

Practical experiences of using CMI models for projecting annuitant mortality

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Workshop A13

Agenda

- Brief history of mortality projections
- Aims when setting improvement basis
- Choice of model
- Use of the model - hints & tips
- Choice of datasets / parameters
- Validating output
- Adjustments to output
- Derivation of capital

Warning !

- This presentation will not contain
 - Formulae
 - In-depth discussions of statistical models

Brief history of mortality projections

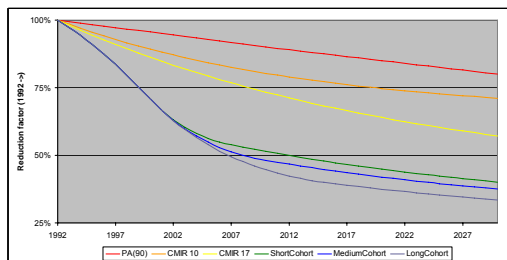
- a(55) – the first formal use of improvements
- PA(90) – projection of 1967-70 mortality to 1990
 - Uniform reduction of 1 year of age per 20 calendar years
- CMIR10 - to run alongside 80 tables
 - First use of geometric method

$$RF(x,t) = \alpha(x) + [1 - \alpha(x)] (0.4)^{t/20}$$

Brief history of mortality projections

- CMIR17 - to run alongside 92 tables
 - Revision to geometric method
- CMI wp1 - 'interim' cohort projections
 - Overlay over CMIR17 improvements
- 2005 – Spline / Lee-Carter methodologies

Brief history of mortality projections (age 65, male)



Aims when setting improvement bases

- Use of output (best estimate / reserves / ICAS)
- Affordability (on your in-force)
- Competitiveness (of new business terms)
- Sensible shape
- Consistency with base table
- Objectivity (not too much of a fiddle)
- Ease of understanding / communication

Choice of model - Lee Carter v P-Spline

- The CMI's initial software allowed both
- But the Lee Carter functionality was flawed
- So it was deactivated for version 2
- CMI paper covering Lee Carter in "late Q3 2006"
- Other methods (eg Logistic/Weibull) are valid too
- Although software does not readily support them
- Rest of presentation will hence relate to P-Spline

Use of the model - hints & tips

- What the model does & does not do
- Practical issues in using model
- Choice of dataset
- Choice of parameters
- Which ages / years to use
- Problems in the data

What the model does (& does not) do

- The model
 - Fits a curve to past data, projecting it into the future
 - Blends goodness of fit with smoothing
 - Projects future as a continuation of the past
- The model does NOT
 - Opine to quality / relevance of data
 - Adjust for future items not in past
 - Adjust for past items not expected to be in the future
 - Adjust for differences in use

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Practical issues in using model

- Use the 2nd version of the software
 - More flexible (allows choice of smoothing params)
 - More informative output
 - Read the user guide !
- Automate back end
 - Log (μ) \rightarrow μ
 - μ \rightarrow improvements
 - Improvements \rightarrow graphical output
 - Improvements \rightarrow annuity factors

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Choice of dataset

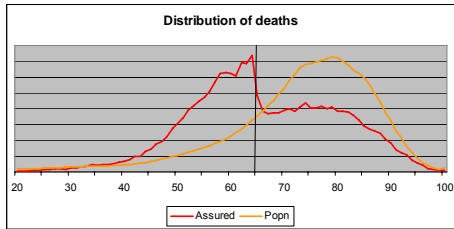
- Several choices of dataset

- CMI assured (underwritten)
- CMI annuity
- Population
- Own data
- SAPP

Volume	Trends	Relevance
√	√	√
√	x	√
√	√	√
?	?	√
√	x	√

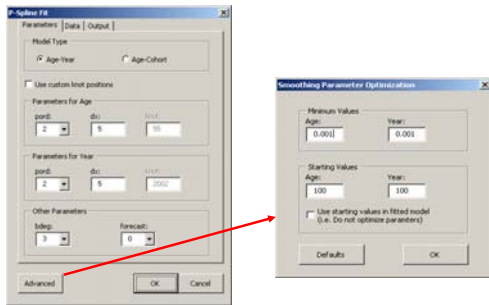
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Choice of dataset



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Choice of parameters



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Choice of parameters

- Model type Age-Year v Age-Cohort
- pord Degree of penalty to be applied
Determines the (approximate) shape of the projection
2nd order penalty gives approximately linear projection
- dx distance between spline peaks
use 5 or less – larger makes good fit impossible
- bdeg Degree of splines (3 = quadratic)
- forecast Longer = slower

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Choice of parameters

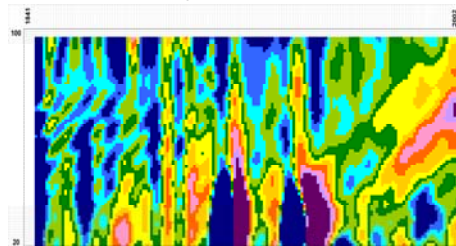
- Knot positions
 - Allows you to specify where the knots (peaks of splines) are
 - You specify one in both the age & year dimensions
 - The rest are determined from these & dx
 - Model behaves best with knots at the last calendar year of the data

Choice of parameters

- Do not optimise smoothing parameters ?
 - If you leave this box unchecked, the model will derive the blend between smoothness v goodness of fit
 - If you check the box, you get control over this.
- Smoothing parameters
 - Allows you to control how closely the model fits to the data (vs how much it smooths)

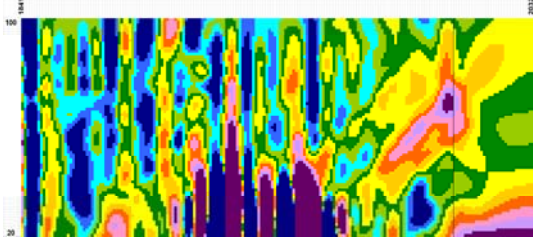
Choice of parameters (Year v Cohort)

- Consider first the past data, smoothed



Choice of parameters (Year v Cohort)

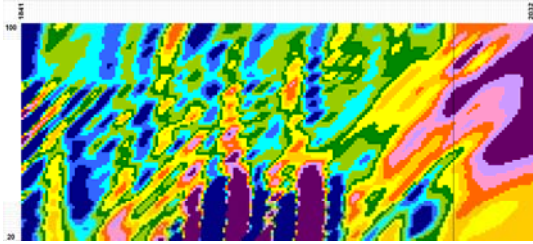
- Age-Year seems to give a sensible fit ...



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Choice of parameters (Year v Cohort)

- ... Age-Cohort does not



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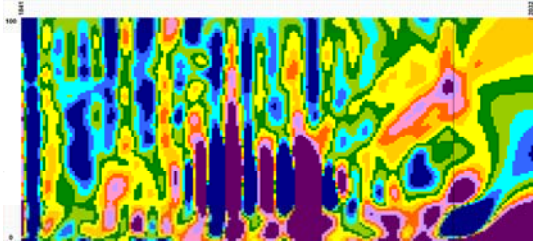
Which ages to use

- Clearly mortality improvements have varied hugely by age
- So it is sensible that the ages selected for our projection will impact on the answer

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Which ages to use

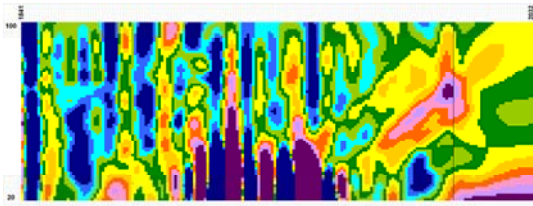
Population data, ages 0 – 100 (all data)



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Which ages to use

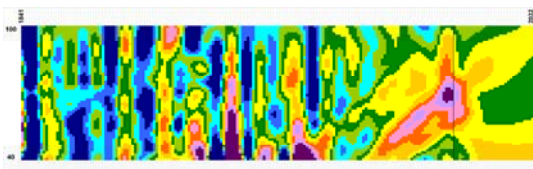
Population data, ages 20 – 100



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Which ages to use

Population data, ages 40 – 100



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Which ages to use

Population data, effect on annuity factors

- Male annuity factors
- at end 2005
- 100% PMA00 base table
- at 4.5%

	Popn. age range	
	age 65	age 75
0-100	13.32	9.11
20-100	13.21	9.04
40-100	13.13	8.99
Range	101.5%	101.3%

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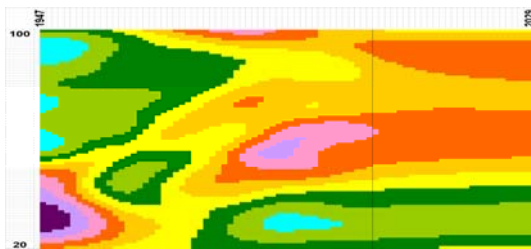
Which calendar years to use

- Assured dataset
- Investigate choice of data
- Default parameters

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Which calendar years to use

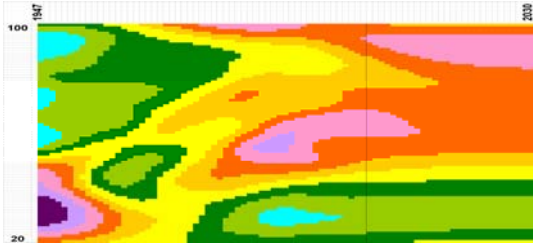
Assured data, to 1999



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Which calendar years to use

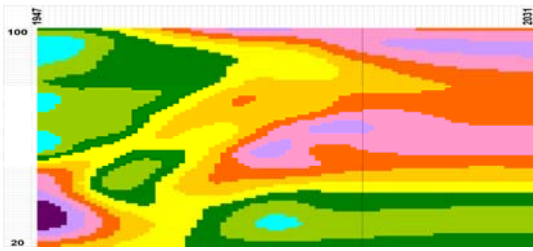
Assured data, to 2000



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Which calendar years to use

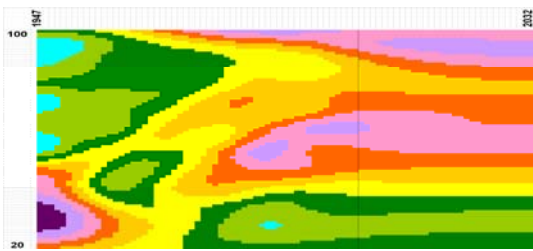
Assured data, to 2001



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Which calendar years to use

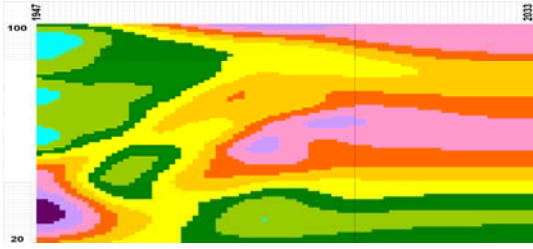
Assured data, to 2002



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Which calendar years to use

Assured data, to 2003



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Which calendar years to use

Assured data, effect on annuity factors

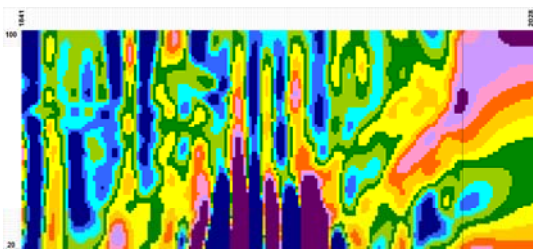
- Male annuity factors
- at end 2005
- 100% PMA00 base table
- at 4.5%

	age 65	age 75
-> 1999	13.44	9.33
-> 2000	13.72	9.54
-> 2001	13.86	9.63
-> 2002	13.83	9.61
-> 2003	13.58	9.38
Range	103.0%	103.2%

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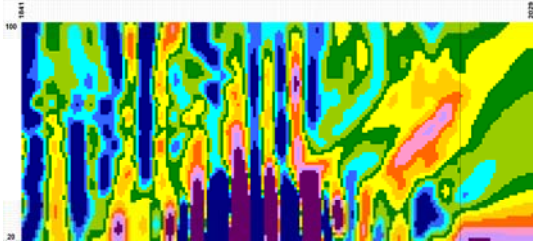
Which calendar years to use

Population data, to 1998



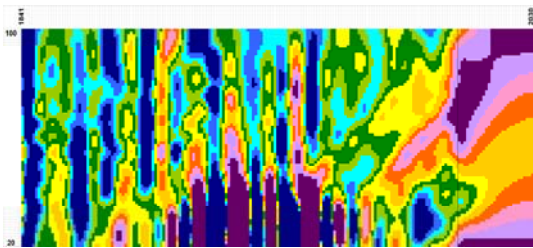
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Which calendar years to use
Population data, to 1999



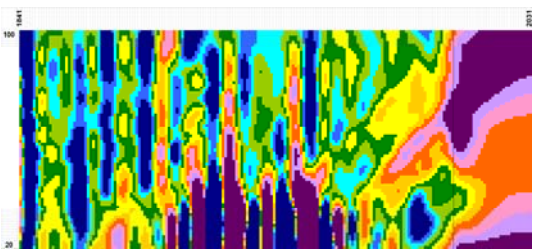
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Which calendar years to use
Population data, to 2000



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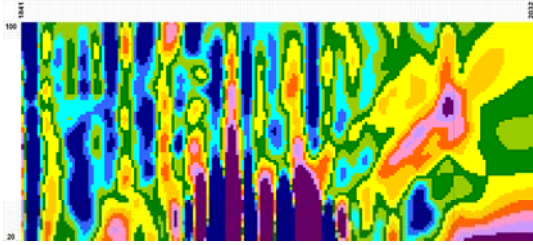
Which calendar years to use
Population data, to 2001



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Which calendar years to use

Population data, to 2002



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Which calendar years to use

Population data, effect on annuity factors

- Male annuity factors
- at end 2005
- 100% PMA00 base table
- at 4.5%

	age 65	age 75
-> 1998	14.60	10.42
-> 1999	12.83	8.66
-> 2000	14.96	10.60
-> 2001	15.45	10.91
-> 2002	13.21	9.04
Range	120.4%	123.1%

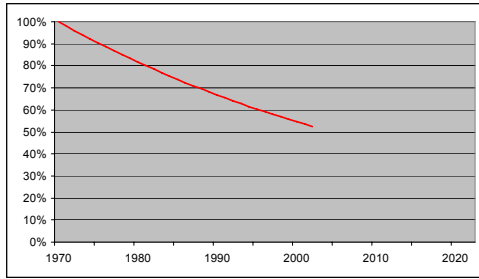
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Problems - data distortions

- Controlled data set
- Initially with smooth qx development
- Looking at age 65 only
- Using population data (1841 – 2002)
- Graphs showing 1970+

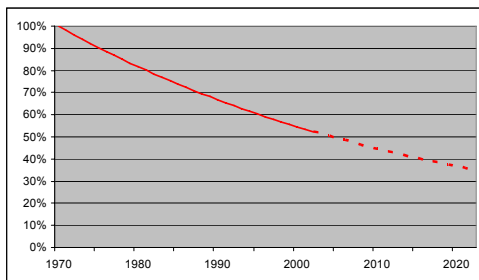
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Problems - data distortions



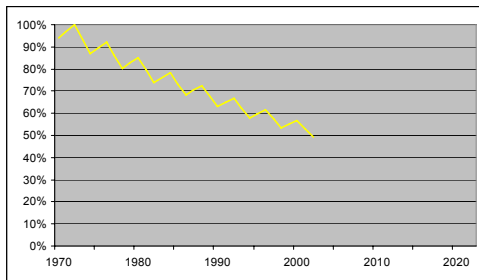
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Problems - data distortions



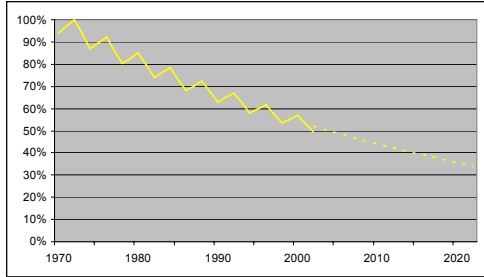
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Problems - data distortions



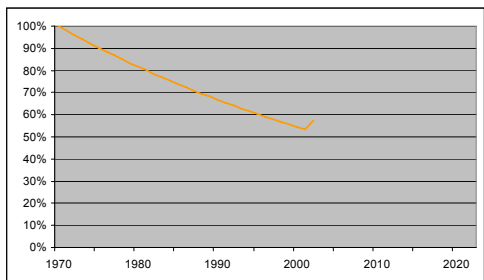
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Problems - data distortions



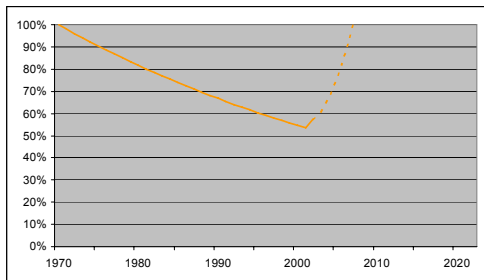
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Problems - data distortions



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Problems - data distortions



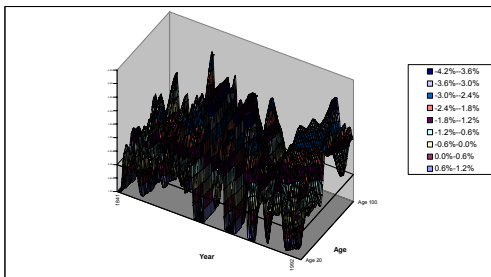
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Validating output

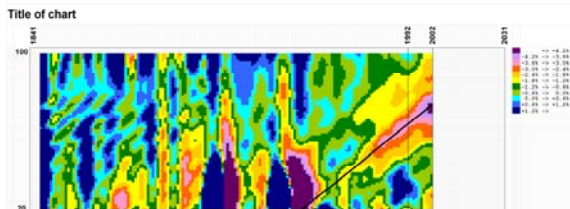
- Annuity factors
- Max / min improvements (over time / ages)
- Equivalent single improvement
- Heat graphs (2-d improvements by age / year)
- Don't forget effect on in-force

Validating output – Heat Graphs

3-d graphs are not easy to interpret !



Validating output – Heat Graphs



- Above is actual past data (not fitted)
- But smoothed (Gaussian 2-d smoothing)
- Thick line is 1930 cohort

Adjustments to output

- Minimum improvements
- Take out smoking prevalence changes in past
- Add any anticipated future changes
- Convert to annuitant improvements
- Over age 100

Derivation of capital

- Model produces mean μ but also $\sigma(\mu)$
- Hence estimate “trend risk”
- But capital required for
 - Trend risk
 - Volatility
 - Miss-pricing risk
 - (Catastrophe)
 - Other capital constituents (ALM, operational)

Derivation of capital

- Model looks at σ of the fitted μ - NOT the raw μ
- So changing smoothing parameters changes σ
- Two projections can give largely similar future μ
- ... but very different implied capital

Derivation of capital

- For example,
 - Population dataset,
 - Years 1841 – 2002, ages 20-100

		Smoothing parameters (age/year)	
		50/75	500/150
Age 65	mean	13.67	13.70
	99.50%	14.18	13.68
	capital %	3.7%	1.3%
Age 75	mean	11.71	11.74
	99.50%	12.18	11.90
	capital %	4.0%	1.4%

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