

Big Data in Practice, and the challenges of Big MI

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Common experiences





What is big data?

- Definition
- A data set whose size is beyond the ability of databases on data management tools to
 - Capture
 - Store
 - Manage
 - Analyse

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Three Categories of big data problems

- Volume
- Velocity
- Variety

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Why look at big data

- · Quote data explosion
 - Phone
 - Internet

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- Aggregation
- Enrichment

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Why look at big data



So is our data big data?

	no not really yet!		
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What did we do?

- · Lots of reading and research
- · Lots of conferences
 - The first things the conferences tell you is this is really hard

.....its really not if you work with the right people

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Sandbox exercise

- · What am I talking about
- · What did we do
- What did we learn





Problem Articulation

- · What did we need to achieve?
- · Six key data requirements
 - We were not looking to mine social media data
 - Nothing here was difficult about designing the reports
 - Nothing earth shattering in terms of data required
 - $\ldots\,$ but was by far the most important phase of the project

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Impact on Impact Analyses

 Hopefully we've all progressed through a programme of increasingly predictive modelling and built confidence in ability to predict impacts of pricing actions



 New faster introduction means one cannot create fully-spoulated quick batches or future renewal batches on which to model. Usually we only have a subset of populated historic data on which to rely.
 Moreover, behavioural/competitiveness models will not contain the new factors
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We must remain aware that our models have high statistical variance, yet we wish to rely on their central estimates more than "gut feel" of stakeholders, whilst not over-engineering analyses relative to these inherent uncertainties

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The timing factor

- We all like to use our models to generate more insightful MI than traditional approaches permit (notwithstanding some self-fulfilling risks... another presentation entriely) For example, we like to predict the YOY changes of upcoming renewals... but by transitioning into bigger data, distortions are introduced
- imple, we like to predict the YoY changes of upcoming renewals ... but by transitioning into bigger data, distortions are introduced



the risk (tighter ot for the benefits that the new data bring Even after a year, we still find ourselves in a less comfortable position: There can often be barriers that don't permit bat enriched with the latest external data, and this will only be re-looked-up as part of the actual issuance of renewal Institute and Faculty of Actuaries

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The timing factor

We also like to put sold data back into our risk m "big data" for past periods. Even with back-fill it's Is to give an indication of net effects of rating actions on expected profitability. We are missing likely to be the same data we would have attached at the time (it possibly didn't exist!)



 Some data is (almost) continually refreshed, but much is only periodically updated – this cau risks creating a cottage industry to explain away es step-changes that id to live with but hstitute and Faculty of Actuaries

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Issues we overcame

- Communication
- Supplier selection and contracts - Final solution required five suppliers, and eight contracts
- Internal challenges
 - IT skill sets
 - Internal IT
 - Executive and Board

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What did we achieve?

- Speed
- Curiosity
- Innovation
- · Data sources

Data accessibility	Actuaries
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What did we achieve that we weren't expecting

- Control of data throughout organisation
 - Standardisation of definitions
 - Reconciled data
 - Standardisation of reports
- Much more data was available for analysis
 - Portal data
 - Sub peril data

	 fraud 			
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What does it allow

Two routes in:

- traditional SAS coding access
- Visual Analytics tool
- · In memory analytics on
 - 25m quotes
 - 3m policies
 - >400k claims



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Living with the consequences

- There will be no "prevailing" mix of new factor conversion/retention rates from which to anchor views of appropriate trading KPIs. Any mix-stability MI is rendered immediately pretty useless. It takes time for conversion/retention and especially LR to evolve
- The usual trap of wanting all the benefits of the new models but the mix from old (underperforming?) rates is heightened when deploying multiple new factors
 Therefore: We need tools to communicate unusual performance versus inherent noise to stakeholders



Living with the consequences

Factor level 3 12.1%

 We also face a growing problem of sheer volume of factors for which KPIs need to be inspected, so desire some mechanical approach One solution may be some kind of automated conversion/retention/contribution "exception report"

Cohort	Quote exposure	Conversion	Exception level	Exception breach
Factor level 1	3.7%	3.23%	2.50%	29%
Easter level 2	4 407	4.259/	2 400/	250/

· Since we've been fighting against the temptation to unnecessarily fiddle with rates, we should already have this Finally, if currently adjusting conversion levels based on market info about general quote to sale ratios to get a more insightful MI pack; the difficulties of enriching external quote data with our new factors means we find curselves back in a distorted position Ind Oursesse And Faculty of Actuaries

1.34% 1.11%

21%

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21

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