



#### 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."

Li Millimar

Leong, Wang & Chen

Accelerate Vision - The Character of the Pald Chain-Ladder Model with Actual Historical Claims Data," CAS E-Forum, Summer 2012, 1-34.

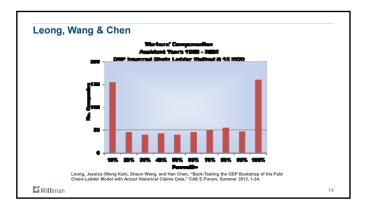
#### Leong, Wang & Chen

"...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 rast"

Leong, Jessica (Weng Kah), Shaun Wang, and Han Chen, "Back-Testing the ODP Bootstrap of the Paid

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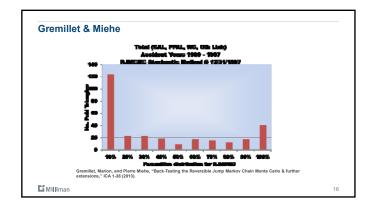
#### Leong, Wang & Chen

"...it appears that the incurred bootstrap model is also underestimating the risk of falling in these extreme percentiles."

Note: This is not the same incurred ODP bootstrap model as described in the Shapland Monograph.

Leong, Jessica (Weng Kah), Shaun Wang, and Han Chen, "Back-Testing the ODP Bootstrap of the Pai

Milliman



## Gremillet & Miehe "...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



Communication Issues		
■ Intended audience		
Intended use of the work product		
<ul> <li>Measurement objective</li> </ul>		
Reliability of the estimates		
<ul><li>Disclosures</li></ul>		-
Li Milliman	19	· ·
ASOP 43		
<ul> <li>Purpose or Use of the Unpaid Claim Estimate –</li> <li>The actuary should identify the intended purpose</li> </ul>		-
or use of the unpaid claim estimate.  Who will be using the work product?		-
<ul><li>What is their training and experience?</li></ul>		
How do they intend to use it?		
<b>Ľ</b> Milliman	20	
Examples of Intended Uses		-
Support for a Statement of Actuarial Opinion		
= M&A		
■ ERM risk assessment, capital modeling, ORSA		
<ul> <li>Internal strategic planning</li> </ul>		
SEC filings		
_		
Milliman Communication Communi	21	

ASOP 43	]
A30F 43	
3.3 Scope of the Unpaid Claim Estimate 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.	
Ľ MIJIman 22	
	_
Basis of Presentation	
<ul><li>Standard deviation</li></ul>	
Coefficient of variation	
<ul><li>Probability distribution</li><li>Probability levels / Confidence Levels / Percentiles</li></ul>	
Flobability levels / Confidence Levels / Percentiles	
Arguably satisfies the letter of the law, but the spirit of the law too?	
<b>Ľ</b> Milliman 23	
	_
Sources of Uncertainty	
ourses of officerality	
Process Risk Parameter Risk Model Risk	
• Inherent • Random noise • All models are	
randomness of in historical wrong data used to Simplifying	
payments estimate assumptions parameters are inherent to	
Risk that past is not predictive of process	
future values	
<b>Ľ</b> Milliman 24	

Sources of Uncertainty		
Independent Risk  Inherent randommess of future payments Random noise in historical data used to estimate parameters  Internal Systemic Risk  Simplifying assumptions inherent to the modeling process Unconscious biases of the reserving actuary Other sources of risk related to the reserve estimation process  Internal Systemic Risk  Risk that historical experience is not predictive of future values  The modeling process  Milliman  External Systemic Risk  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.	26	
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.		
<b>™</b> Milliman	27	

Reliability of the Estimates	
Reliability of the Estimates	
Suitability of the data for bootstrapping	
calculations?	
<ul> <li>Data issues that could impact bootstrapping</li> <li>Calendar Year Effects</li> </ul>	
Trend	
<ul> <li>Known material changes to exposure (e.g. Law change)</li> <li>Others?</li> </ul>	
- Others:	
Li Milliman 28	
Other Disclosures	
Judgmentally selected risk drivers for bootstrap?	
Coefficient of variation for Bornhuetter-Ferguson expected loss ratio     Coefficient of variation for tail factors	
Coefficient of variation for tall factors     Correlation between lines of business?	
Indications from multiple models?	
<ul> <li>Known risks not captured by statistical analysis of loss</li> </ul>	
development triangles?	
Others?	
L'Milliman 29	
Analysis Summary	

# Comparison of Analyses Item Meyers & Shi Leong, Wang & Chen Gremillet & Miehe Shapland Data 50 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 50 ~4,950 296 30,707

#### **Analysis Details**

MIIIIman

- 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**

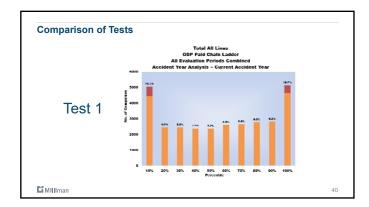
- Beginning Data
- NAIC Schedule P 4,796 Companies (& Groups)
- Remove all triangles without 10 years of data (Paid, Incurred, etc.)
- Other data quality tests → "quality data"
- Test whether next 9 years are identical → "complete data"
- Test Data
- 2,104 Companies with at least 2 Schedule P LOBs of "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

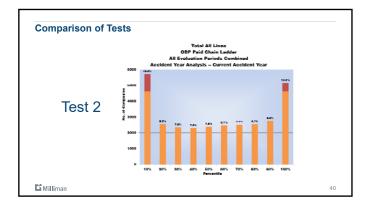
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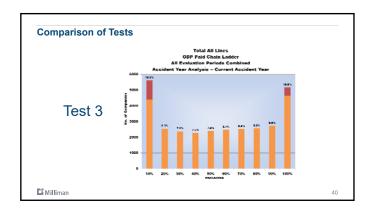
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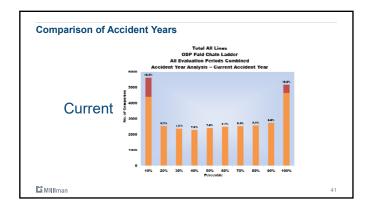
Analysis Details	
■ Model Output  ■ Accident Year Totals (by Year & All Years Combined)  ■ Calendar Year Totals (by Year)  ■ Calendar Year Runoff Totals (by Year)  ■ Ultimate Loss Ratios (by Year)  ■ Incremental Results (by Year and Development Period)  ■ Diagnostic Statistics	
Analysis Pateila	
Analysis Details	
Model Options (Tests)	
<ul> <li>Test 1 – Defaults</li> <li>No Tail factors (i.e., 1.000)</li> </ul>	
<ul><li>BF – a priori based on hindsight L/R, No CoV</li></ul>	
<ul> <li>CC – Trend = 2.5%, Decay Ratio = 90%</li> <li>Test 2 – Selected Limiting of Incrementals</li> </ul>	
<ul> <li>Test 3 – Selected Limiting &amp; Suggested</li> </ul>	
Heteroscedasticity Groups	
<b>E</b> Milliman 35	
	-
Model Limitations	
Woder Limitations	

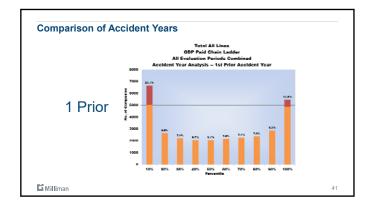
Model Limitations	
<ul> <li>Model Risk</li> <li>Limited to known data</li> <li>A single model can underestimate variability</li> <li>Systemic risk</li> <li>In addition to model risk</li> <li>A shift in claims environment</li> <li>Need to Understand Assumptions</li> </ul>	
Major Assumption	
Bootstrap models (ODP & Mack) assume Chain Ladder projections are unbiased	
Li Milliman 38	
Model Projections	
Are they Unbiased?	



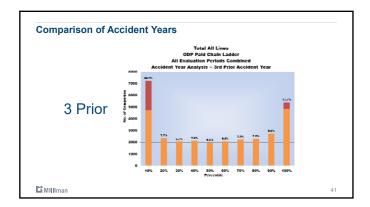


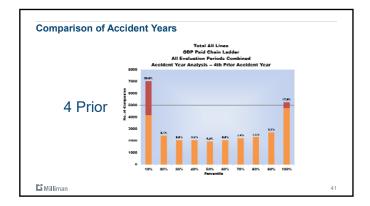


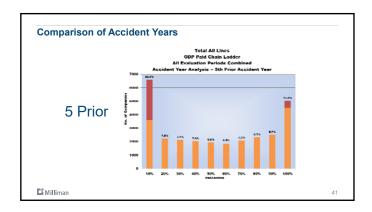


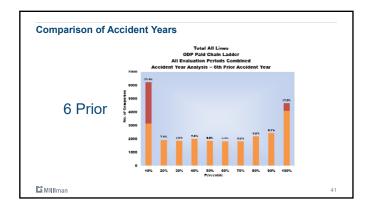


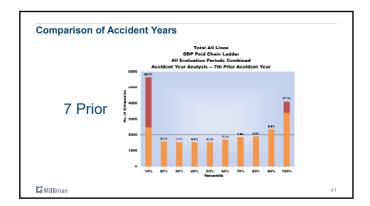


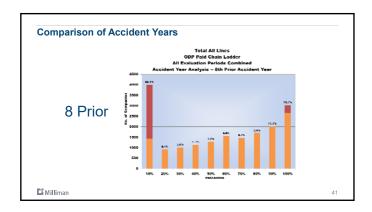


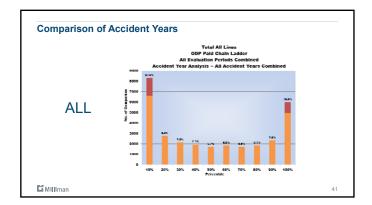


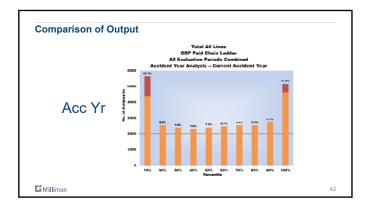


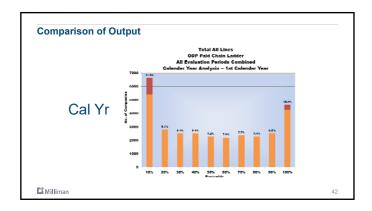


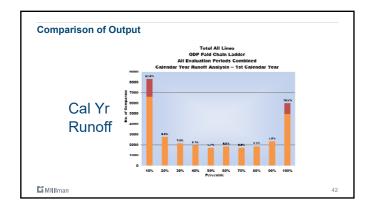


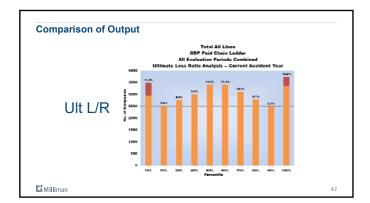


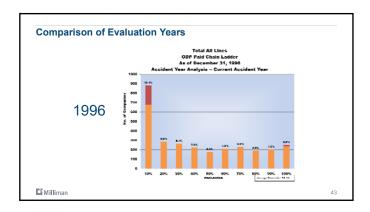


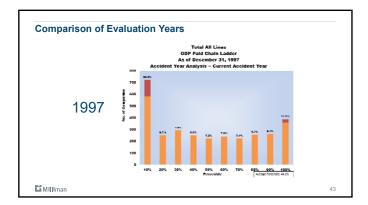


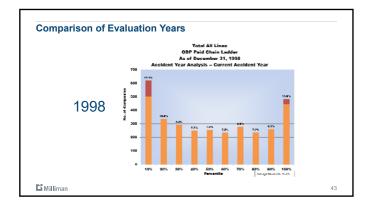


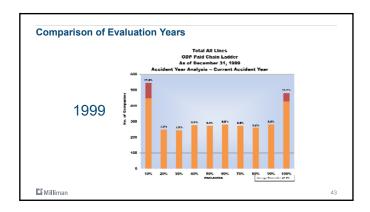


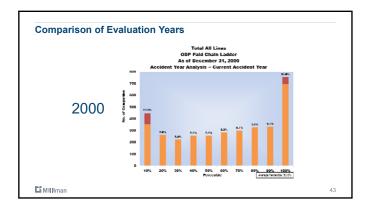


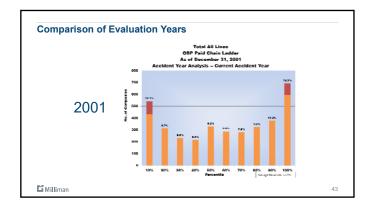


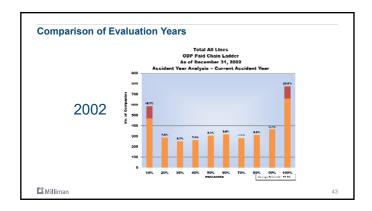


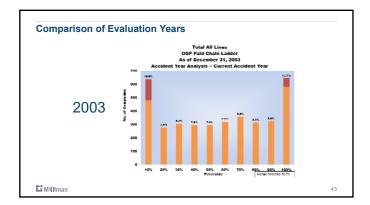


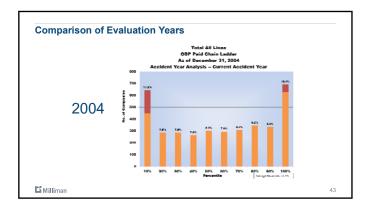


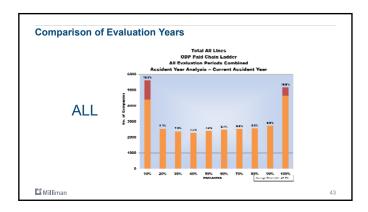


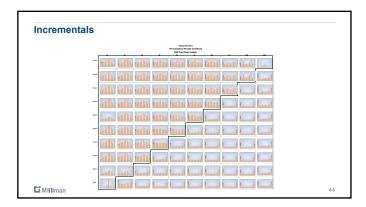


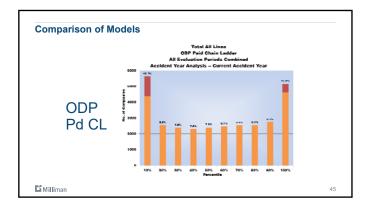


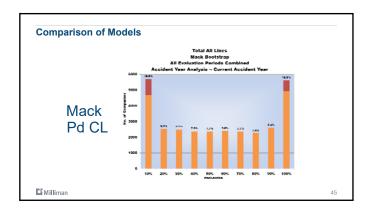


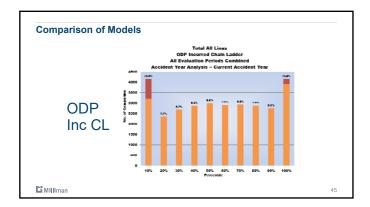


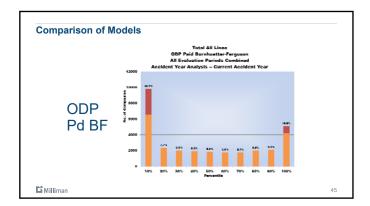


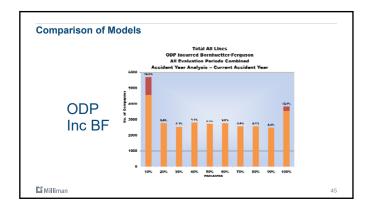


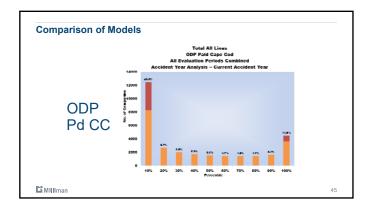


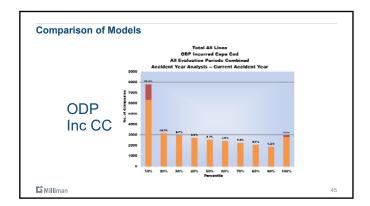


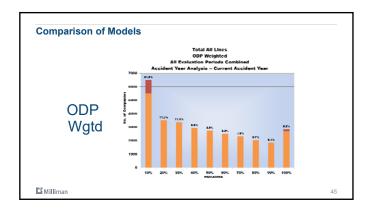














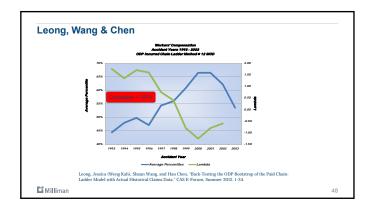
#### Leong, Wang & Chen

- 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.

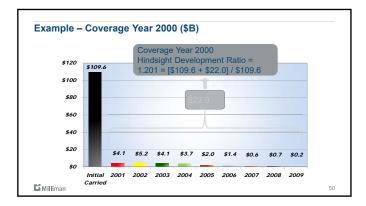
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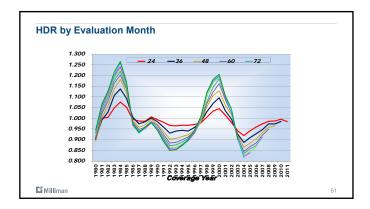


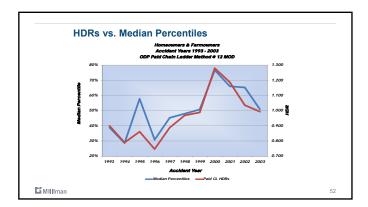
#### **HDR Adjustment**

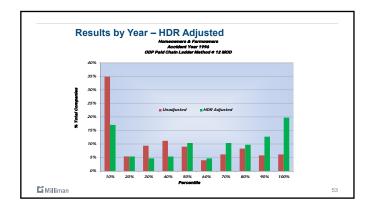
- Shift distribution by multiplying unpaid claim estimates by the HDR
- Coefficient of variation unchanged
- Additive shift will not address variance
- Hindsight adjustment, but we are not advocating, just testing how much bias vs. not enough variance

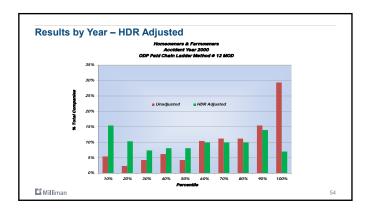
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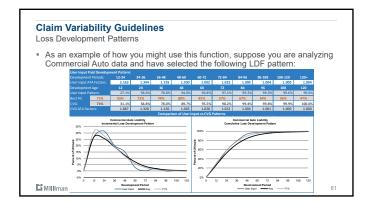


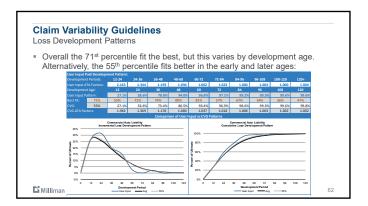


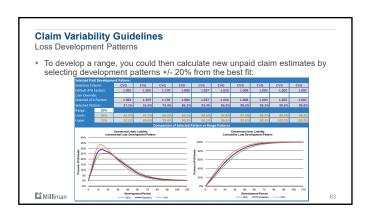


Conclusions	
Conclusions	
Conclusions	
<ul> <li>Goal of Ideal Histogram Unrealized by Paid CL Bootstrap</li> </ul>	
Both ODP Bootstrap and Mack Bootstrap	
Confirms Other Research	
Other ODP Bootstraps – Much Closer to Theoretical Ideal	
Milliman Incurred models different (Shapland Monograph)	
Bornhuetter-Ferguson and Cape Cod models	
<ul> <li>Cyclical Bias in Reserve Distributions – Paid and Incurred</li> </ul>	
Consistent with Deterministic Projections	
Consider Man Determined Frequency	
<b>™</b> Milliman	56
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	
<ul> <li>Based on hindsight, but forward looking</li> </ul>	
Correlations	
<ul> <li>Distributions by LOB and Premium</li> </ul>	
<b>□</b> Milliman	57









Claim Variability Guidelines Types of Benchmarks			
Types of Benefittation			
1 Loss Development Patterns			
2 Unpaid Claim Distributions			
3 Correlation Between Segments			
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#### **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, ...)\*

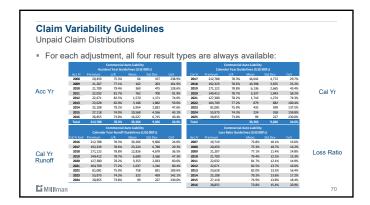
\* Additional optional parameters not shown

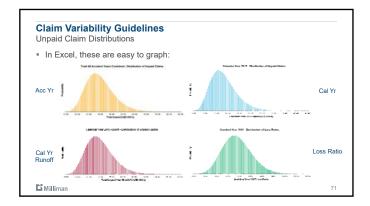
#### **Claim Variability Guidelines** Unpaid Claim Distributions Algorithm also includes Variance Adjustment Factors to correct for back-testing results Separate variance adjustments factors for Loss Ratio distributions For example, this is the Acc Yr adjustment for Commercial Auto Optional parameters allow the user to further increase or decrease the variance

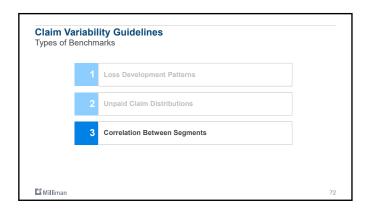
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	Claim Variability Guidelines Unpaid Claim Distributions										
The algorithm allows for other customizations.											
For example, consider a faster development pattern:											
Average Development Faster Dev							velopmen	t			
		Commercial ent Year Guid					Accid	Commercial a ent Year Guid			
					CoV						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%

Claim Variability Guidelines											
Unpaid Claim Distributions											
The algorithm allows for international use.											
For example, consider a European insurer with the same development pattern:								rn:			
		US In	surer					European	n Insurer		
		Commercial ent Year Guid				Commercial Auto Liability Accident Year Guidelines (€ 000's)					
					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	166.4% 132.2%
	21,709 22,032	79.4% 81.7%	369 762	475 700	128.6% 91.9%	2010 2011	21,709 22,032	79.4% 81.7%	370 763	489 722	132.2% 94.7%
2010 2011 2012	21,709 22,032 22,671	79.4% 81.7% 82.5%	369 762 1,570	475 700 1,171	128.6% 91.9% 74.6%	2010 2011 2012	21,709 22,032 22,671	79.4% 81.7% 82.5%	370 763 1,572	489 722 1,205	132.2% 94.7% 76.6%
2010 2011 2012 2013	21,709 22,032 22,671 23,628	79.4% 81.7% 82.5% 82.0%	369 762 1,570 3,188	475 700 1,171 1,882	128.6% 91.9% 74.6% 59.0%	2010 2011 2012 2013	21,709 22,032 22,671 23,628	79.4% 81.7% 82.5% 82.0%	370 763 1,572 3,191	489 722 1,205 1,926	132.2% 94.7% 76.6% 60.4%
2010 2011 2012 2013 2014	21,709 22,032 22,671 23,628 25,108	79.4% 81.7% 82.5% 82.0% 79.2%	369 762 1,570 3,188 5,954	475 700 1,171 1,882 2,832	128.6% 91.9% 74.6% 59.0% 47.6%	2010 2011 2012 2013 2014	21,709 22,032 22,671 23,628 25,108	79.4% 81.7% 82.5% 82.0% 79.2%	370 763 1,572 3,191 5,961	489 722 1,205 1,926 2,884	132.2% 94.7% 76.6% 60.4% 48.4%
2010 2011 2012 2013 2014 2015	21,709 22,032 22,671 23,628 25,108 27,118	79.4% 81.7% 82.5% 82.0% 79.2% 74.9%	369 762 1,570 3,188 5,954 10,568	475 700 1,171 1,882 2,832 4,556	128.6% 91.9% 74.6% 59.0% 47.6% 43.1%	2010 2011 2012 2013 2014 2015	21,709 22,032 22,671 23,628 25,108 27,118	79.4% 81.7% 82.5% 82.0% 79.2% 74.9%	370 763 1,572 3,191 5,961 10,581	489 722 1,205 1,926 2,884 4,638	132.2% 94.7% 76.6% 60.4% 48.4% 43.8%
2010 2011 2012 2013 2014 2015 2016	21,709 22,032 22,671 23,628 25,108 27,118 28,855	79.4% 81.7% 82.5% 82.0% 79.2% 74.9% 73.8%	369 762 1,570 3,188 5,954 10,568 16,627	475 700 1,171 1,882 2,832 4,556 6,715	128.6% 91.9% 74.6% 59.0% 47.6% 43.1% 40.4%	2010 2011 2012 2013 2014 2015 2016	21,709 22,032 22,671 23,628 25,108 27,118 28,855	79.4% 81.7% 82.5% 82.0% 79.2% 74.9% 73.8%	370 763 1,572 3,191 5,961 10,581 16,647	489 722 1,205 1,926 2,884 4,638 6,834	132.2% 94.7% 76.6% 60.4% 48.4% 43.8% 41.1%
2010 2011 2012 2013 2014 2015 2016 Total	21,709 22,032 22,671 23,628 25,108 27,118	79.4% 81.7% 82.5% 82.0% 79.2% 74.9%	369 762 1,570 3,188 5,954 10,568	475 700 1,171 1,882 2,832 4,556	128.6% 91.9% 74.6% 59.0% 47.6% 43.1%	2010 2011 2012 2013 2014 2015	21,709 22,032 22,671 23,628 25,108 27,118	79.4% 81.7% 82.5% 82.0% 79.2% 74.9%	370 763 1,572 3,191 5,961 10,581	489 722 1,205 1,926 2,884 4,638	132.2% 94.7% 76.6% 60.4% 48.4% 43.8%







## **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 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 MIJIman **Claim Variability Guidelines** Correlation Between Segments For example, consider the weighted results for 5 LOBs using 1996 data: Standard Deviations Milliman **Claim Variability Guidelines** Potential Uses of Software Creating aggregate distributions for guidelines at the company 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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Any Final Questions?	
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