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RESERVE AUTOMATION FRAMEWORK

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- Significant degree of uncertainty and complexity associated with Insurance and Health Liability estimates due to the complex and randomness of the claims process – incidence, reporting, settlement.
- Reserving practice is not just the selection of method, but also the management of data, the application of judgement and peer review.
- Many reserving techniques available. Important to consider the type, level and reliability of data available, the features of the portfolio and how best to identify and model the loss drivers.
- It is not enough for Boards to be presented with the actuarial "best estimate" view of reserves. To make informed decisions, actuaries and Boards need to understand the assumptions and methods relied upon and the sensitivity of – and potential volatility in – the reserves.









Three themes are influencing the way we need to work

1) The world is becoming more complex, and more complex in modelling requirements.

2) Deadlines for sharing information around a business and for reporting information to external stakeholders are shortening.

3) Technology and processing power available to improve the way we work and our expertise needs to keep up with this. Tech to augment what we do.









This session will:

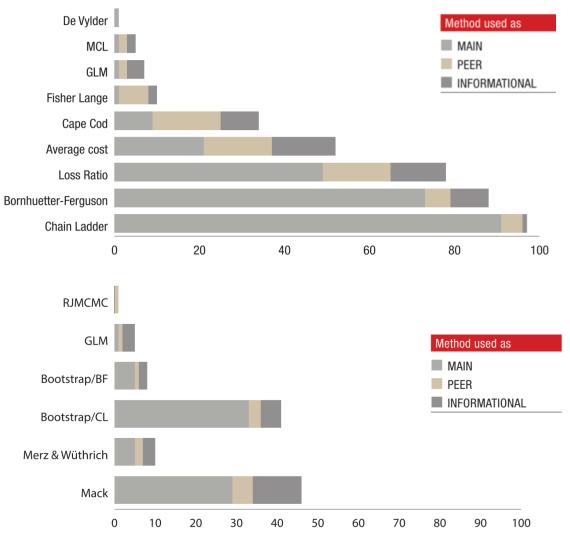
- Highlight some of the reserving methods used in the market beyond the basic Chain Ladder and Bornhuetter-Fergusson
- Demonstrate a Framework for Automation and validation
- Diagnostics how do you know when machine gets in wrong
- Show you a case study of method performance and measurement over a 20 year history of a bodily injury liability reserving class.







ASTIN 2016 SURVEY METHODS USED



Range of methods available, but few methods used. Graphs shows global responses, table shows South Africa specific:

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	Unused
Percentage	86%
Loss ratio	57%
Chain ladder	9%
Bornhuetter-Ferguson	27%
Cape Cod	86%
Average cost	52%
De Vylder	100%
Fisher-Lange	100%
GLM	100%
Munich Chain Ladder	100%
Market-based std dev	91%
Internal calibration	95%
Mack	57%
Merz & Wüthrich	90%
GLM	100%
Bootstrap / CL	39%
Bootstrap / BF	90%
RJMCMC	100%



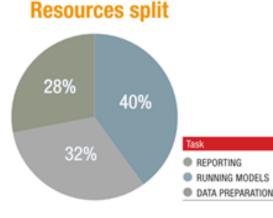






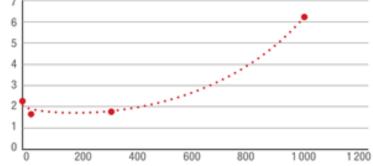
If so many methods available, why such a constrained response?

- Resource constraints, lack of appropriate software or systems which requires lots of resources on data prep.
- Tools used not conductive to Automation. Almost 40% of companies perform reserving modelling exclusively in Excel. Specialist tools focused only on Actuarial Technical method, not process and validation.
- SA 35% on data prop and 28% on reporting only about 1/3rd of time spend on Analysis and Modelling



Average number of companies vs companies size

Running model is the main task for actuaries for most insurers (40%). Then comes data preparation (32%), and reporting (28%).



The number of actuaries required to make a reserving exercise seems to be constant around 2, until the company size exceeds USD 500M.



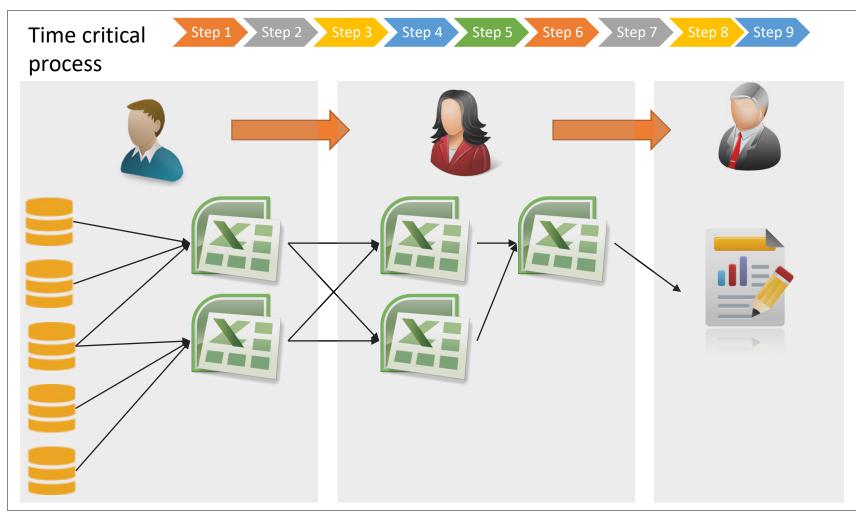




TYPICAL PROCESS



Data manipulation, fixing formulas, updating links....



...no time to meaningfully consider the appropriateness of methods or assumptions, or to conduct sensitivity testing, or extend use of other methods.

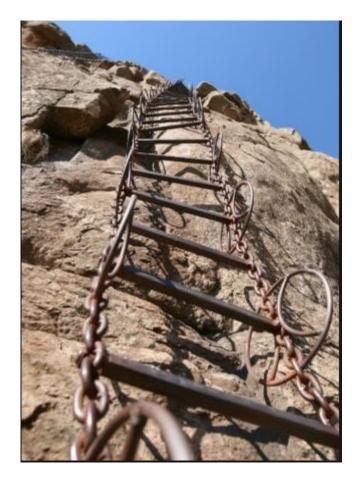
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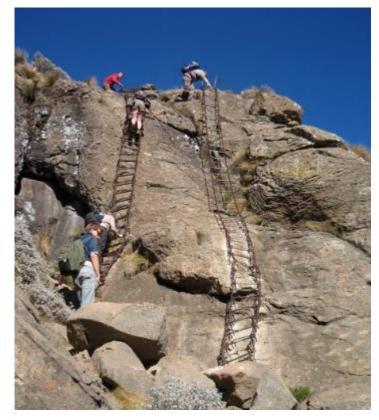




TYPICAL ANALYSIS APPROACH



















- 1. Think Rules based, not link ratio based.
- 2. Automate the data manipulation, segmentation and adjustment process
- 3. Use a platform that can allow for Script or Code based selections
- 4. Allow for wide number of methods
- 5. Segment highly volatile claim types and model stochastically, e.g. special treatment of large claims.
- 6. Know limitations of methods, and set 'selection rules' to accommodate them
- 7. Once rules are set let the process run 'hands free'
- 8. Diagnostics and Visual tools, A vs E, Back-Testing, Residuals, Graphs
- 9. Apply actuarial judgement, smoothing and prior selection refinement.

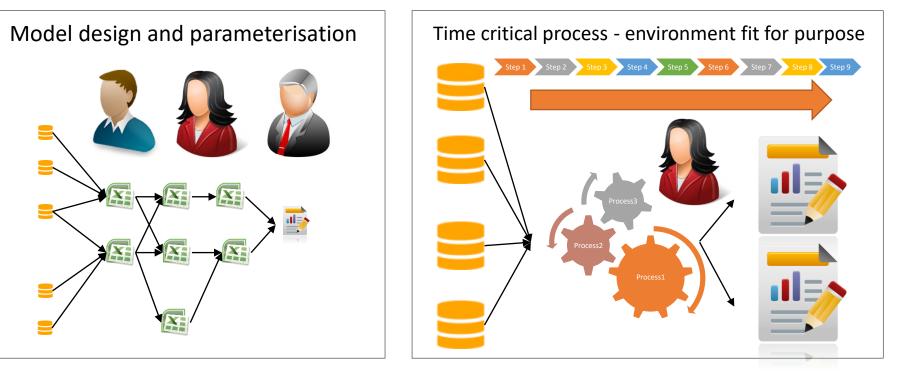








Technical Process Industrialisation



- Repeatable processes are defined in a coded environment.
- Securing data flow from process to process as well as securing calculation engines.
- An "actuary-in-a-box" approach to parameterisation of reserving data.
- Gives a first cut that can then be improved upon by experts









Chain Ladder and **Bornhuetter-Ferguson (BF)** clearly favoured across the globe, with Average cost per claim methods.

Focus on Triangle-based techniques, range of methods very useful to understand underlying trends

- Projected Payments per Claim Incurred (PPCI) paid in each development period as a % of ultimate claims reported – Ultimate numbers reported and paid per development period
- Projected Payments per Claim Finalised (PPCF) projects average cost of claims finalised in each development period – Finalised numbers projected per development period AND paid amounts per development period
- Projected Case Estimates (PCE) Project movements in case estimates, and payments as % of such case estimates per development period.
- **Benktander** Iterative BF method, where posterior from previous analysis used as prior.
- Munich Chain Ladder regression on paid to incurred relativities
- Other methods such as Stochastic Case Estimation (SCE) or Frequency / Severity good to consider as can supplement with other variables other than delays in settlement.







TECHNIQUE LIMITATIONS



Method	Limitation						
Chain Ladder Paid	Assume stability in historic settlement rates and amounts. As its proportional to development to date can be very volatile for immature years.						
Bornhuetter-Ferguson	Prior assumption may not be appropriate, but full weight given to future development. BT method would increase credibility on experience over time.						
Chain Ladder Incurred	Follows trends in data without identifying the cause and can lead to erratic results unless those trends are stable. Assume case estimates either reliable or stable in approach.						
PPCI	Can be particularly unreliable in the tail when there are only a few claims open. A PCE more appropriate in the tail. Basic model not allow for case estimation.						
PPCF	Number of claims finalised can be unstable, and sensitive to movements in number of open claims						
PCE	Not really suitable for recent cohorts of long tailed liabilities, better used in the tail.						
Munich CL	Regression on paid to incurred ratios so will always project between 2 methods. Management may feel 'exposed' to allow large savings relative to case estimates.						



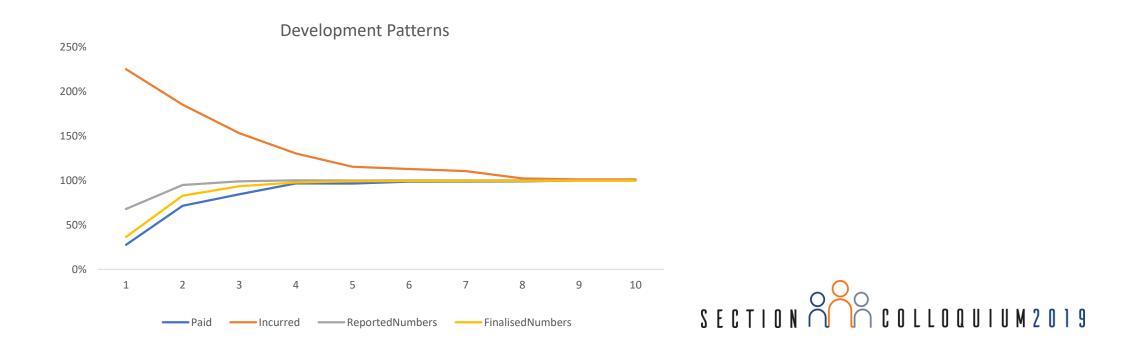






Case study: Bodily Injury Portfolio of risks. Applied 6 methods to compare.

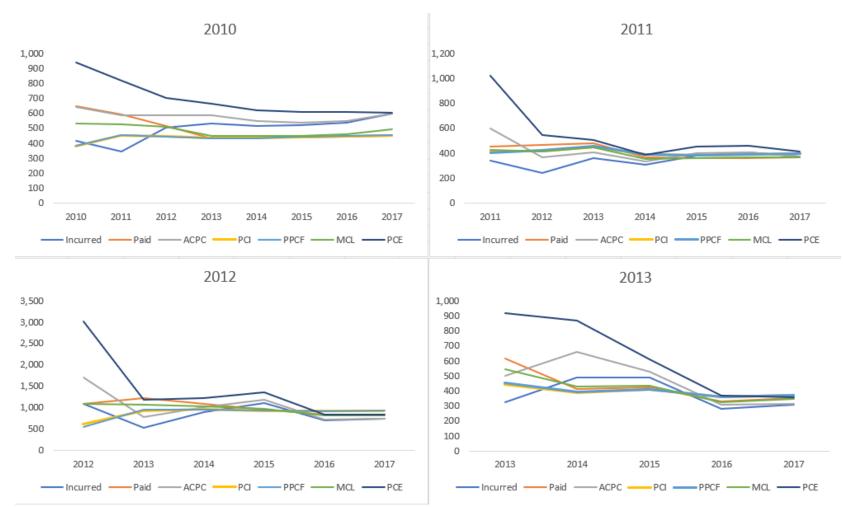
- 20 year claims history of a stable portfolio
- Reasonable consistent development process
- Following shows statistical projected ultimate for each accident year, at different points in time as the cohort matures.





ULTIMATE LOSS DEVELOPMENT PER METHOD





Choice of GO-TO method might be the most reliable, but change in relationship with other methods may tell you something important.

Paid / MCL seems to be more credible in early periods, PCE or incurred better in later periods.

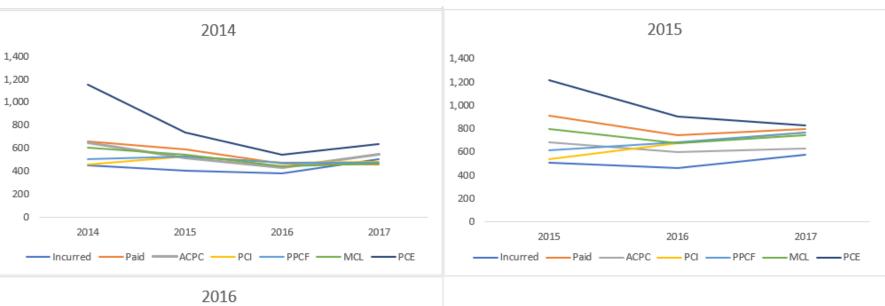


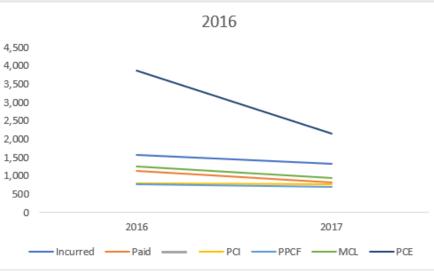




Case estimate based methods better in later periods







Incurred Ultimate projection has consistently been lower than other methods in early development. What does this mean for 2016 selection?

Once assessed your algorithm need to codify it, to make selections based on ratio / relativity of method results.

SECTION TO COLLOQUIUM 2019







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ChainLadderRegressionUltimate @= IF(IncurredChainladder/PaidChainLadder >1.1, MunichPaid, MunichIncurred) TailCaseUltimate @= IF(PaidPercDeveloped >= 1.0 AND CaseEstimateReserve > 0, PCE_PaidUltimate, Incurred_CLUItimate) APrioriThresholdUltimate @= IF(IncurredPercDeveloped <0.5, BFUItimate, Incurred CLUItimate) MinimisedErrorPPCFUltimate @= IF(PaidLambda > 0.4, PPCFOT_Ultimate, PPCF_Ultimate) Editing Full 🖬 🏷 🗮 | 🕄 Section Overview PaidCL IncurredCL FinNumbersCL RepNumbersCL MCL PCE Finalisations PPCFOT Selected Section Full × fl ⊕ ⊖ ⊒ ≡ SummaryFormulae Parameters Operators Insert Functions Options ChainLadderRegressionUltimate @= IF(IncurredChainladder/PaidChainLadder >1.1, MunichPaid, MunichIncurred) TailCaseUltimate @= IF(PaidPercDeveloped >= 1.0 AND CaseEstimateReserve > 0, PCE_PaidUltimate, Incurred_CLUltimate) Search APrioriThresholdUltimate @= IF(IncurredPercDeveloped <0.5, BFUltimate, Incurred_CLUltimate) MinimisedErrorPPCFUltimate @= IF(PaidLambda > 0.4, PPCFOT Ultimate, PPCF Ultimate) Full Ulti ତ୍ର୍୍ 🕻 😜 ChainLadderRegres... TailCaseUltimate APrioriT MinimisedErrorPPC... Over FullUltimatesChart ΘQ Initial selections can 2006.1 134 181.47 134 181.47 134 181.47 134 083.00 Selected Ultimates 2006.2 98 529.70 98 529.70 98 529.70 98 457.40 2007.1 162 850.50 162 731.00 OverrideUltimates 162 850.50 162 850.50 over time be automated MinimisedErrorPPCEUItimate 2007.2 147 146.98 147 146.98 147 146.98 147 039.00 APrioriThresholdUltimate 2008.1 144 429 99 144 429 99 143 131 44 144 324.00 600000 ✓ TailCaseUltimate 2008.2 325 351.00 325 589.93 325 589.93 322 662.59 as well. Opens up for ChainLadderRegressionUltimate 2009.1 209 054 35 209 066.98 207 128.97 208 844.00 PPCFOT_Ultimate_Paid 2009.2 251 006 55 251 057.03 248 558.87 250 585.00 PCE_Ultimate_Paid 2010.1 101 657.77 101 910.00 99 417.87 100 843.00 MCL_UltimateMunichPaid 500000 2010.2 358 489.06 346 289.68 450 290.96 342 664.00 MCL_UltimateMunichIncurred 2011.1 259 196.79 255 922.53 283 325.31 253 243.00 2011.2 109 386.60 110 058.79 104 174.22 108 721.00 400000 Paid Blended Lower Triangle Cumulative ChainLadderRegressionUltimate 2016.1 : 334862.125343148 1 1995.2 0.00 0.00 0.00 0.00 1996.1 0.00 0.00 0.00 110.00 300000 1996.2 0.00 0.00 0.00 0.00 1997.1 0.00 977.60 1 804.20 2 516.00 1997.2 0.00 0.00 30.00 3 201.00 200000 1998.1 0.00 0.00 5 739.40 5 739.40 1998.2 809.85 5 195.41 14 124.30 20 217.40 1999.1 16.00 953.00 2 725.55 7 438.85 1999.2 2 305.00 7 113.98 95.00 25 799.30 100000 2000.1 0.00 1 969.55 16 971.80 19 709.00 2000.2 518.00 1 840.71 10 930.00 14 108.40 2001.1 4 956.05 22 810.90 34 521.30 35 878.70 Incurred Blended Lower Triangle Cumulative Œ 2 14 1995.1 0.00 0.00 0.00 0.00 1995.2 0.00 10 000.00 13 000.00 13 000.00 2010.2 2011.2 2012.2 2013.2 2014.2 2015.2 2016.2 1996.1 Hosted by 0.00 0.00 500.00 110.00 SECTION 🔏

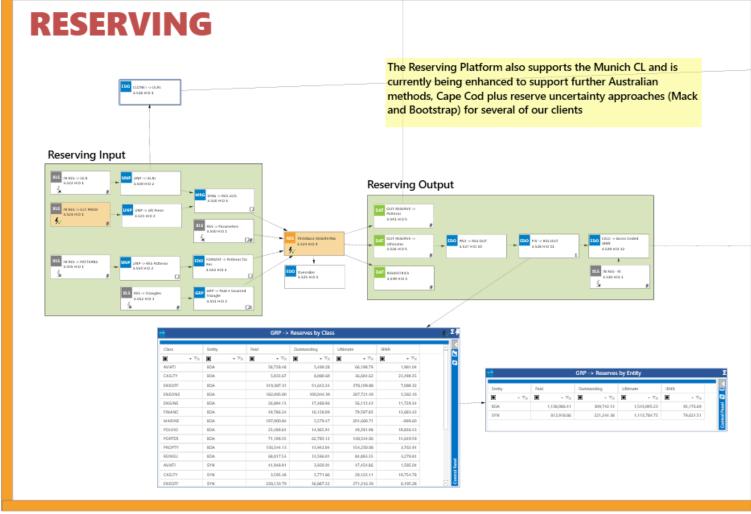


AI.



NOW PRESS PLAY







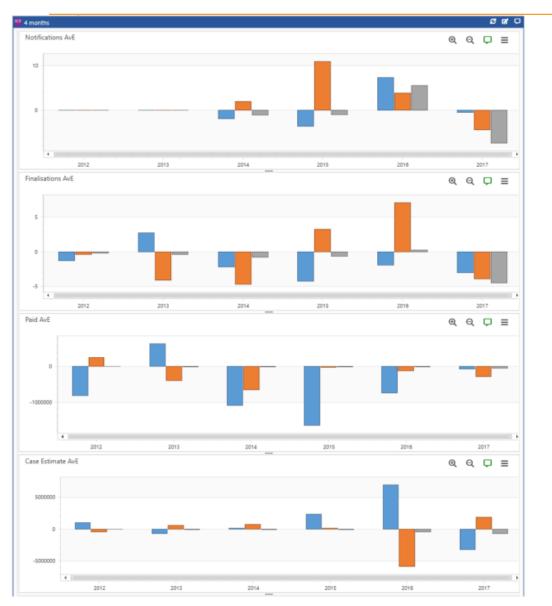






DIAGNOSTICS REVIEW OF RESULTS





Actual vs Expected and Back Testing:

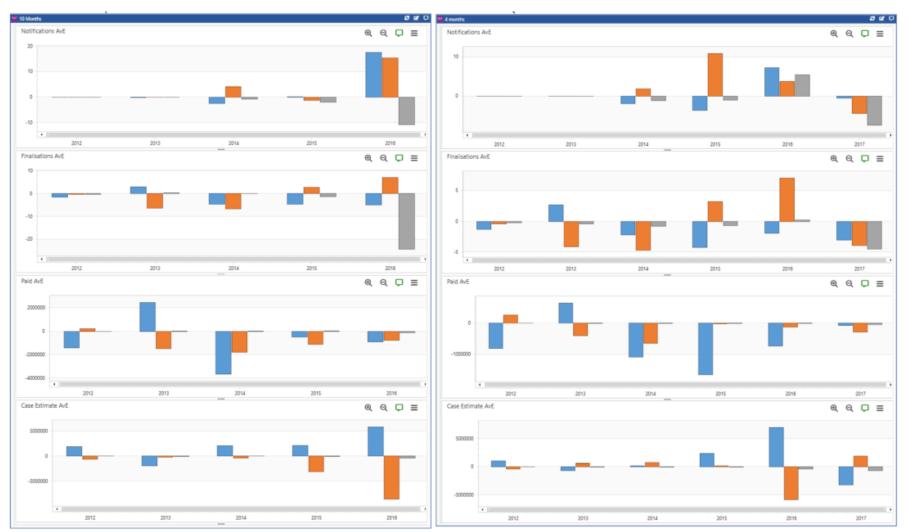
- Dashboard of A vs E by method gives insight into over / under projections
- Based on previous selected pattern

 assess how previous assumptions
 perform against experience.
- Based on newly selected pattern back-test new assumptions as driver of change in ultimate claims projection.
- Case Position lower than expected relating to number of finalisations greater than expected



DIAGNOSTICS A vs. E Time Periods Considered









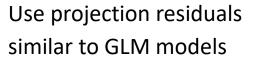


DIAGNOSTICS RESIDUAL GRAPHS



Res	idual Triangle Valu	ies			Ð	Q	. . .	=	Residu	ualTriangleValue	s								Ð	Q	Q	≡
	1	2	3	4	5	6																
1997	-0.32	0.15	-3.76	14.05	8.78			-5.76														
1998	-0.30	0.31	-3.15	10.14	-3.48			3.10	20		•											
1999	-0.19	0.10	-2.83	11.06	-2.93			-1.83														
2000	-0.21	0.23	-2.50	5.51	-0.70			-0.23							•							
2001	-0.03	0.03	-2.87	0.06	-0.86			-3.41		•							•					
2002	0.11	-0.26	-2.73	-6.26	-3.91		-	11.12	10					•								
2003	0.06	0.21	-1.78	-2.88	-21.67			-8.55		•		•										
2004	0.16	0.18	-2.10	-21.24	-15.94			-1.17			1				•		۰.					
2005	-0.03							-2.55	0			. :	:	: 1	:	1	. :					
2006	0.00			-8.85	-5.84			-0.25	. °.			11	1	1			•		•		0	
2007	-0.14	-0.17	-1.20	15.58	-5.15			2.16			. :	•		•			•			•		
2008	-0.25				5.96			5.55		· · · ·	:	• .		• 1				•				
2009	-0.15				13.08			5.33	-10					•								
2010	-0.03	-0.30	-1.99	2.02	2.83			-0.89			•••	-										
2011	-0.05	-0.04	-2.22	-6.71	-0.61			0.93														
2012	-0.04	0.16	0.70	3.17	-1.50																	
2013	0.19	-0.34		-3.83					-20				•									
2014	-0.02																					
2015	0.03	-0.09)																			
2016	0.00										2000		2	005			201	10			201	5

PaidBootstrap Origin Year Metrics **PaidBootstrapOriginYearMetrics** 99.5th ▼ ## € ▼ 🖬 ▼ 🗁 Default ▼ 99th 2014 2015 2016 95th 68.208.558 142,295,792 460,606,340 685.683.87 90th 1000000000 StandardDeviation 21.362.841 44,169,260 75.235.146 138.306.386 220,399,609 75th 25th 52,691,469 110,261,032 210,648,934 359,123,812 522,607,403 SOth 683.929.129 50th 65.816.129 138.479.655 258.034.860 454.001.147 25th 75th 81.670,698 170.341.394 312.008.960 552,833,475 829,526,722 500000000 StandardDeviation 97,039,195 201.232.830 363.695.127 642,876,698 90th 968,462,350 Mean 95th 106.278,224 220,203,919 397,690,700 696.338.567 1.054,386,185 99th 127,103,717 259,249,720 463,658,921 808.092,372 1,247,696,810 99.5th 135,837,734 276.881.324 491.007.769 854.342.843 1.332.413.567 2000 2005 2010 2015



- Have a set of diagnostics • and residual plots.
- **Residuals of chose** method but also of supporting methods.

> Use stochastic ranges to determine if movement from all methods are within reasonable variance, e.g. within 75th or single Standard Deviation.









Should make use of Automation, Process industrialisation and Visual diagnostics to significantly improve actuarial practices and the value of actuaries in business decision making.

Let software do the heavy lifting so we can run the process faster, perform more methodologies, generate better metrics...

Automated reserving processes and automated parameter selection, including identification of model failure or adjustments to be made...

and consequently help businesses better **understand** their reserves and the movements in their reserves.









QUESTIONS

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