

IAA Health Section

Stochastic Modeling Webcast

Theory and Reality from an Actuarial Perspective

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Overview of Discussion

- General Concerns and Health Insurance Specific Issues
- Stochastic Modeling for Health Actuaries
- Examples
- Questions

Overview of Discussion

- Based on the recently published book by the IAA
- Stochastic Modeling: Theory and Reality from an Actuarial Perspective
- Available on the IAA website

General Concerns

- What factors should be stochastically generated?
- When should I be using stochastic models?
- When should I not use stochastic models?
- Are there alternatives to stochastic models?
- What are the disadvantages of stochastic models?

General Concerns

When should I be using stochastic models?

- Are these required by regulation or professional guidelines?
- Do we need a better understanding of the effects of extreme outcomes?
- Do we need a better understanding of the tail risk or risks in general?
- What is the probability of an event?
- What is the probability of ruin?
- Are certain risk measures needed for reporting?

General Concerns

When should I **not** use stochastic models?

- Can you calculate a probability distribution?
- Can you calibrate the model?
- Can you validate the model?

General Concerns

Are there alternatives to stochastic models?

- Stress Testing/Scenario Testing
- Static Factors/PADs/MADs
- Range testing

General Concerns

What are the disadvantages of stochastic models?

- “Black Box”
- Inappropriate distributions
- Inappropriate parameters
- Improper calibration
- Validation – Beware of False Positives
- Model size
- Computer power

General Concerns

Model Calibration

- Two General Approaches
 - Calibration to Historical Experience
 - Calibration to Current Market Conditions
- Considerations
 - Does the model track to expected assumptions?
 - Reflect expectations today?
 - Experience period
 - Range of possible outcomes?
 - Extremes

General Concerns

Model Calibration

- Calibration to Historical Experience
 - Can you create a distribution?
 - Expected
 - Correlations
 - Volatility
 - Mean reversion

General Concerns

Model Calibration

- Calibration to Current Market Conditions
 - Observed market prices or conditions
 - Closed form formula
 - Market consistent results

General Concerns

Model Validation

- How do I validate a model?
 - Cellular checking
 - Reasonableness review
 - Assumption review
 - Formula testing
 - Calibration review
 - Distribution of outcomes

General Concerns

Model Validation

- Documentation Review
 - Source of data
 - Experience period
 - Testing
 - Audit/Checking/Peer Review

General Concerns

Model Peer Review

- Reviewed for Accuracy?
- Credible data?
- Model and parameter development?
- Correlations?
- Testing
- Validation

General Concerns

Model Communication

General Ideas

- Average
- Outliers – Worst Case and Best Case
- Specific Scenario
- Type of Audience

General Concerns

Model Audit

Parameter Development

- Source of Information
- Development of Assumptions
- Development of Correlations
- Expected Assumptions

General Concerns

Model Audit

Stochastic Generator Analysis

- Cell Checking
- Review of Distributions, range of outcomes, extreme cases
- Correlation checking
- New set of scenarios produce similar results

General Concerns

Health Specific Issues

- Identification of All Factors
- Identification of Key Material Factors

Example of Long Term Health Insurance Risk Model



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Risk Factors of Interest to Health Actuaries

General Categories of Risk Factors of Interest to Health Actuaries:

- Economic Factors
- Demographic Factors
- Morbidity/Claims Factors

All categories include assumptions that could be modeled stochastically

Risk Factors of Interest to Health Actuaries

- Economic Factors Include:
 - Interest Rates
 - Unemployment
 - Inflation

Risk Factors of Interest to Health Actuaries

- Demographic Factors Include:
 - Policy Terminations (Claim termination considered separately)
 - Shock Lapses
 - Mortality

Risk Factors of Interest to Health Actuaries

- **Morbidity/Claims Factors Include:**
 - Claim Costs (or Incidence and Severity)
 - Morbidity Improvement
 - Medical Trend

Economic Risk Factors

- Interest Rates
 - Many commonly used interest rate generators are available.
 - Approach to selecting an interest rate generator:
 - How will the projection be used?
 - Should scenarios be “real-world” or “risk-free”?
 - Is mean-reversion incorporated in the model?
 - Do you want to model stochastic volatility?
 - Evaluate arbitrage-free properties of the model
 - If appropriate, validate scenarios produced by the model against observable market prices or conditions.

Commonly used Interest Rate Generators: Equilibrium Models

Commonly used Interest Rate Generators - Equilibrium Models

Model	Formulaic Description	Comments
Rendleman and Bartter Model	<ul style="list-style-type: none"> ■ $dr = \mu r dt + \sigma r dz$ 	<ul style="list-style-type: none"> ■ μ and σ constant ■ r follows geometric Brownian motion ■ μ is a drift component. σ is a volatility parameter. ■ No mean reversion ■ Negative interest rates are possible ■ Drift is time-independent
Vasicek Model	<ul style="list-style-type: none"> ■ $dr = a(b-r) dt + \sigma dz$ 	<ul style="list-style-type: none"> ■ Incorporates mean reversion through the term $a(b-r)$ ■ The σdz term incorporates stochastic element ■ Negative interest rates are possible ■ Drift is time-independent
Cox, Ingersoll and Ross Model	<ul style="list-style-type: none"> ■ $dr = a(b-r) dt + \sigma \sqrt{r} dz$ 	<ul style="list-style-type: none"> ■ Very similar to Vasicek model, except for the presence of \sqrt{r} in the second term ■ The \sqrt{r} exists in order to preclude negative interest rates ■ Drift is time-independent

Note: dz is a Wiener process satisfying the following two properties:

- (1) $dz = Z (dt)^{(1/2)}$, where Z is standard normally distributed
- (2) Values of dz for separate time intervals are independent

Commonly used Interest Rate Generators: No-Arbitrage Models

Commonly used Interest Rate Generators - No-Arbitrage Models

Model	Formulaic Description	Comments
Ho-Lee Model	<ul style="list-style-type: none"> ■ $dr = \theta(t) dt + \sigma dz$ ■ $\theta(t) = F_t(0,t) + \sigma^2 t$ 	<ul style="list-style-type: none"> ■ σ constant ■ $\theta(t)$ term exists to calibrate model to initial term structure ■ Drift is time-dependent ■ $F_t(0,t)$ is the partial derivative with respect to t of the forward rate for a maturity t. This term helps ensure that future short rates move in line with initial term structure
Hull-White Model	<ul style="list-style-type: none"> ■ $dr = [\theta(t) - ar] dt + \sigma dz$ ■ $\theta(t) = F_t(0,t) + aF(0,t) + (\sigma^2/2a)(1-e^{-2at})$ 	<ul style="list-style-type: none"> ■ Effectively, a Ho-Lee model with mean-reversion

Note: dz is a Wiener process satisfying the following two properties:

- (1) $dz = Z (dt)^{(1/2)}$, where Z is standard normally distributed
- (2) Values of dz for separate time intervals are independent

Economic Risk Factors

- Inflation/Unemployment Models
 - Purchasing Power Parity
 - ARIMA (Autoregressive Integrated Moving Average Models)
 - Phillips Curve Models

Economic Risk Factors

- Purchasing Power Parity
 - In a no-arbitrage world, forward exchange rates are completely determined by:
 - The current foreign exchange rate
 - Projection of US-denominated risk-free interest rates
 - Projection of Fx-denominated risk-free interest rates
 - Using this fact, we can model inflation stochastically without adding additional complexity to the model.
 - Key advantage: your model will be internally consistent and arbitrage-free

Economic Risk Factors

- Purchasing Power Parity
 - Key Point:
 - Arbitrage is avoided if and only if:
 - Investing \$1 in USD for 5-years at the risk-free rate is expected (on a risk-free basis) to be equivalent to:
 - Converting \$1 to GBP
 - Investing for 5-years at the risk-free rate
 - Converting the cumulative value to USD at the end of 5-years

Economic Risk Factors

- Purchasing Power Parity
 - Example

Time	US Dollar	GBP	Spot	No-Arbitrage
	Risk-Free	Risk-Free	Exchange	Forward
	Forward Rates	Forward Rates	Rate (USD/GBP)	Rate (USD/GBP)
0			1.4000	
1	0.75%	5.00%		1.3433
2	1.00%	5.00%		1.2922
3	1.25%	5.00%		1.2460
4	1.50%	6.00%		1.1931
5	1.50%	6.00%		1.1425

Economic Risk Factors

- Purchasing Power Parity
 - Conclusions:
 - Over 5-years, USD has appreciated relative to GBP
 - This is due to relatively high GBP risk-free rates relative to USD risk-free rate
 - USD Inflation has been low (relative to GBP)
 - The implied forward exchange rate can be used to develop inflation relativities
 - Doing so requires some professional judgment, but this technique can be quite useful if you compare your currency to a currency where the central bank is known to target a certain inflation range (and does so credibly).

Economic Risk Factors

- ARIMA Models
 - Many different variations of this approach using time-series/regression models
 - Both continuous-time and discrete-time models exist

Economic Risk Factors

- Phillips Curve Models
 - Basic Idea: there is a trade-off between inflation and unemployment (i.e., the “Phillips Curve”)
 - If your model already incorporates a stochastic projection of unemployment, you can make use of a Phillips curve to extract the implied level of inflation given the projected unemployment
 - Alternatively, your inflation model (using either Purchasing Power Parity or an ARIMA model) can be used to project unemployment (again through use of the Phillips curve)

Demographic Risk Factors

- Policy terminations and Shock Lapses
 - Generally modeled using commonly available distributions:
 - Normal distribution is the most commonly used
 - Establish a deterministic lapse assumption, which may vary by duration
 - Model additions/subtractions to the lapse rate as a stochastic, normally distributed random variable with mean 0 and standard deviation, (Σ)
 - Σ generally estimated based on historical data (for the block of business being modeled or from general industry experience), but may involve element of judgment or provision for adverse deviation
 - Shock Lapses
 - The shock is sometimes modeled as a stochastic, normally distributed random variable (similar to routine lapses)
 - Other models may express shock lapse as a randomly-generated percentage of some other observed value (e.g., x% of rate increases)

Demographic Risk Factors

- Mortality
 - Often a smaller concern for health products (relative to traditional life products)
 - Several approaches exist:
 - Normal distribution probably the most widely-used
 - May fail to capture “regime changes” (e.g., pandemics, terrorist attacks or other “1 in x-hundred year events”)
 - IAA Stochastic Modeling text presents a sophisticated 3-component approach
 - Baseline Model
 - Disease Model
 - Terrorism Model

Morbidity/Claims Factors

- Claim Incidence
 - Commonly-used approaches:
 - “Pure” Monte Carlo
 - Normal Distribution
 - Log-Normal Distribution
 - Poisson Distribution

- Claim Severity
 - Stochastic models are generally constructed off of pre-existing continuance curves/tables
 - Stochastic generation of curves is generally complex and (often) unnecessary
 - Approach:
 - Generate random number from some distribution (e.g., uniform)
 - Use the random number to “choose” the severity of a particular claim from the appropriate continuance curve
 - Calculate total claim based on “chosen” severity

Morbidity/Claims Factors

- Comments on Medical Trend
 - Medical trend may be linked to inflation
 - Can model medical trend stochastically as:
 - Inflation + “Real Cost Spread”
 - Some choose to model the “Real Cost Spread” stochastically, as well
 - Advantage: Model is internally consistent
 - Disadvantage: May desire a more “scientific” approach to determine the “Real Cost Spread.”

Example: Long Term Care Product

- Illustrative Example for “Long-Tailed” Health Product:
 - Long Term Care Policy
 - Policy Characteristics:
 - 5-year Benefit Period
 - \$100/day benefit
 - No inflation protection or other special riders
 - Annual premium = \$1,000
 - Policyholder Data:
 - Female
 - Attained age 70 as of the valuation date
 - Purchased the policy 10 years ago (i.e., is in policy duration 10)
 - Other Information:
 - Policy is expected to have a rate increase of 50% in projection year 3
 - We will do a 5-year projection of premium and incurred claims for a single cell, under a single scenario

Claims Modeling - Incidence

- Claim Incidence
 - Example (Poisson Distribution for Claim Incidence)
 - Estimated parameter for the Poisson Distribution
 - $\lambda = 0.3$

Probability of a Claim (Poisson)		0.2222
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Projection Year	Random Number Generated	Claim Incidence
1	0.901	0
2	0.930	0
3	0.164	1
4	0.363	0
5	0.007	1

Claims Modeling - Continuance

- Claim Severity (Continuance)
 - Actuary has estimated claim continuance curve as follows:

Illustrative Claim Continuance Curve									
Claim Duration (Years)	0.5	1.5	2.0	2.5	3.0	3.5	4.0	4.5	5.0
Probability of Survival	98.00%	90.00%	80.00%	60.00%	30.00%	15.00%	7.50%	3.75%	0.00%

Claims Modeling - Continuance

- Claim Severity (Continuance)
 - Actuary selects a uniform distribution to stochastically model claim continuance
 - Following considers claim continuance only; we will combine incidence and severity next:

Projection Year	Random Number Generated	Random		Modeled Claim Severity (Years)
		High	Low	
1	0.213	30.00%	15.00%	3.29
2	0.001	3.75%	0.00%	4.99
3	0.699	80.00%	60.00%	2.25
4	0.920	98.00%	90.00%	1.25
5	0.048	7.50%	3.75%	4.35

- For random numbers in between points on the claim continuance curve, we use linear interpolation

Claims Modeling – Incurred Claims

- Incurred Claims

- To model total claim costs, we combine claim incidence and claim severity from the previous slides, as follows:

Projection Year	Claim Incidence	Claim Severity (Years)	Annual Benefit	Total Incurred Claim (Undiscounted)
1	0.000	3.29	36,500	-
2	0.000	4.99	36,500	-
3	1.000	2.25	36,500	82,191
4	0.000	1.25	36,500	-
5	1.000	4.35	36,500	158,947

Claims Modeling – Incurred Claims

■ Policy Termination Rates

- We use the “stochastic addition” method discussed earlier
- The additive term is normally distributed with mean = 0 and standard deviation = 5%

Sigma 5.00%

Policy Year	Projection Year	Deterministic Voluntary Lapse	Random Number	Stochastic Adjustment to Lapse	Total Policy Termination Rate
10	1	5.0%	0.317	-2.4%	2.6%
11	2	5.0%	0.663	2.1%	7.1%
12	3	5.0%	0.812	4.4%	9.4%
13	4	4.0%	0.213	-4.0%	0.0%
14	5	3.0%	0.049	-8.3%	0.0%

Note: The Voluntary Lapse is assumed to include mortality; Mortality is not modeled stochastically in this example.

Claims Modeling – Incurred Claims

▪ Rate-Increase Shock Lapses

– Modeled as follows:

- Deterministic assumption that shock lapse = 10% of Rate Increase Amount
- In the stochastic projection, the deterministic assumption is multiplied by a stochastic adjustment that is normally distributed with mean = 1, standard deviation = 25%

Projection Year	Deterministic		Random Number	Stochastic Adjustment to Lapse	Projected Shock Lapse
	Rate Increase	Shock Lapse			
1	0.0%	0%	0.317	0.881	0.0%
2	0.0%	0%	0.508	1.005	0.0%
3	50.0%	5%	0.632	1.084	5.4%
4	0.0%	0%	0.618	1.075	0.0%
5	0.0%	0%	0.832	1.240	0.0%

Claims Modeling – Incurred Claims

- Cash Flow Projection Detail

- Based on the components developed on the previous slide, we have the following cash flow projections:

Time	Policy Count	Paid Premium	Incurred Claims
0	1.000		
1	0.974	987	-
2	0.905	939	-
3	0.775	1,260	82,191
4	0.775	1,162	-
5	0.775	1,162	158,947

- Example ignores accounting implications, expenses, etc.

Example of Short Term Medical Insurance Risk Model



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Stochastic Risk Models for Short Term Medical Insurance

- What is Stochastic Modeling Used for in Short Term Medical Insurance?
 - Claim Level Estimation
 - Surplus Requirements (Economic Capital)
 - Distribution of Medical Loss Ratios
 - Stop Loss Rating
 - Other

Stochastic Risk Models for Short Term Medical Insurance

- What are Major Risks to Insurance Companies Selling Short-Term Medical Insurance Products?
 - Rating Parameter Adequacy
 - Regulatory Issues / Delays
 - Catastrophic Events
 - Expense Recoupment
 - Other

Stochastic Risk Models for Short Term Medical Insurance

- Is Stochastic Modeling Necessary to Establish Surplus Requirements?
 - No
 - However, it is superior to deterministic models that involve projection of a limited set of likely scenarios
 - Also, it is superior to peer group analysis
 - Where conclusions are drawn from companies with “similar” characteristics
 - Stochastic models allow for simultaneous consideration of multiple risk factors and ranges of possible outcomes

Stochastic Risk Models for Short Term Medical Insurance

- What are some of the considerations in developing a Stochastic Model?
 - Establish risk level – high likelihood, sufficiency, virtual certainty – corresponding to 90th, 95th, 98th percentiles
 - Determine risks to include in model
 - Develop distributions of outcomes for each risk, based on ranges of potential outcomes
 - Some risks can be easily measured and parameterized
 - Other risks may be more subjective and harder to define
 - Interdependent risks need to be evaluated

Stochastic Risk Models for Short Term Medical Insurance

- How are Stochastic Models Tested to Ensure Meaningful Results?
 - Sufficient number of iterations are run to ensure stability of result
 - Underlying distributions are calibrated to observed data history
 - Model results are validated by comparison to other independent approaches or results

Stochastic Risk Models for Short Term Medical Insurance

- Examples of Stochastic Modeling Used in Short Term Medical Insurance:
 - Risk Associated with Catastrophic Events
 - Already established claim risk model
 - Need to evaluate additional risk due to pandemics
 - Develop Optimal Surplus Target Range
 - For non-profit Short Term Medical policies sold by a stand-alone health insurance company

Risk Associated with Pandemics

- Claims Trend



Risk Associated with Pandemics

■ Pandemic Cost Component

Event	Probability	Outcome
• No Pandemic	• 98.0%	• 0% Additional Cost
• Mild Pandemic	• 1.0%	• 2% Additional Cost in Year
• Major Pandemic (duration 1 year)	• .75%	• 14% Additional Cost in Year
• Major Pandemic (duration 2 years)	• .25%	• 14% Additional Cost in 1 st Year and 2% Residual Cost in 2 nd Year

Risk Associated with Pandemics

- Premium Development

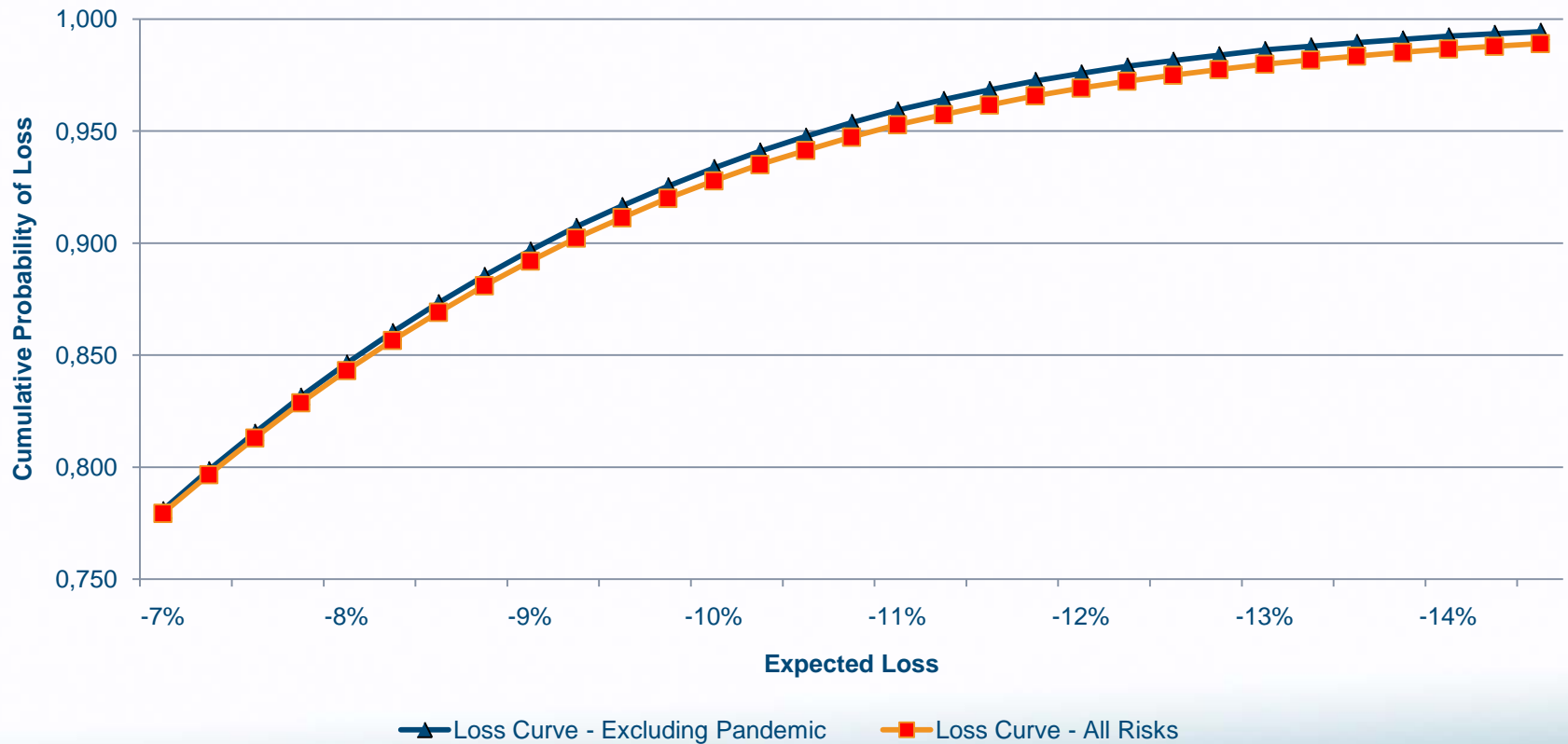


Risk Associated with Pandemics

- Recovery of Losses from Pandemic
 - Recovery of Loss, if any, over 5 years after event
 - Potential for not being able to recover losses due to MLR limitations under Federal Health Care Reform
- Distribution of results with and without pandemic risk

Risk Associated with Pandemics

- Illustrative Distribution of Losses



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Optimal Surplus

- Analytical Approach
 - Approach has been refined over time
 - Best described as a variant of Enterprise Risk Management
 - Involves the use of a model to reflect company-specific financial situation and risk exposure:
 - Forward looking projection of potential outcomes
 - Based on historical and anticipated loss distributions
 - Detailed approach to development and documentation of assumptions
 - Externally peer reviewed
 - Approach found to be reasonable and acceptable

Optimal Surplus

- Analytical Approach
 - Basic Approach to the Evaluation:
 - Analysis of historical underwriting loss cycles
 - > Company-specific loss history
 - > Comparison (by relative size) set of Insurers
 - Identify and evaluate major risks
 - > Develop risk and contingent distributions
 - > Monte Carlo simulation of loss cycles
 - Identify reserve target thresholds and criteria
 - > Surplus target goals
 - > Loss cycle provisions to satisfy goals

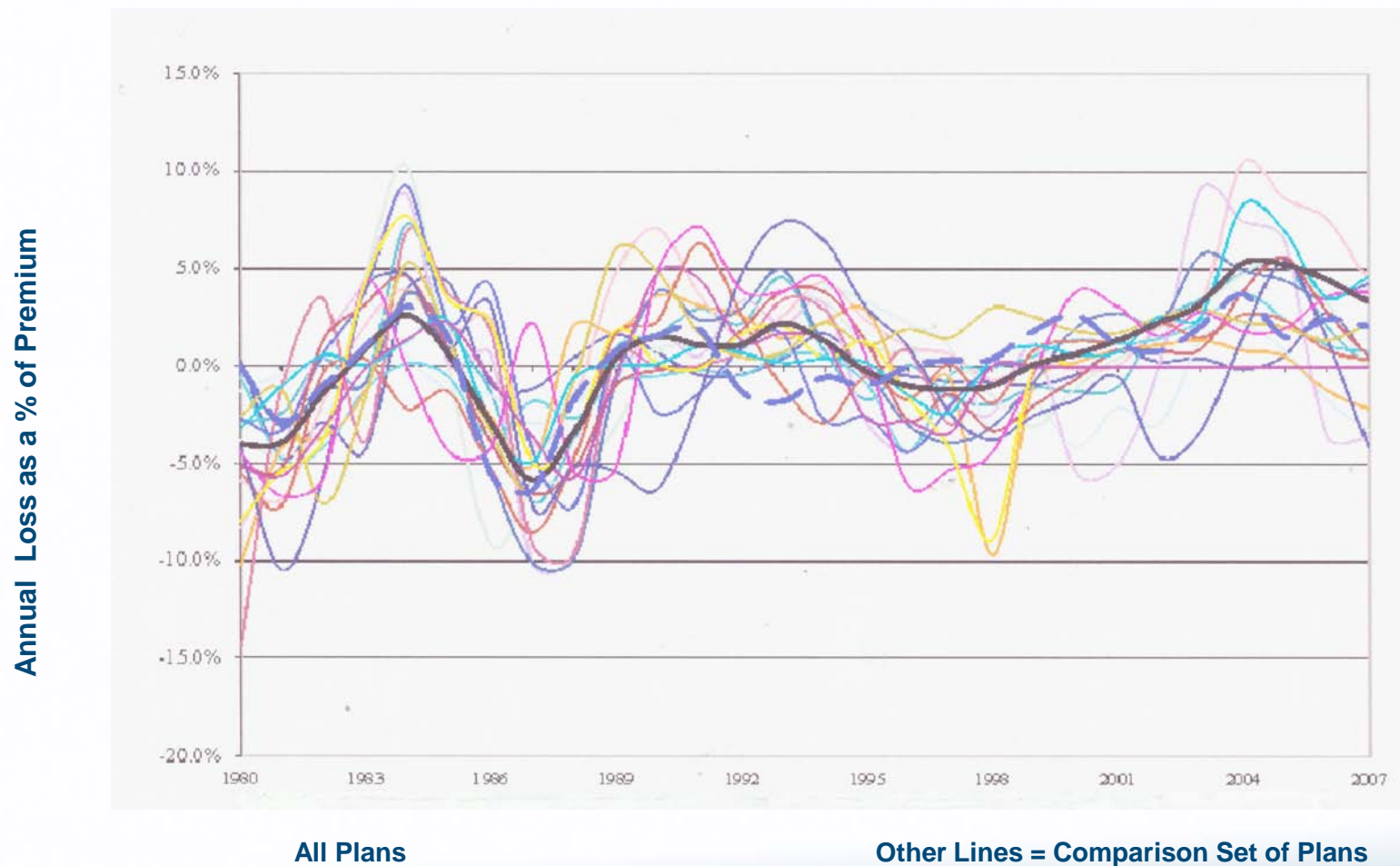
Optimal Surplus

- Analytical Approach
 - Basic Approach to the Evaluation:
 - Develop and initialize pro forma projection model
 - > Pro forma assumptions for company
 - Develop reserve target range
 - > “Stress test” loss cycles against reserve thresholds
 - > Develop reserve levels necessary to withstand cycle losses

Optimal Surplus

- Analytical Approach
 - Basic Approach to the Evaluation:
 - Analysis of historical underwriting loss cycles

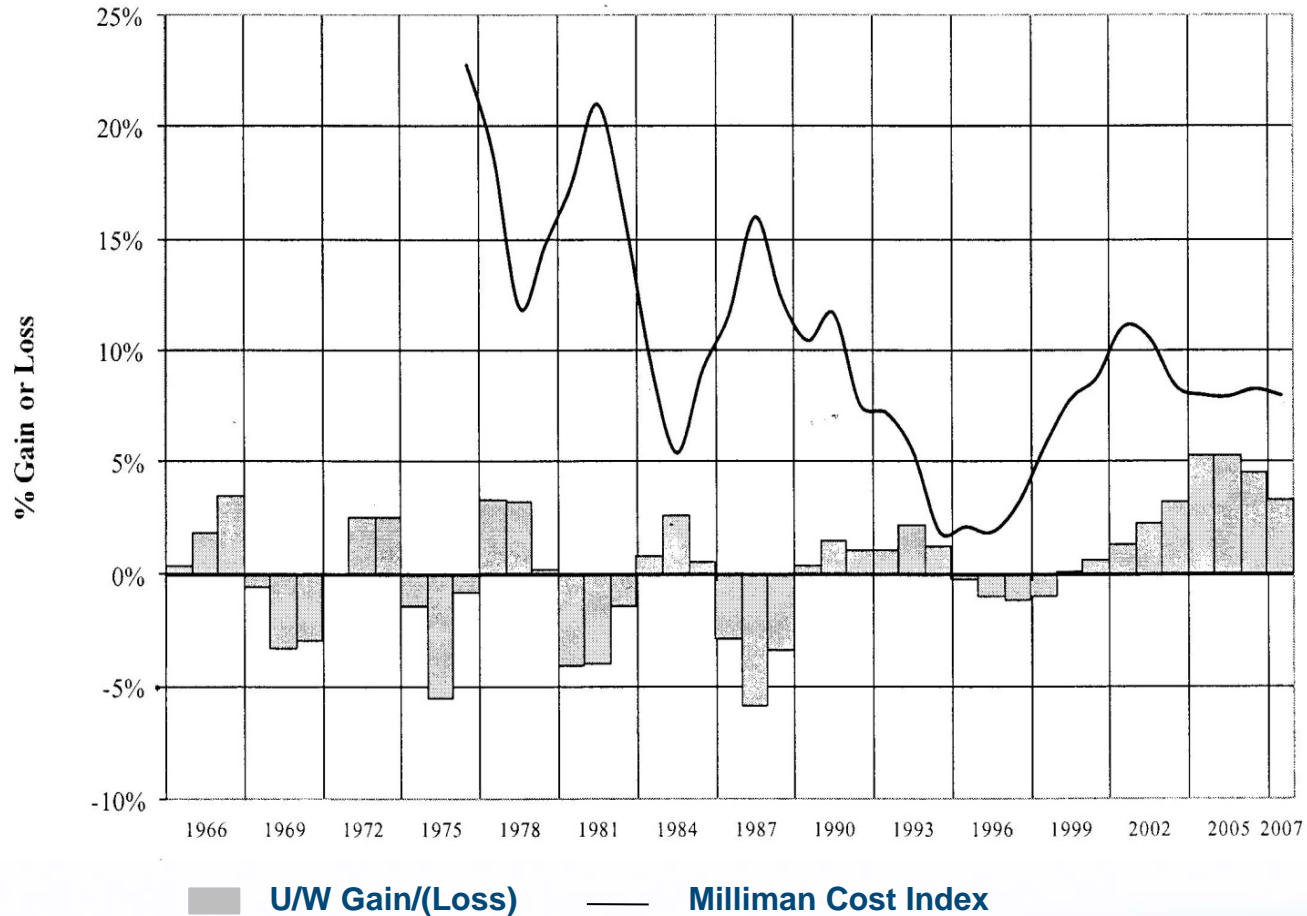
Underwriting Gain/(Loss) for Selected Insurers



All Plans

Other Lines = Comparison Set of Plans

Average Underwriting Gain / (Loss) for a Sample of Health Insurers



Optimal Surplus

- Analytical Approach

- Basic Approach to the Evaluation:

- Identify and evaluate major risks

- Develop risk and contingency distributions

- > Rating adequacy and fluctuation

- > Unpaid claims and other estimates

- > Portfolio risks (equity market, interest rates, impairment)

- > Overhead expense recovery and fee income risk

- > Other business risks, including self-funded

- > Catastrophic events, including Health Care Reform

- > Provision for unidentified development and growth

- > Other factors

Optimal Surplus

- Analytical Approach
 - Basic Approach to the Evaluation:
 - Identify reserve target thresholds and criteria
 - Surplus target goals
 - > Early warning monitoring threshold avoidance (375% RBC)
 - > Company Action Level (200% - 300% RBC)
 - > Loss of trademark avoidance (200% RBC)
 - Loss cycle provisions established to satisfy goals – for example:
 - > Early warning – 90th to 95th percentile of Monte Carlo simulated loss cycles
 - > CAL – 95th percentile of simulated loss cycles
 - > Loss of trademark – 98th percentile of simulated losses

Optimal Surplus

- Analytical Approach
 - Basic Approach to the Evaluation:
 - Develop and initialize pro forma projection model
 - > Pro forma assumptions for company
 - Generally based on company forecast
 - Normalized assumptions to reflect longer term view of operations
 - GAAP to Stat adjustments

Optimal Surplus

- Analytical Approach
 - Basic Approach to the Evaluation:
 - Develop reserve target range
 - > “Stress test” loss cycles against reserve thresholds
 - Use Pro Forma model to project company balance sheet
 - Based on various starting surplus amounts
 - Tested under a range of forecast projection assumptions
 - Ending surplus target goals
 - > Develop reserve levels necessary to withstand cycle losses
 - Expressed as ranges of RBC

Optimal Surplus

Recap of Resulting Surplus Ranges

- 11 Distinct Studies in Recent Years
- Circumstances of Individual Companies Different
- Range of Results 650% - 1250% of ACL-RBC

IAA Health Section

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Theory and Reality from an Actuarial Perspective

Questions ??

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