

Benchmarking Unpaid Claim Estimates

Benchmark: A standard, or a set of standards, used as a point of reference for evaluating performance or level of quality. Benchmarks may be drawn from a firm's own experience, from the experience of other firms in the industry, or from legal requirements such as environmental regulations.

Source: businessdictionary.com

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Benchmarking Unpaid Claim Estimates

Have you ever calculated an estimate of unpaid claims?

P&C (General) Insurance, any LOB or segment

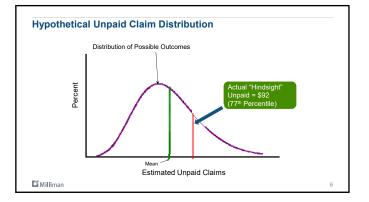
For any reason, reserves, pricing, ERM, etc.

Have you ever used a benchmark to help with your estimated unpaid claims or range of estimates?

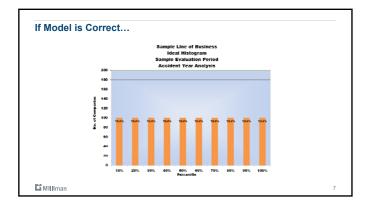
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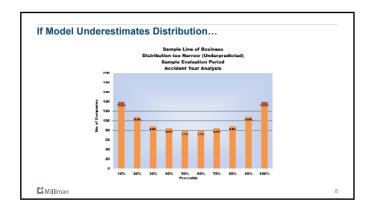
1	Background
2	Analysis Summary
3	Model Limitations
4	Model Projections – Are they Unbiased?
5	Proposed Adjustments
6	Conclusions
7	Claim Variability Guidelines



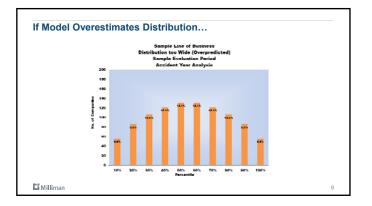


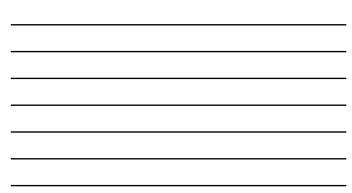












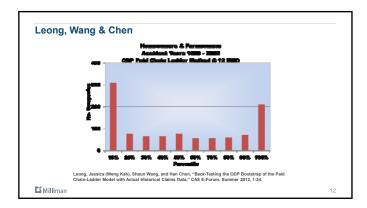


Meyers & Shi

"...study suggests that there might be environmental changes that no single model can identify."

"If this continues to hold, the actuarial profession cannot rely solely on stochastic loss reserve models to manage its reserve risk."

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Leong, Wang & Chen

Leong, Wang & Chen

percentiles."

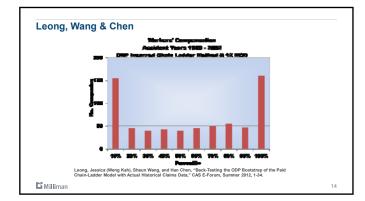
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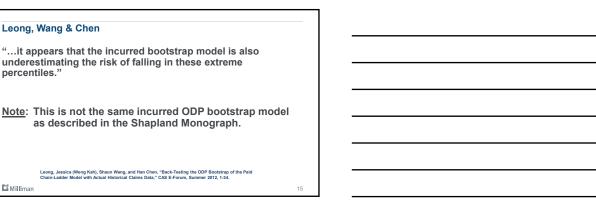
"...the popular ODP bootstrap of the paid chain-ladder method is underestimating reserve risk."

"...the bootstrap model does not consider systemic risk, or, to put it another way, the risk that future trends in the claims environment – such as inflation, trends in tort reform, legislative changes, etc. – may deviate from what we saw in the past."

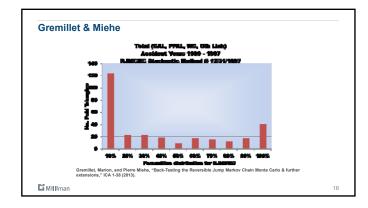
Leong, Jessica (Weng Kah), Shaun Wang, and Han Chen, "Back-Testing the ODP Bootstrap of the Paid Chain-Ladder Model with Actual Historical Claims Data," CAS E-Forum, Summer 2012, 1-34.

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Gremillet & Miehe

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"...it is core to have adjustments by actuaries prior to running the stochastic methods 'automatically.' " "Actuary in the box" dream for stochastic reserves valuation not yet happening

> Gremillet, Marion, and Pierre Miehe, "Back-Testing the Reversible Jump Markov Chain Monte Carlo & further extensions." ICA 1-38 (2013).

Background Communication Issues

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Communication Issues

- Intended audience
- Intended use of the work product
- Measurement objective
- Reliability of the estimates
- Disclosures

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ASOP 43

- <u>Purpose or Use of the Unpaid Claim Estimate</u> The actuary should identify the intended purpose or use of the unpaid claim estimate.
- Who will be using the work product?
- What is their training and experience?
- How do they intend to use it?

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Examples of Intended Uses

- Support for a Statement of Actuarial Opinion
- = M&A
- ERM risk assessment, capital modeling, ORSA
- Internal strategic planning
- SEC filings

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ASOP 43

3.3 <u>Scope of the Unpaid Claim Estimate</u> The actuary should identify the following:

a. the intended measure of the unpaid claim estimate;

 Examples of various types of measures for the unpaid claim estimate include, but are not limited to, high estimate, low estimate, median, mean, mode, actuarial central estimate, mean plus risk margin, actuarial central estimate plus risk margin, or specified percentile.

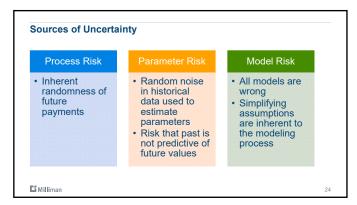
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Basis of Presentation

- Standard deviation
- Coefficient of variation
- Probability distribution
- Probability levels / Confidence Levels / Percentiles

Arguably satisfies the letter of the law, but the spirit of the law too?

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Independent Risk	Internal Systemic Risk	External Systemic Risk
Inherent randomness of future payments Random noise in historical data used to estimate parameters	 Simplifying assumptions inherent to the modeling process Unconscious biases of the reserving actuary Other sources of risk related to the reserve estimation process 	Risk that historical experience is not predictive of future values

ASOP 43

3.6.1 Methods and Models – The actuary should consider methods or models for estimating unpaid claims that, in the actuary's professional judgment, are appropriate. The actuary should select specific methods or models, modify such methods or models, or develop new methods or models based on relevant factors including, but not limited to, the following:

e. the reasonableness of the assumptions underlying each method or model.

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ASOP 43 Cont.

The actuary should consider the use of multiple methods or models appropriate to the purpose, nature and scope of the assignment and the characteristics of the claims unless, in the actuary's professional judgment, reliance upon a single method or model is reasonable given the circumstances. If for any material component of the unpaid claim estimate the actuary does not use multiple methods or models, the actuary should disclose and discuss the rationale for this decision in the actuarial communication.

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Reliability of the Estimates

- Suitability of the data for bootstrapping calculations?
- Data issues that could impact bootstrapping
- Calendar Year Effects
- Trend
- Known material changes to exposure (e.g. Law change)

Others?

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Other Disclosures

- Judgmentally selected risk drivers for bootstrap?
 Coefficient of variation for Bornhuetter-Ferguson expected loss ratio
 Coefficient of variation for tail factors
- Correlation between lines of business?
- Indications from multiple models?
- Known risks not captured by statistical analysis of loss development triangles?
- Others?

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Item	Meyers & Shi	Leong, Wang & Chen	Gremillet & Miehe	Shapland
Data	50 Companies	21 (MPL) to 78 (PPAL) Companies	?	1,679 Companies
Evaluations	1	11	5	9
Models	2	2	3	8
Lines of Business	1	9	4	16
Triangle Sets	50	~4,950	296	30,707

Analysis Details

- ODP Bootstrap
- Paid Chain Ladder
- Incurred Chain Ladder
- Paid Bornhuetter-Ferguson
- Incurred Bornhuetter-Ferguson
- Paid Cape Cod
- Incurred Cape Cod
- Weighted
- Mack Bootstrap
- Paid Chain Ladder

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Analysis Details

Beginning Data

- NAIC Schedule P 4,796 Companies (& Groups)
- Remove all triangles without 10 years of data (Paid, Incurred, etc.)
- Test whether next 9 years are identical → "complete data"

Test Data

- 2,104 Companies with at least 2 Schedule P LOBs of "quality data"
- Total of 75,000+ LOBs with "quality data"
- 1,679 Companies with at least 1 Schedule P LOB of "complete data"
- Total of 30,707 LOBs with "complete data"
- Approx. 27,000 LOBs with at least 2 for same Company

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Analysis Details

- Model Output
- Accident Year Totals (by Year & All Years Combined)
- Calendar Year Totals (by Year)
- Calendar Year Runoff Totals (by Year)
- Ultimate Loss Ratios (by Year)
- Incremental Results (by Year and Development Period)
- Diagnostic Statistics

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Analysis Details

- Model Options (Tests)
- Test 1 Defaults
- No Tail factors (i.e., 1.000)
- BF a priori based on hindsight L/R, No CoV
- CC Trend = 2.5%, Decay Ratio = 90%
- Test 2 Selected Limiting of Incrementals
- Test 3 Selected Limiting & Suggested Heteroscedasticity Groups

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Model Limitations

- Model Risk
- Limited to known data
- A single model can underestimate variability
- Systemic risk
- In addition to model risk
- A shift in claims environment
- Need to Understand Assumptions

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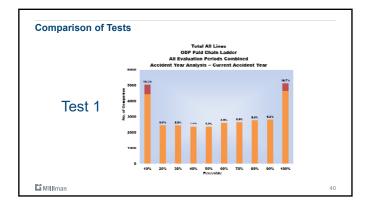
Major Assumption

Bootstrap models (ODP & Mack) assume Chain Ladder projections are unbiased

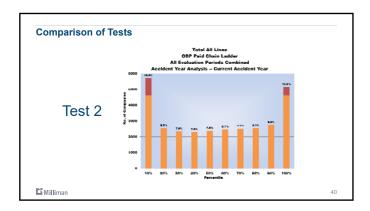
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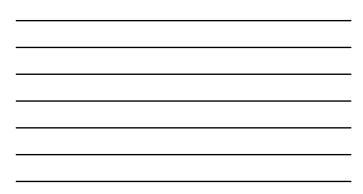
Model Projections Are they Unbiased?

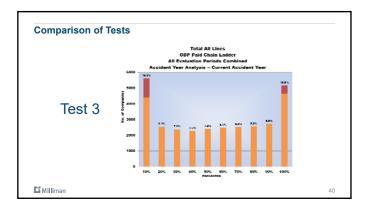
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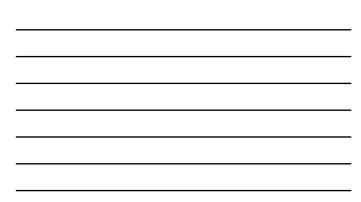


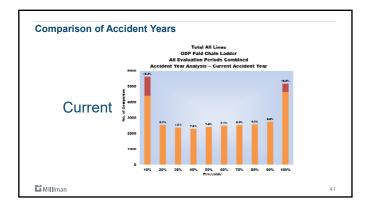


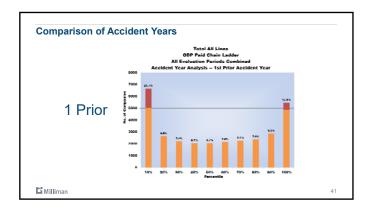


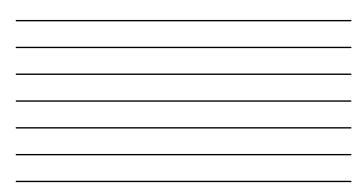


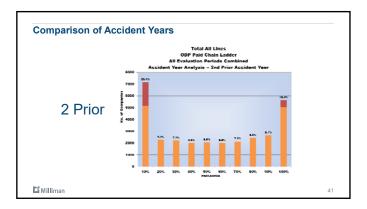


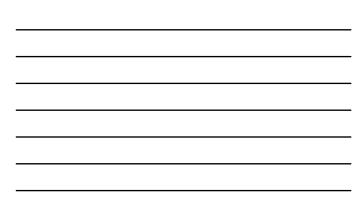


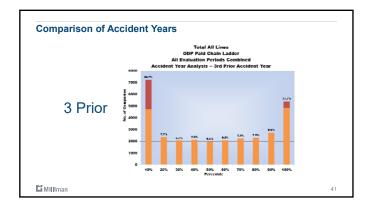




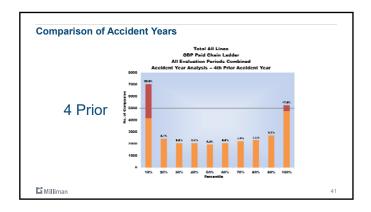




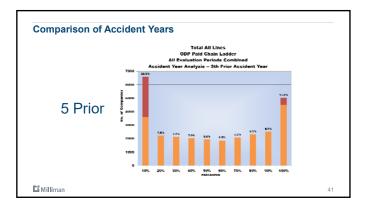


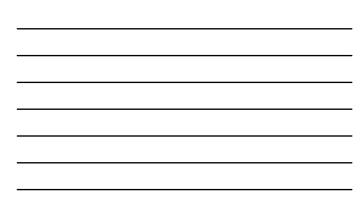


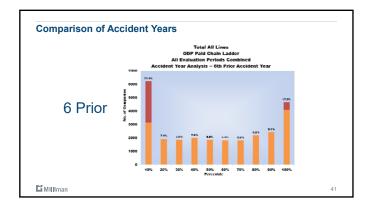


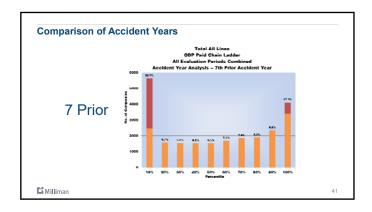




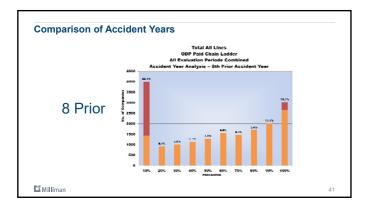


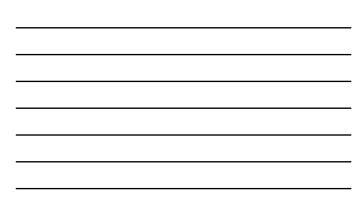


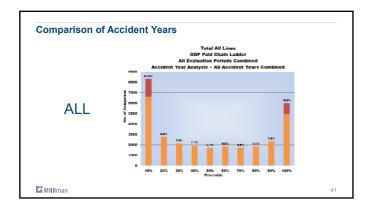


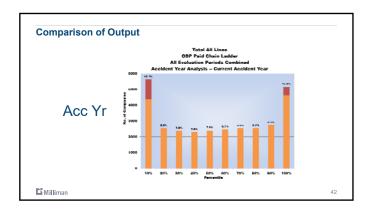




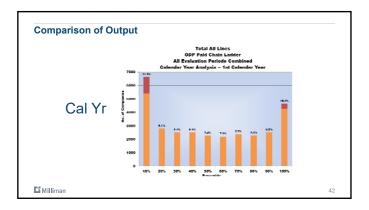


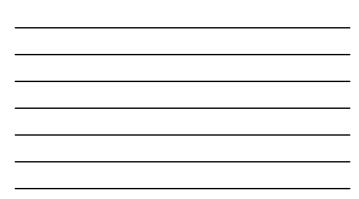


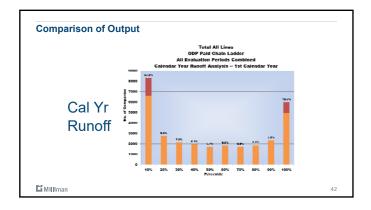


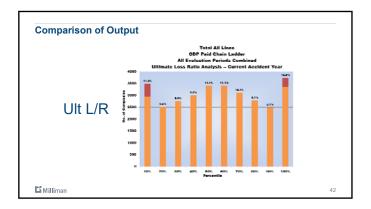




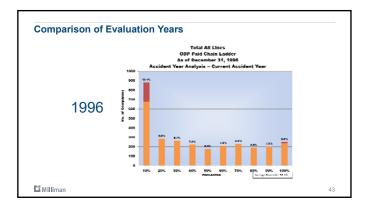


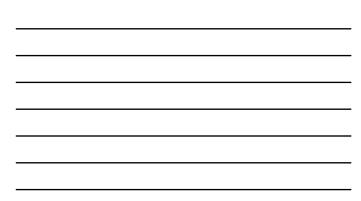


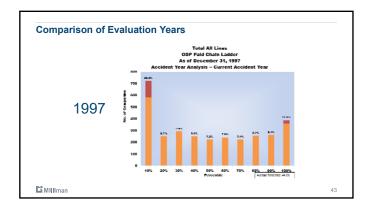




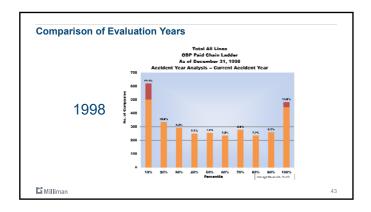




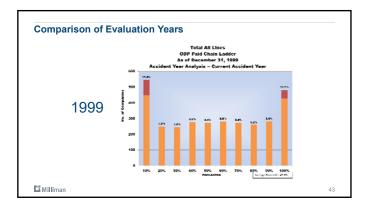


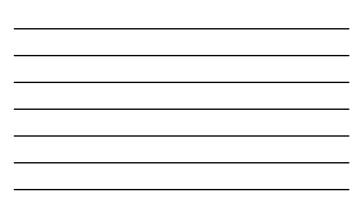


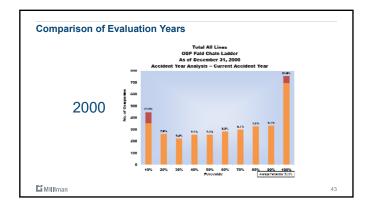


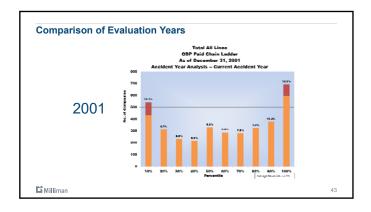




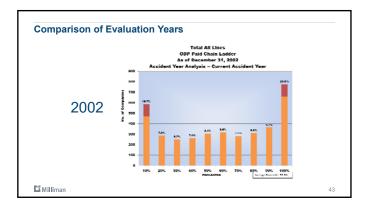


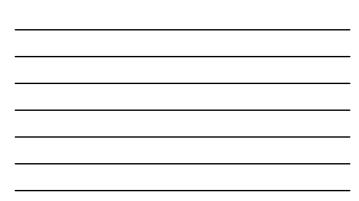


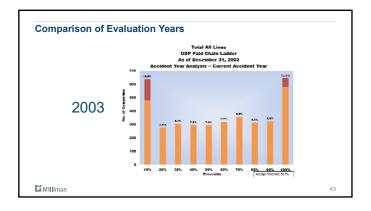


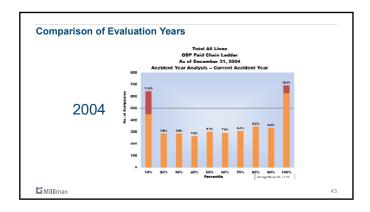


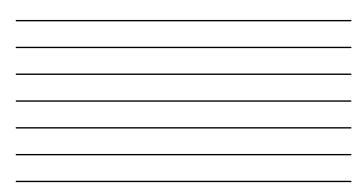


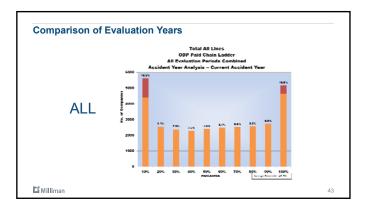


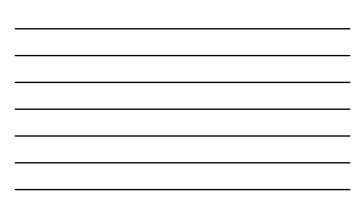






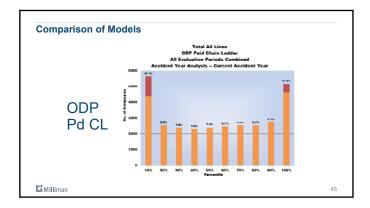




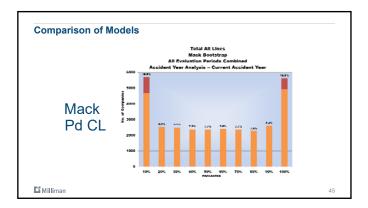


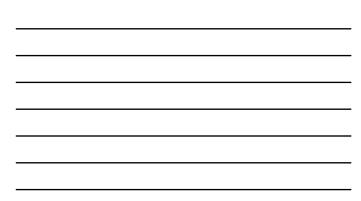
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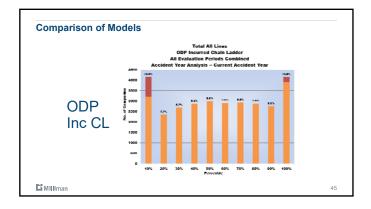




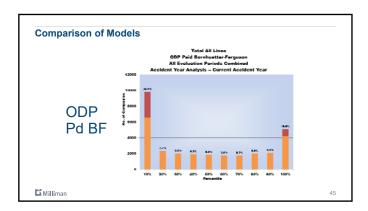




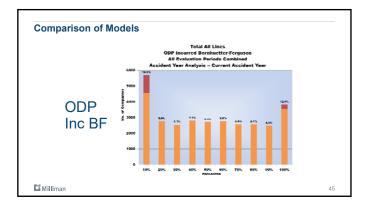


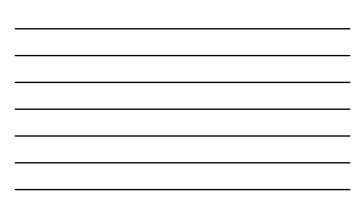


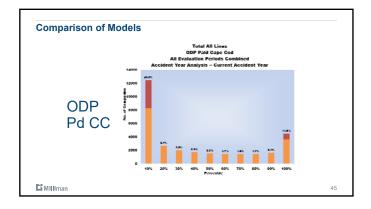


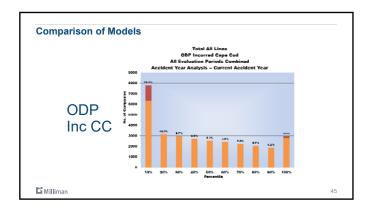




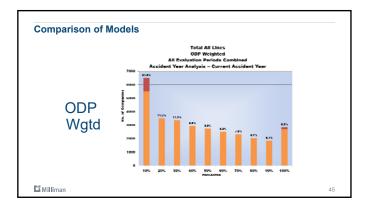


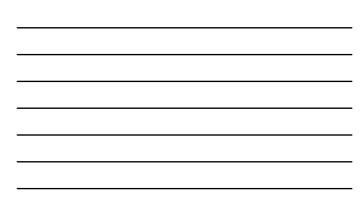














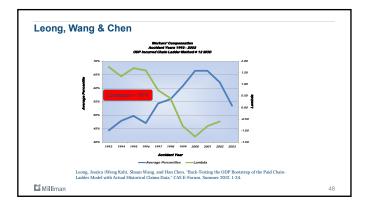
Leong, Wang & Chen

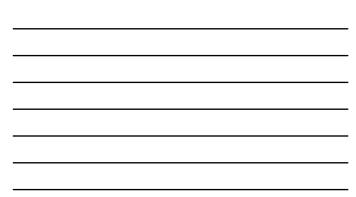
- Systemic Risk Distribution Method
 Multiply each simulated bootstrap result by a "systemic" factor
- Wang Transform Adjustment
- Increase the variability of the original unpaid loss distribution
- Shift the percentiles to account for bias in methods over time
- Relies on a parameter "Lambda" targeting an ideal histogram

Assumes Model Risk is Systemic! Based on Hindsight only!

Leong, Jessica (Weng Kah), Shaun Wang, and Han Chen, "Back-Testing the ODP Bootstrap of the Paid Chain-Ladder Model with Actual Historical Claims Data," CAS E-Forum, Summer 2012, 1-34.

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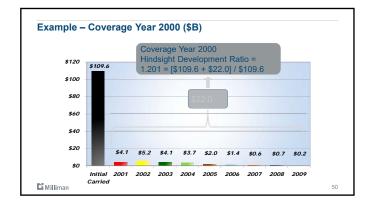
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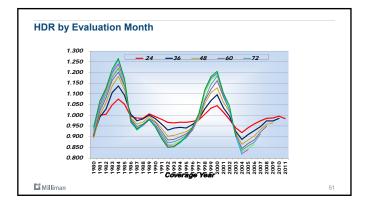
HDR Adjustment

- Shift distribution by multiplying unpaid claim estimates by the HDR
- Coefficient of variation unchanged
- Additive shift will not address variance
- Hindsight adjustment, but we are not advocating, just testing how much bias vs. not enough variance

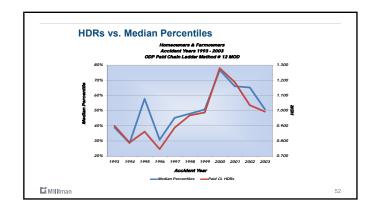
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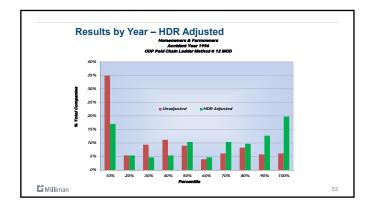




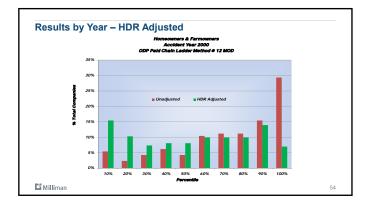


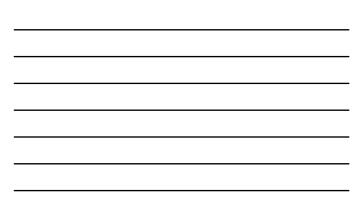




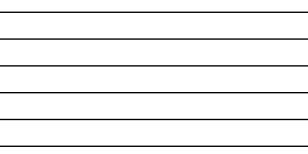












Conclusions

- Goal of Ideal Histogram Unrealized by Paid CL Bootstrap
 Both ODP Bootstrap and Mack Bootstrap
- Confirms Other Research
- Other ODP Bootstraps Much Closer to Theoretical Ideal
- Milliman Incurred models different (Shapland Monograph)
- Bornhuetter-Ferguson and Cape Cod models
- Cyclical Bias in Reserve Distributions Paid and Incurred
- Consistent with Deterministic Projections

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Conclusions

- "Corrections" to Other ODP Models may be Unnecessary
- Addressing Model Risk is very important
- Can't "blindly" accept model results
- · Use diagnostics to assess model strengths / weaknesses
- Implications for weighting
- Still need to address systemic risks
- Guidelines (i.e., benchmarks) to Assess Results
- · Based on hindsight, but forward looking
- Correlations
- Distributions by LOB and Premium

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Claim Variability Guidelines The Way Forward

1	Loss Development Patterns	
2	Unpaid Claim Distributions	
3	Correlation Between Segments	

Claim Variability Guidelines Loss Development Patterns

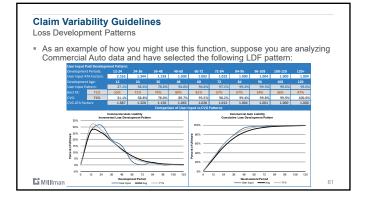
- Back-testing output includes VWA factors for all paid data triangles
- Back-testing output includes VWA factors for simulated paid data
- Actual incurred data is part of the data set, but output for incurred simulations is not readily available
- By Schedule P Line of Business, a "distribution" of the patterns were created for both actual and simulated data
- The Claim Variability GuidelinesTM product (the "Product") includes a function for calling any pattern:

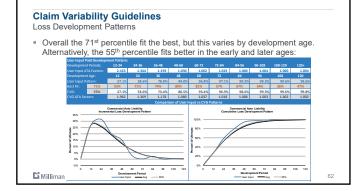
cvgLDFPattern(LOBCode, LDFPercentileCode, LDFDataCode)*

* Italicized parameters are optional

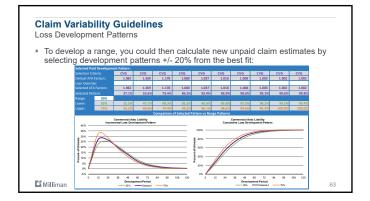
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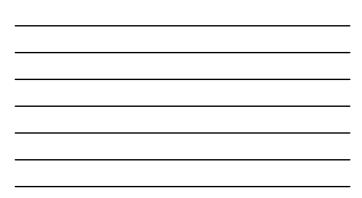
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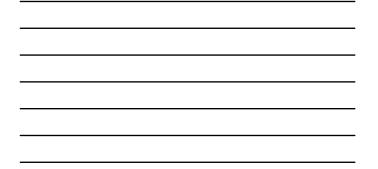








1 Loss Development Patterns	
2 Unpaid Claim Distributions	
3 Correlation Between Segments	



Claim Variability Guidelines Unpaid Claim Distributions

- For each Schedule P LOB, the back-testing results contain thousands of simulated distributions for companies of all different sizes
- Regression models were used to fit the distributions by premium volume for each of the Acc Yr, Cal Yr, Cal Yr Runoff, and Loss Ratio distributions
- Fitted results were smoothed to be consistent between distribution types and to conform with statistical properties
- This resulted in a Product function to calculate the unpaid claim benchmark: cvgUnpaid(EarnPrem, APrioriLR, LOBCode, UnpaidCode, ...)*

* Additional optional parameters not shown

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Inde Con ΔII Year Analysis – All Accident Years C 1000 600 Optional parameters allow the user to further increase or 66



- Algorithm also includes Variance Adjustment Factors to correct for
- back-testing results Separate variance adjustments factors for Loss Ratio distributions
- For example, this is the Acc Yr adjustment for Commercial Auto

decrease the variance





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Claim Variability Guidelines

Unpaid Claim Distributions

The algorithm allows for changes in assumptions to fit statistical properties.

	For example, consider smaller vs larger number of exposures:										
		Small I	nsurer				Large	Insurer			
	Commercial Auto Liability Accident Year Guidelines (US\$ 000's)							Commercial ent Year Guid			
Acc Yr	Premium	L/R	Mean	Std Dev		Acc Yr	Premium	L/R	Mean	Std Dev	CoV
2008	5,115	75.3%	17	63	369.8%	2008	40.918	75.3%	131	284	216.4%
2009	5,302	77.1%	42	112	268.7%	2009	42,415	77.1%	323	464	143.5%
2010	5,427	79.4%	95	203	213.1%	2010	43,419	79.4%	735	838	114.0%
2011	5,508	81.7%	196	308	157.3%	2011	44,064	81.7%	1,516	1,223	80.6%
2012	5,668	82.5%	404	498	123.4%	2012	45.343	82.5%	3.124	2,067	66.2%
2013	5,907	82.0%	820	737	89.9%	2013	47,256	82.0%	6,344	3,409	53.7%
2014	6,277	79.2%	1,532	1,019	66.5%	2014	50,215	79.2%	11,850	5,250	44.3%
2015	6,780	74.9%	2,719	1,640	60.3%	2015	54,236	74.9%	21,034	8,442	40.1%
2016	7,214	73.8%	4,278	2,401	56.1%	2016	57,710	73.8%	33,093	12,465	37.7%
Total	53,197	78.3%	10,102	3,654	36.2%	Total	425,576	78.3%	78,152	17,681	22.6%
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Claim Variability Guidelines Unpaid Claim Distributions

- The algorithm allows for other customizations.
- For example, consider a faster development pattern:

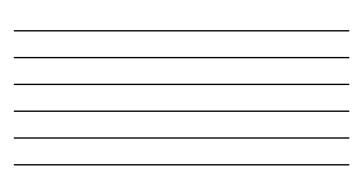
Average Development								Faster De	velopmen	t	
Commercial Auto Liability Accident Year Guidelines (US\$ 000's)								Commercial a ent Year Guid			
Acc Yr											CoV
2008	20,459	75.3%	66	157	238.9%	2008	20,459	75.3%	2	25	1506.9%
2009	21,207	77.1%	162	263	161.9%	2009	21,207	77.1%	18	79	430.9%
2010	21,709	79.4%	369	475	128.6%	2010	21,709	79.4%	69	173	249.2%
2011	22,032	81.7%	762	700	91.9%	2011	22,032	81.7%	275	360	131.0%
2012	22,671	82.5%	1,570	1,171	74.6%	2012	22,671	82.5%	794	721	90.8%
2013	23,628	82.0%	3,188	1,882	59.0%	2013	23,628	82.0%	2,029	1,320	65.0%
2014	25,108	79.2%	5,954	2,832	47.6%	2014	25,108	79.2%	4,481	2,227	49.7%
2015	27,118	74.9%	10,568	4,556	43.1%	2015	27,118	74.9%	8,926	3,945	44.2%
2016	28,855	73.8%	16,627	6,715	40.4%	2016	28,855	73.8%	15,589	6,351	40.7%
Total	212,788	78.3%	39,266	9,666	24.6%	Total	212,788	78.3%	32,182	8,202	25.5%
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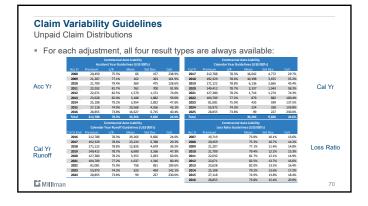
Claim Variability Guidelines

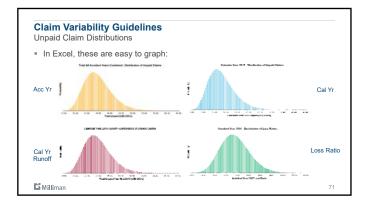
Unpaid Claim Distributions

- The algorithm allows for international use.
- For example, consider a European insurer with the same development pattern:

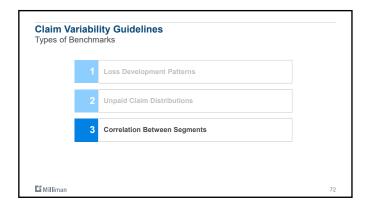
		US In	surer				Europea	n Insurer			
	Commercial Auto Liability Accident Year Guidelines (US\$ 000's)							Commercial ident Year Gu			
Acc Yr											CoV
2008	20,459	75.3%	66	157	238.9%	2008	20,459	75.3%	66	161	244.5%
2009	21,207	77.1%	162	263	161.9%	2009	21,207	77.1%	163	271	166.4%
2010	21,709	79.4%	369	475	128.6%	2010	21,709	79.4%	370	489	132.2%
2011	22,032	81.7%	762	700	91.9%	2011	22,032	81.7%	763	722	94.7%
2012	22,671	82.5%	1,570	1,171	74.6%	2012	22,671	82.5%	1,572	1,205	76.6%
2013	23,628	82.0%	3,188	1,882	59.0%	2013	23,628	82.0%	3,191	1,926	60.4%
2014	25,108	79.2%	5,954	2,832	47.6%	2014	25,108	79.2%	5,961	2,884	48.4%
2015	27,118	74.9%	10,568	4,556	43.1%	2015	27,118	74.9%	10,581	4,638	43.8%
2016	28,855	73.8%	16,627	6,715	40.4%	2016	28,855	73.8%	16,647	6,834	41.1%
Total	212,788	78.3%	39,266	9,666	24.6%	Total	212,788	78.3%	39,313	9,870	25.1%
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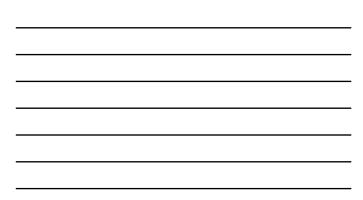












Claim Variability Guidelines

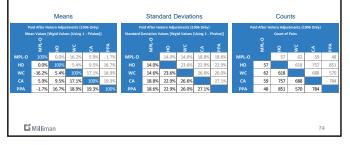
Correlation Between Segments

- Back-testing output includes correlation statistics between all pairs of LOBs within a company (i.e., if there was more than one 'complete' LOB)
- Data for all years combined or individual years is available
- Output includes both paid and incurred, before and after optimal hetero adjustments
- The mean and std dev (unweighted and weighted) for all specific pairs (i.e., between two specific LOBs) was measured
- The Product includes a function for calling any statistic: cvgCorrelation(LOBCodes, DataCode, ValueCode, TypeCode, Output)* * Italicized parameters are optional

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Claim Variability Guidelines Correlation Between Segments

For example, consider the weighted results for 5 LOBs using 1996 data:



Claim Variability Guidelines Potential Uses of Software

- Creating aggregate distributions for guidelines at the company level
- Calculating average durations for future cash flows
- Calculating reserve risk margins based on the expected unpaid claim runoff
- Assessing the variance parameter for a priori loss ratio assumptions in models
- Other uses which are only limited by your imagination

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