
|------|--------|
| A    | 0.95   |
| B    | 1.00   |
| C    | 1.09   |
| D    | 1.15   |
| E    | 1.18   |
| F    | 1.27   |
| G    | 1.36   |
| H    | 1.44   |

Subject to  
max +20%  
min -10%



# Example of use of moderator





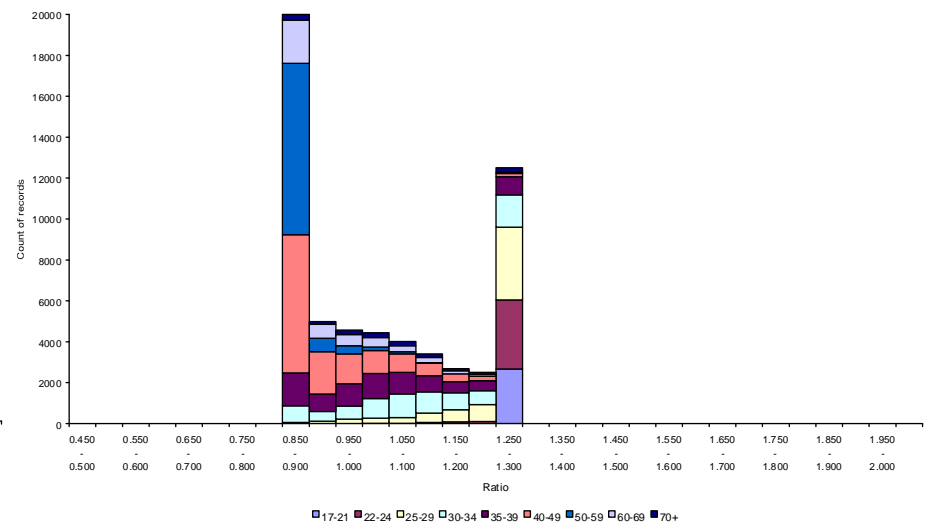
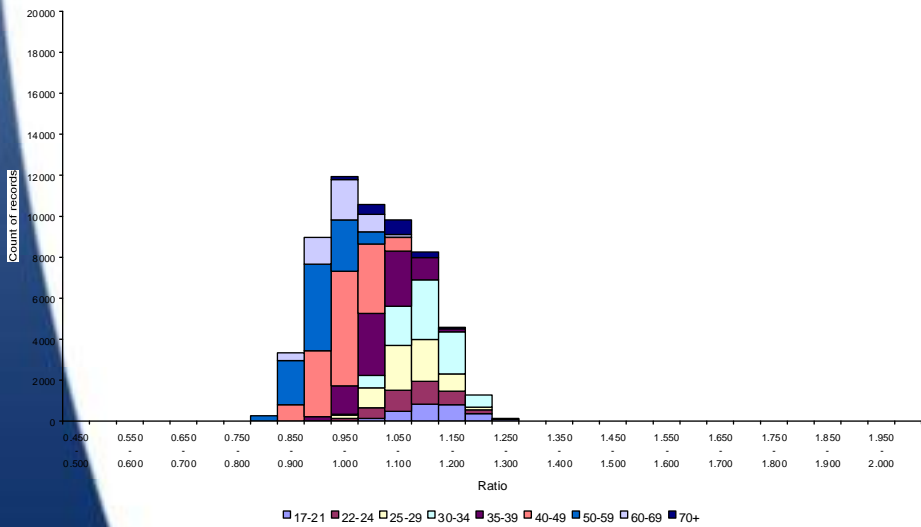
£621.50 x

| Age Factor | Group Factor | Sex Factor |
|------------|--------------|------------|
| 17         | 2.95         | M 1.04     |
| 18         | 2.81         | F 1.23     |
| 19         | 1.91         | M 0.75     |
| 20         | 1.85         | F 0.86     |
| 21-24      | 1.75         | M 0.92     |
| 24-26      | 1.64         | F 0.96     |
| 27-34      | 1.42         | M 1.01     |
| 31-34      | 1.24         | F 1.08     |
| 35-41      | 1.01         | M 1.14     |
| 41-44      | 0.85         | F 1.26     |
| 45-54      | 0.81         | M 1.39     |
| 55-64      | 0.74         | F 1.43     |
| 65+        | 0.71         | M 1.51     |

£621.50 x

| Age Factor | Group Factor | Sex Factor |
|------------|--------------|------------|
| 17         | 2.95         | M 1.04     |
| 18         | 2.81         | F 1.23     |
| 19         | 1.91         | M 0.75     |
| 20         | 1.85         | F 0.86     |
| 21-24      | 1.75         | M 0.92     |
| 24-26      | 1.64         | F 0.96     |
| 27-34      | 1.42         | M 1.01     |
| 31-34      | 1.24         | F 1.08     |
| 35-41      | 1.01         | M 1.14     |
| 41-44      | 0.85         | F 1.26     |
| 45-54      | 0.81         | M 1.39     |
| 55-64      | 0.74         | F 1.43     |
| 65+        | 0.71         | M 1.51     |

Subject to max +20%  
min -10%





# Moderators: pros/cons

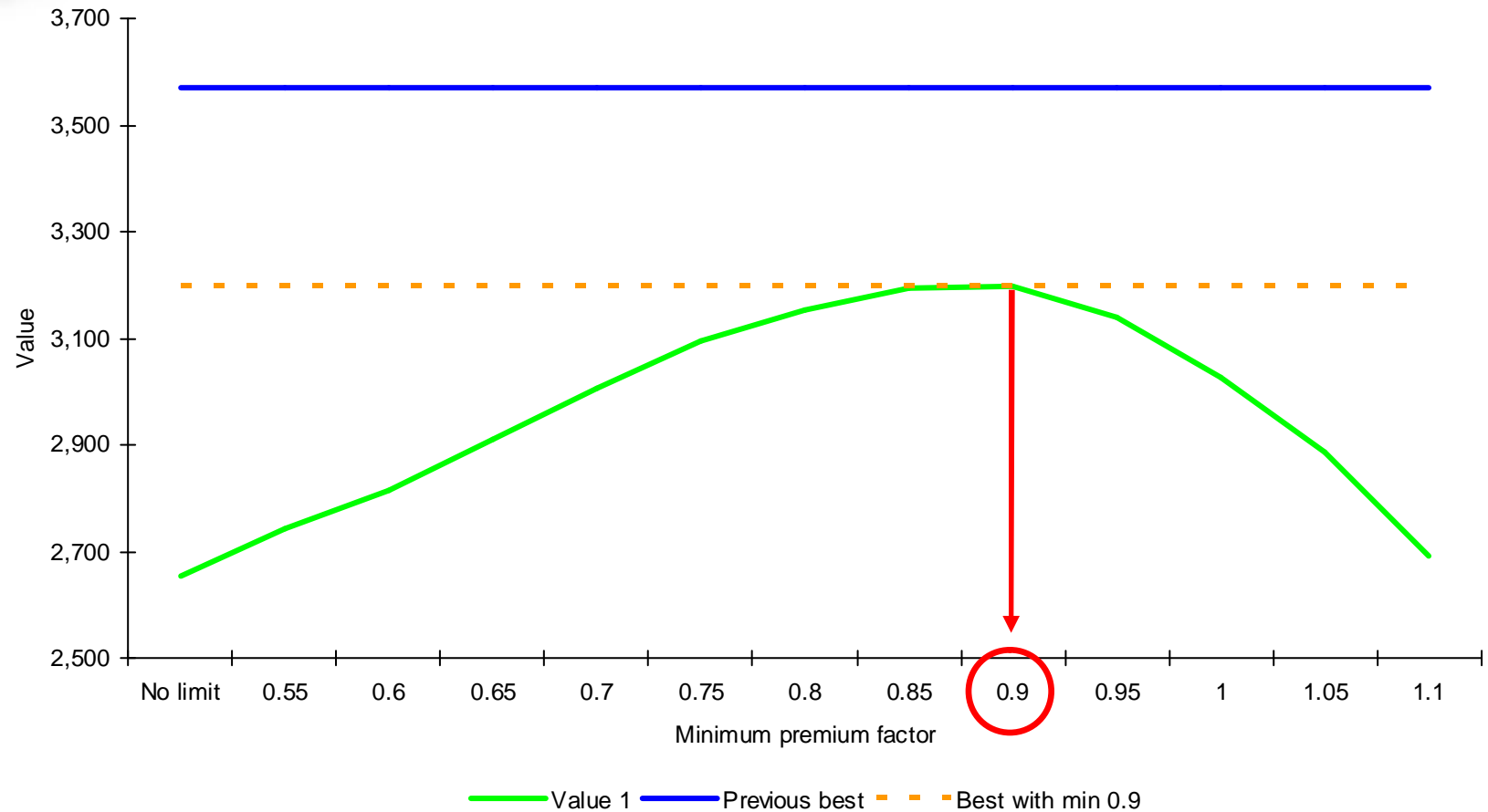
---

- Advantages of moderators include:
  - moves everyone to optimal position (subject to acceptable premium increases) more quickly
  - can take into account elasticity for the type of person in question
  - can be less detailed work required regarding underlying parameterization
  - less work required to parameterize in future
- Disadvantages
  - more onerous system requirements
  - harder to understand rating structure
  - likely to result in different quotes for renewals and new business for an identical risk



# Parameterizing the moderator

## Investigation of limiting premium decreases

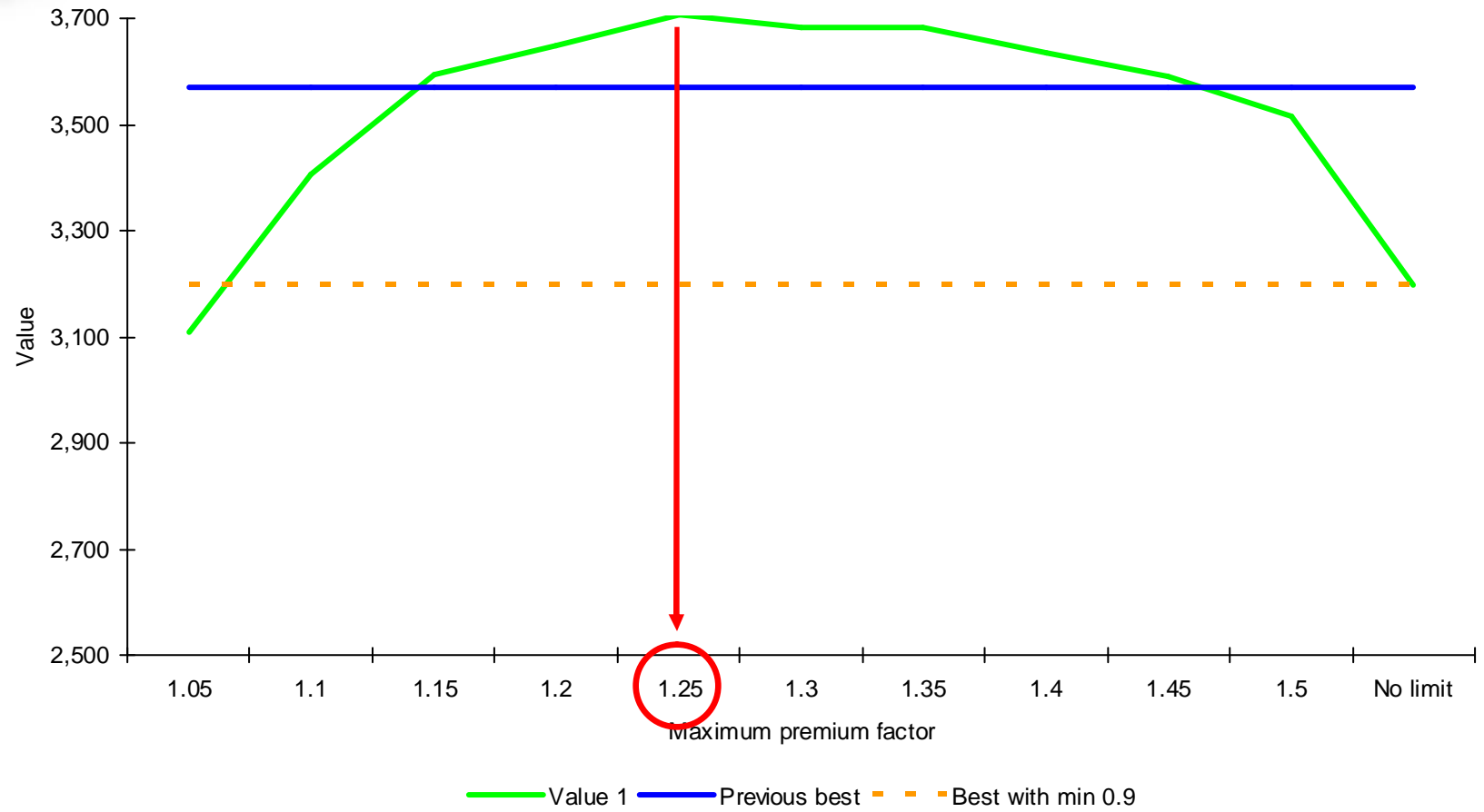






# Parameterizing the moderator

## Investigation of limiting premium increases given 10% limit on decreases





# **Types of optimization**

---

- Optimization via scenario tests, for example:
  - base rate change
  - base rate change with simple relativity change
  - moderator algorithms
- Full optimization for each individual policy at point of sale

# Full optimization

| Exposure | Age of driver | Gender | Marital status | Territory | Credit score | Earned Premium | # claims BI | Incurred losses BI | # claims PD | Incurred losses PD |
|----------|---------------|--------|----------------|-----------|--------------|----------------|-------------|--------------------|-------------|--------------------|
| 1        | 1.00          | 22     | M              | S         | 12           | 178            | 2,331       | 0                  | -           | 0                  |
| 2        | 0.65          | 39     | F              | D         | 2            | 569            | 512         | 0                  | -           | 1                  |
| 3        | 0.35          | 39     | F              | D         | 4            | 569            | 440         | 0                  | -           | 0                  |
| 4        | 1.00          | 58     | F              | M         | 6            | 715            | 968         | 0                  | -           | 0                  |
| 5        | 0.66          | 47     | M              | M         | 19           | 202            | 760         | 1                  | 16,138      | 0                  |
| 6        | 1.00          | 35     | M              | M         | 32           | 550            | 815         | 0                  | -           | 0                  |
| 7        | 1.00          | 46     | M              | S         | 17           | 420            | 1,012       | 0                  | -           | 0                  |

Optimal premium

2,651  
561  
412  
745  
699  
894  
1,242

- For each policy optimize desired success criteria
- Result is individual premium for each renewal
- For new business and mid-term changes, and if required for renewals, can approximate results with a single structure by fitting GLM to optimized individual rates



# Constrained and multiple year optimization

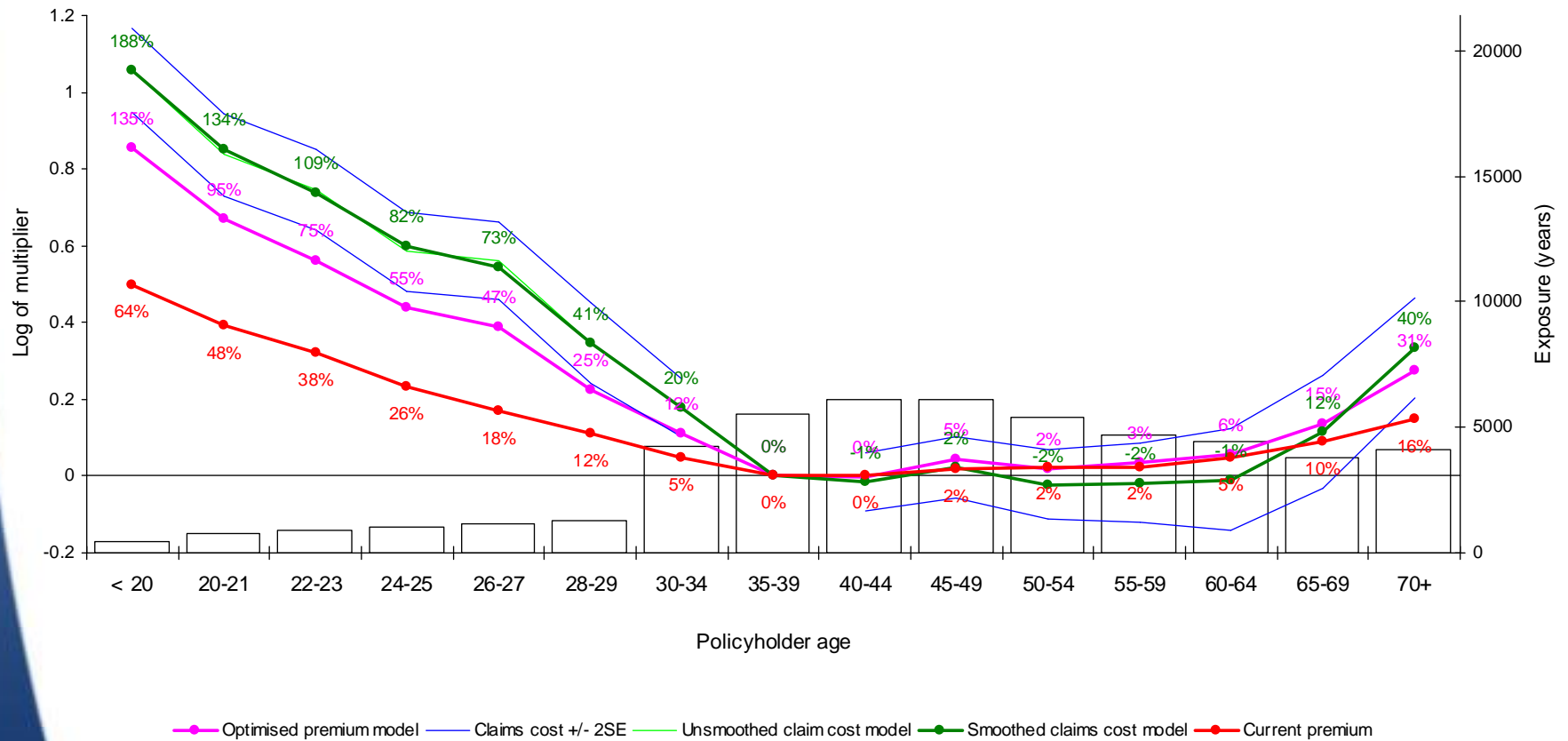
---

- Specified constraints, eg minimum business volume, can be incorporated in optimization algorithm
- For a given policy, the best action next year is interdependent with the action the following year(s) - embedded/looped optimizations allow consideration of multiple year strategies

# Full optimization

## Optimized premium

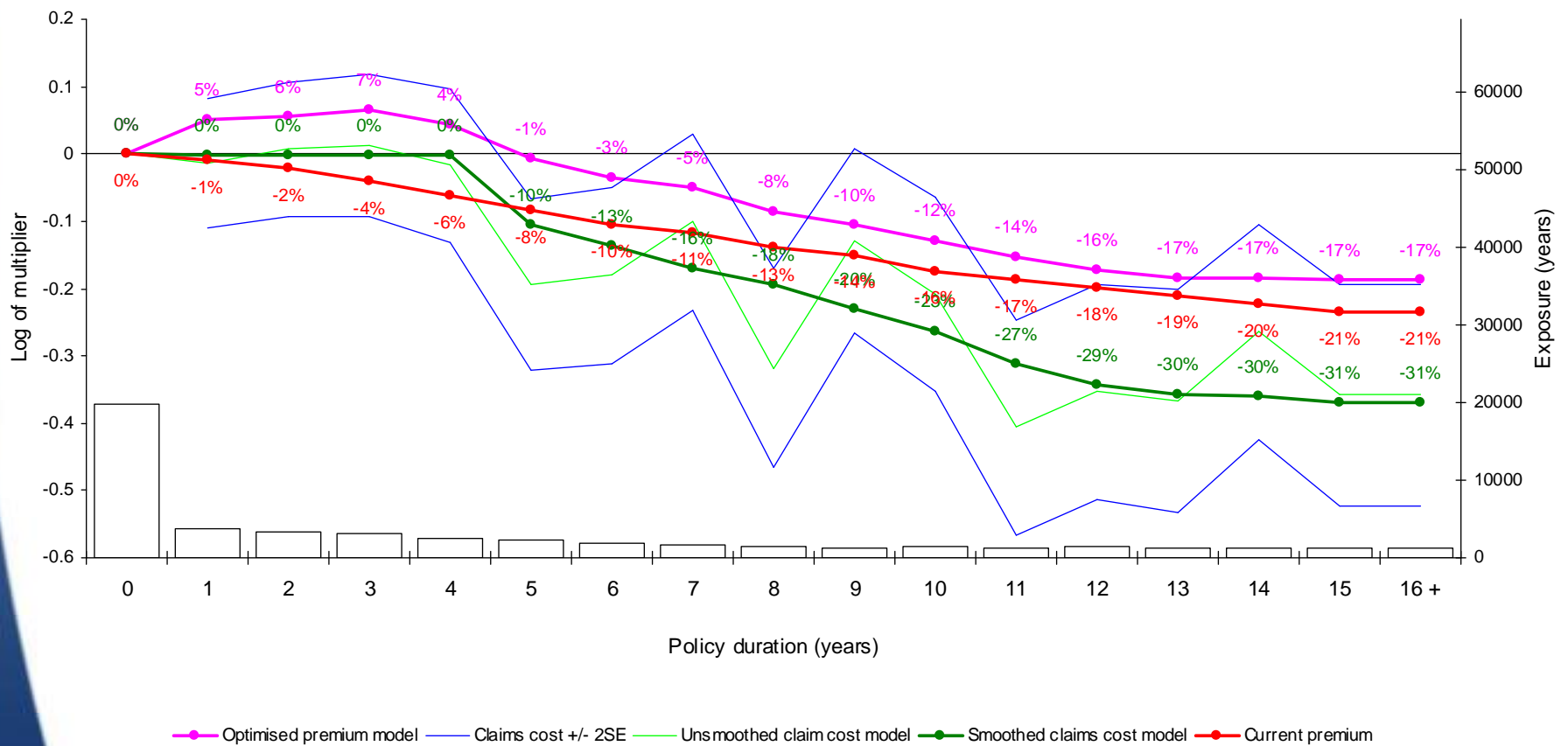
Comparison with claims model and current premium



# Full optimization

## Optimized premium

Comparison with claims model and current premium





**SPE-1  
Policyholder Retention  
and its Impact on  
Pricing**

**2006 CAS Seminar on  
Ratemaking**

**Claudine Modlin, FCAS**

**Watson Wyatt Worldwide**



[WWW.WATSONWYATT.COM](http://WWW.WATSONWYATT.COM)

