

REach FOR GOLD



CANADIAN REINSURANCE CONFERENCE / CONGRÈS CANADIEN DE RÉASSURANCE

THE 59th ANNUAL CANADIAN
REINSURANCE CONFERENCE

REach
FOR **GOLD**



Jean-Yves Rioux

Big Data and Analytics' dramatic impacts in the Life Insurance Industry

Agenda

- Drivers and barriers
- Data sources
- The process
- Life and health insurance applications
- Modeling/analyzing longevity



Famous quotes

“Not everything that can be counted counts, and not everything that counts can be counted.” – Albert Einstein

“If you can’t explain it simply, you don’t understand it well enough.”
– Albert Einstein

“Those who do not remember the past are condemned to repeat it.”
– George Santayana

“The price of light is less than the cost of darkness.” – Arthur C. Nielsen

“War is ninety percent information.” – Napoleon Bonaparte

“Facts do not cease to exist because they are ignored.” – Aldous Huxley



The environment

The interest in predictive modeling has increased substantially in recent years.

Drivers

Decrease in cost of storage

Increase in computer processing speed/memory

Proliferation of data from third parties

Increase interest in learning more with regard to own data

Barriers

Quality of data

Rudimentary or basic IT infrastructure

Inadequate size of or expertise from analytics staff

Lack of executive advocacy



Internal data sources

There is increased interest in understanding, restructuring and using own data.

Representative data categories

Policyholder info Age Gender Smoking status Amount inforce Province of residence Occupational code	Past/current claims Date incurred Date reported Amount Cause	Experience results Incidence Lapse Termination Mortality Persistency
Reinsurance info Premium Rates Allowances Coverage	Product design Surrender rights/charges Vesting Market value adj. Renewal rights Rollover/reset rights Conversion rights	Asset info Coupon rates Notional amounts Market values



External data sources

Companies who are succeeding in advanced analytic analysis are doing so by their commitment to exploring new data.

Representative data categories

Data providers

Axiom
 Agriculture and Agri-Food Canada
 AM Best
 AWCBC
 Bank of Canada
 Bloomberg
 BDC
 Canada Hospital Directory
 CMHC
 Canada Revenue Agency
 Canadian Cancer Society
 CIHI
 CMA
 Cap Index
 Citizenship and Immigration Canada
 Dun & Bradstreet
 HRSDC
 Environics
 Equifax
 Experian
 GHDx
 Industry Canada
 Ipsos
 LifeGuide
 Natural Resources Canada
 Office of the Superintendent of Bankruptcy
 Canada
 OECD
 PAHO
 Public Health Agency of Canada
 Public Safety Canada
 Statistics Canada
 Undata
 UIS
 World Bank
 World Values Survey

Economic

Real Estate
 Equities
 Commodities
 Interest Rates
 Foreign Exchange
 Inflation
 Economic/Bus. trends

Demographic

Age
 Gender
 Ethnicity
 Income
 Immigration Data

Financial

Credit Score
 Gross/Total Debt
 Service Ratio
 Credit Ratings

National indices

Wage Data
 Wealth/Net Worth
 Unemployment Stats
 Aggregate CRA Data

DB, LB and health

Death
 Diabetes
 Cancer
 Cardiovascular
 Disability
 Injury
 Depression/Mental

Medical & drugs

Disability Data
 Hospital Directory
 Nursing Home Data
 Hospital Visit Statistics
 Prescription Drug Usage
 Physician Data

Competitive data

Premium Rates
 Crediting Rates
 Guaranteed Rates
 Product Features

Geographic

Crime Statistics
 Climate Data
 Geographic Mapping
 Population Concentration

Behaviors & lifestyle

Physical Activity Level
 Hobbies
 Lifestyle Clusters
 Social Values

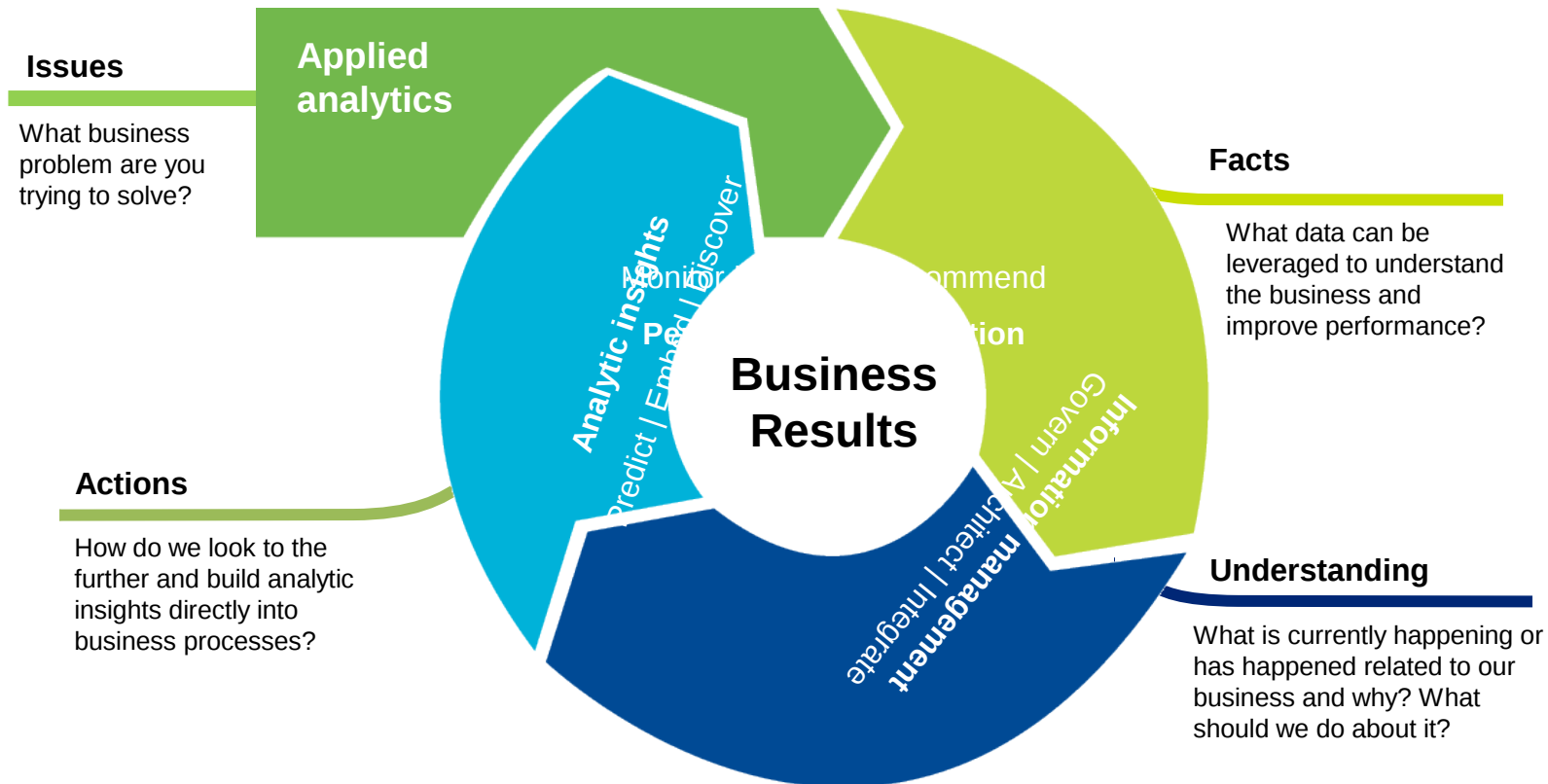
Purchase behaviors

Purchase Propensities
 Spend by Category
 Purchase Triggers

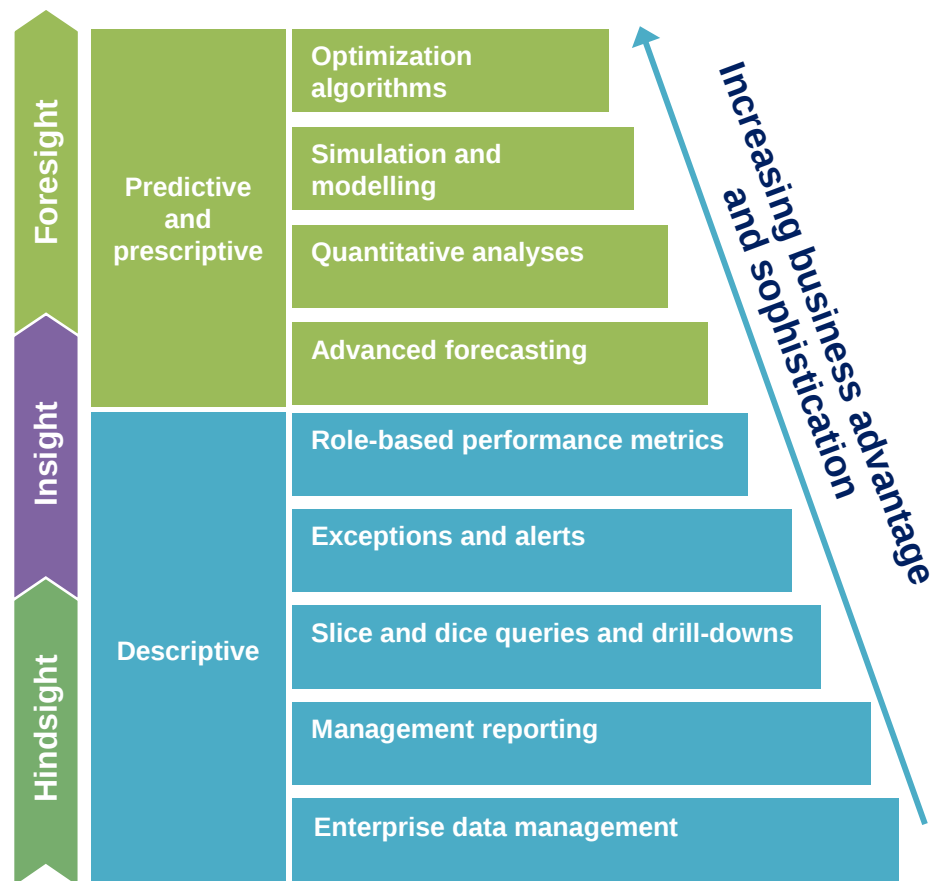
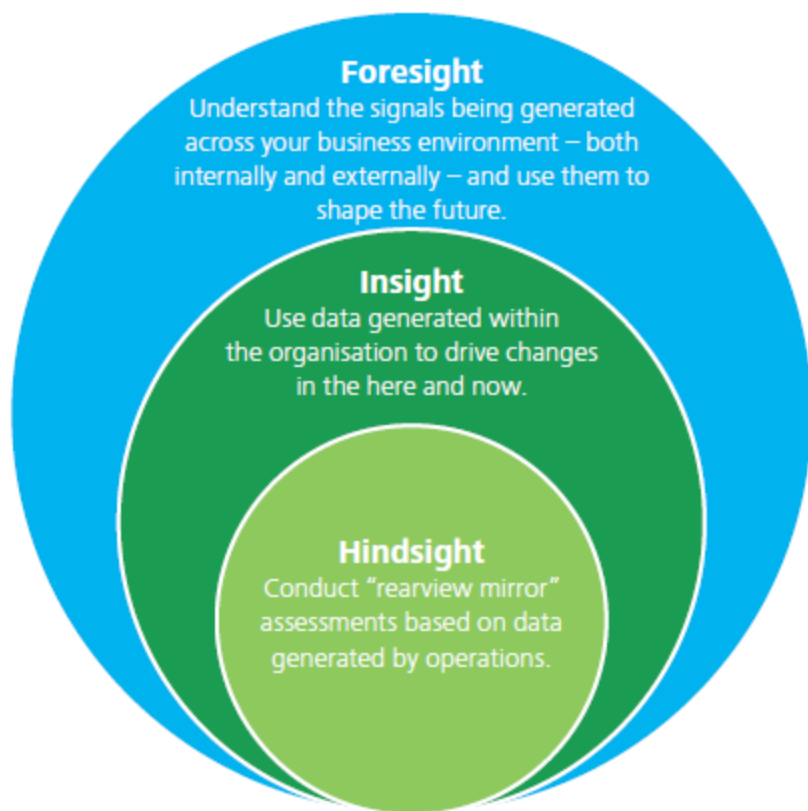


A continuous process...

The process of delivering business analytics results is one of continuous improvement.



... that needs to evolve



Applications – Life and Health

Sales & marketing

- Identify target groups
- Identify characteristics correlated with purchase decision
- Understand purchase behaviors and recommend the right product
- Recruit agents whose characteristics are similar to successful agents
- Monitor existing agents

Claims

- Predict claim frequency and severity
- Prioritize resources
- Identify likely fraudulent/rescinded claims

Pricing/reserving

- Improve pricing accuracy
- Project impact of deviations from pricing parameters
- Reserve more accurately

Underwriting

- Identify best risks and prioritize acceptance efforts
- Identify applicants for whom additional underwriting is needed
- Support simplified underwriting

In-force management

- Identify and retain policyholders likely to surrender
- Offer additional products to current customers
- Profile customers

Experience analysis

- Identify experience drivers
- Handle low credibility data by enhancing the data
- Create own mortality/lapse tables



Application: Modeling Longevity

- **Questions**

- What is the empirical average death rate (DR)?
- What is the empirical implied mortality improvement (MI)?
- What would modeling predict?
- How does the modeled MI compares to CIA promulgated scale?
- Are all variables important for forecasting?



The Data

- Canadian Standard Ordinary Mortality 2005-2012, Canadian Institute of Actuaries [Insured population experience]
- Human Mortality Database 1921-2012, Statistics Canada [General population experience] (as a secondary data set)



Data (CIA CSO) – Empirical rates

Death rates

DR	'05	'06	'07	'08	'09	'10	'11	Avg
20-25	0.0004	0.0003	0.0003	0.0004	0.0003	0.0002	0.0001	0.000
40-45	0.0013	0.0016	0.0018	0.0020	0.0023	0.0032	0.0008	0.002
60-65	0.0088	0.0064	0.0063	0.0067	0.0053	0.0091	0.0057	0.007
80-85	0.0547	0.0498	0.0546	0.0497	0.0410	0.0401	0.0493	0.048

Mortality improvement rates

MI	'06	'07	'08	'09	'10	'11	Avg
20-25	29.2%	3.4%	-48.5%	11.1%	39.5%	29.1%	10.6%
40-45	-20.1%	-12.6%	-13.6%	-13.7%	-41.2%	74.3%	-4.5%
60-65	26.8%	2.7%	-6.5%	20.1%	-69.9%	37.2%	1.7%
80-85	8.9%	-9.5%	8.9%	17.5%	2.1%	-22.9%	1.7%



The Approach

Item	Specifics
Target variable	Death rate (\$ claims/ \$ exposure)
Predictive variables	Year, Sex, Smoker, Type of Underwriting, Insured amount, Duration, Attained Age
Predictive models	<ul style="list-style-type: none">• Generalized Linear Model (GLM)• Lee-Carter (L-C)
Approach	<ul style="list-style-type: none">• Fitted model to 2005-2011 data• Tested the model fit• Tested the predictive powers on 2012• Derived and compared the MI rates



The GLM Model (Probit version)

- $$\phi^{-1}(DR_i) = \beta_0 + \beta_1 x_{1,i} + \dots + \beta_M x_{M,i} + \varepsilon_i$$

Where

- DR_i is the death rate for occurrence i
- $\beta_0 \dots \beta_M$ are the regression coefficients indicating the relative effect of a particular explanatory variable on the outcome
- $x_{1,i} \dots x_{M,i}$ are the explanatory variables
- ε_i is the error term



GLM Model – Predictive variables

- Most important predicting variables have lower p-values
- Eliminated of product type and Preferred classification due to high p-values
- Odd ratios indicate the level of change per unit increase in the variable

Year	Odds ratio
t	0.99

Sex	Odds ratio
Male	1.00
Female	0.84

Duration	Odds ratio
D	1.00

Amount insured	Odds ratio
0-10K	1.00
10K-50K	0.95
50K-100K	0.91
100K-250K	0.87
250K-500K	0.88
500K-1M	0.78
1M+	0.83

Attained Age	Odds ratio
x	1.04

Smoker	Odds ratio
Smoker	1.00
Non-Smoker	0.79
Smoker status unknown	0.93

Type of Underwriting	Odds ratio
Medical	1.00
Non-medical	1.00
Paramedical	0.96
Unknown	0.97



GLM Model – Fit

- **Model is a good fit**

- Coefficients of determination:

- $R^2 = 22\%$

- *Adjusted* $R^2 = 44\%$

- Goodness of fit test (Chi Square) = 5506

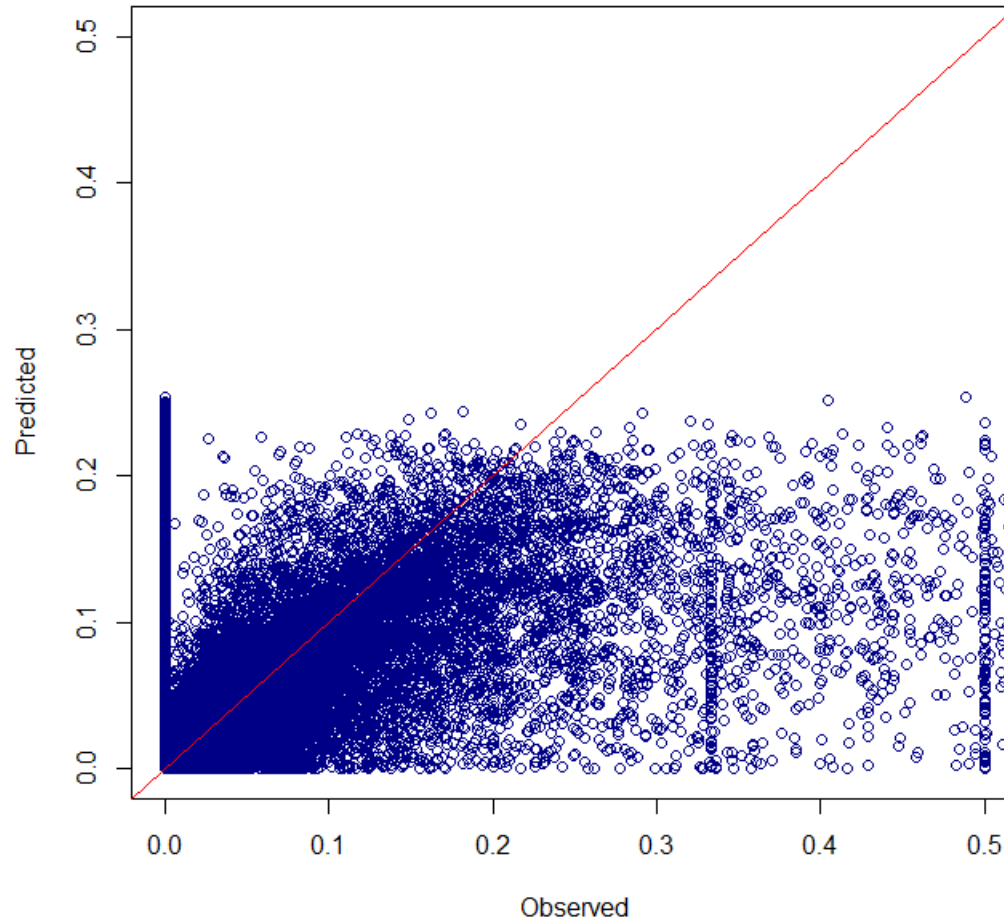
- P-value from Chi Square test <0.001

- Modeling error and projection range



GLM Model – Fit (cont'd)

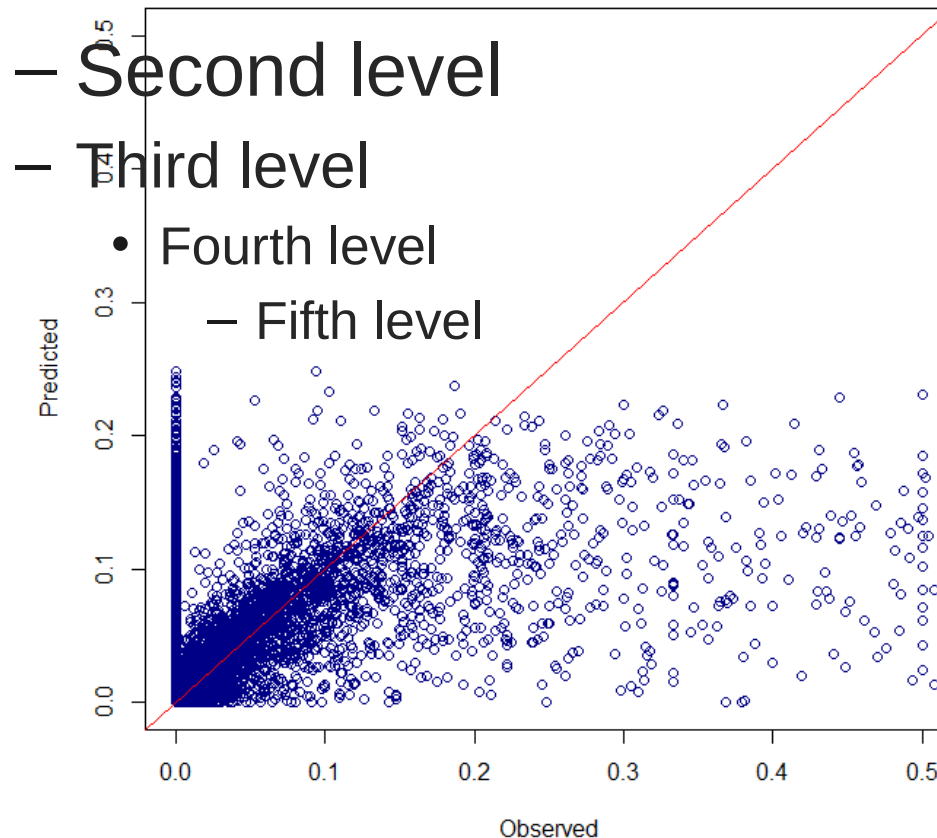
GLM model - Observed DR vs. Predicted DR



GLM Model – Power of Predictability

Coefficient of determination (R^2) for 2012 predicted DR is 22%

- Click to edit Master text styles



GLM Model – Results

Death rates

DR	'05	'06	'07	'08	'09	'10	'11	Avg
20-25	0.00002	0.00002	0.00002	0.00002	0.00002	0.00002	0.00002	0.00002
40-45	0.00049	0.00049	0.00048	0.00048	0.00047	0.00047	0.00047	0.00048
60-65	0.00738	0.00729	0.00725	0.00716	0.00707	0.00701	0.00692	0.00715
80-85	0.05679	0.05642	0.05551	0.05517	0.05486	0.05438	0.05402	0.05531

Mortality improvement rates

MI	'06	'07	'08	'09	'10	'11	Avg
20-25	1.5%	1.1%	1.0%	2.9%	0.2%	1.2%	1.3%
40-45	0.9%	0.5%	1.5%	0.9%	0.9%	0.9%	0.9%
60-65	1.1%	0.6%	1.3%	1.2%	0.8%	1.3%	1.0%
80-85	0.7%	1.6%	0.6%	0.6%	0.9%	0.7%	0.8%



The Lee-Carter Model

- $$\log(DR_{xt}) = a_x + b_x k_t + \varepsilon_{xt}$$

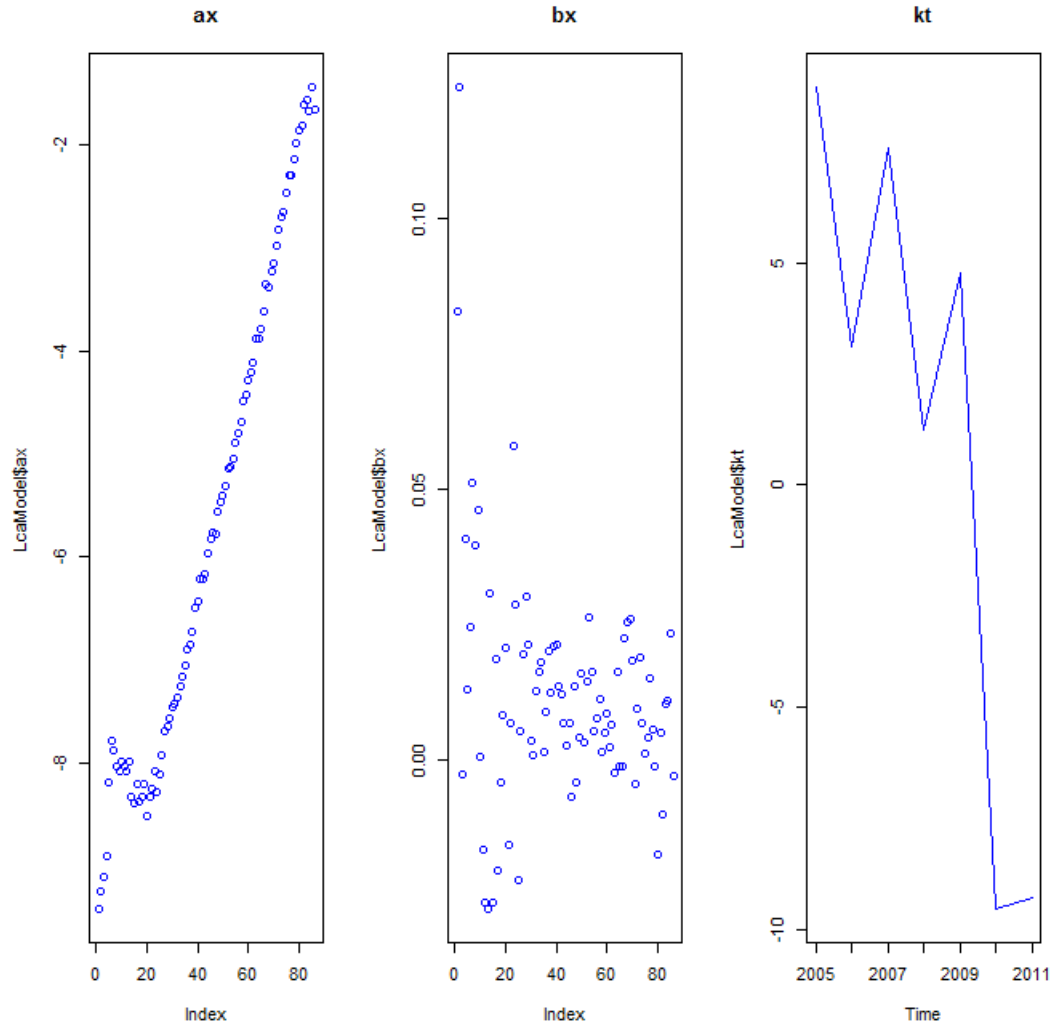
Where

- DR_{xt} is the death rate for age x at time t
- a_x and b_x are the regression coefficients relating to the age
- k_t is the regression coefficient relating to the time
- ε_{xt} is the error term

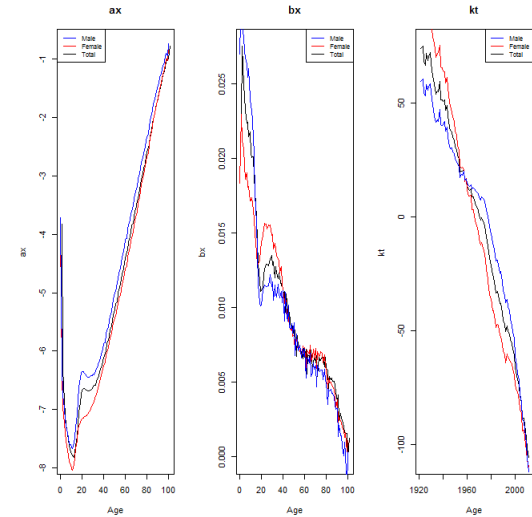


LC Model – Calibration

CIA CSO Data



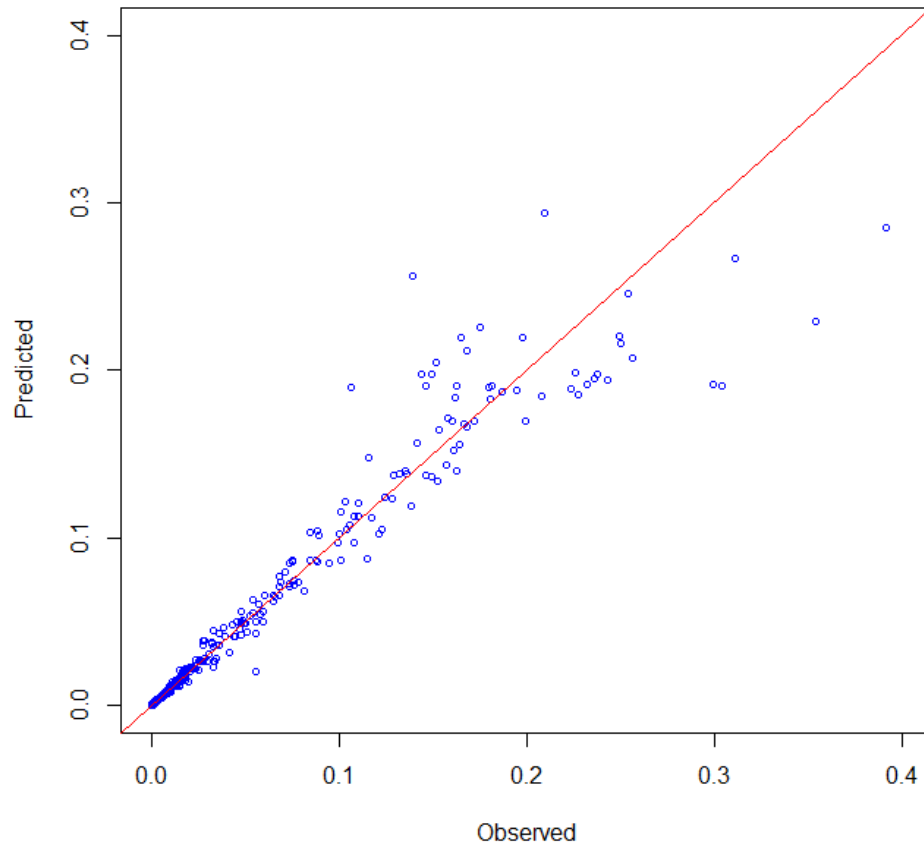
HMD Data



LC Model – Fit

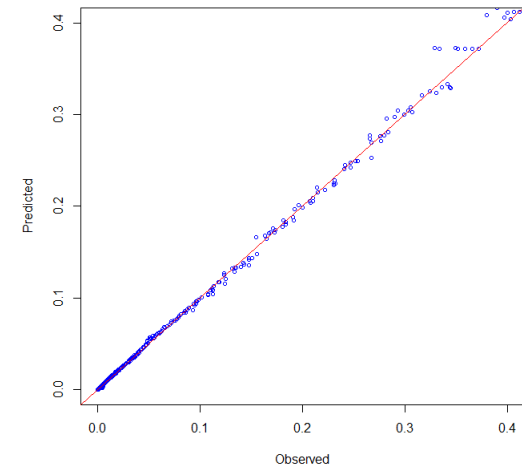
CIA CSO Data

LC CIA model - Observed DR v.s. Predicted DR



HMD Data

LC HMD model - Observed DR v.s. Predicted DR



LC Model – Power of Predictability

- *Coefficient of determination (R^2)* for 2012 observed v.s. predicted death rate is 85%

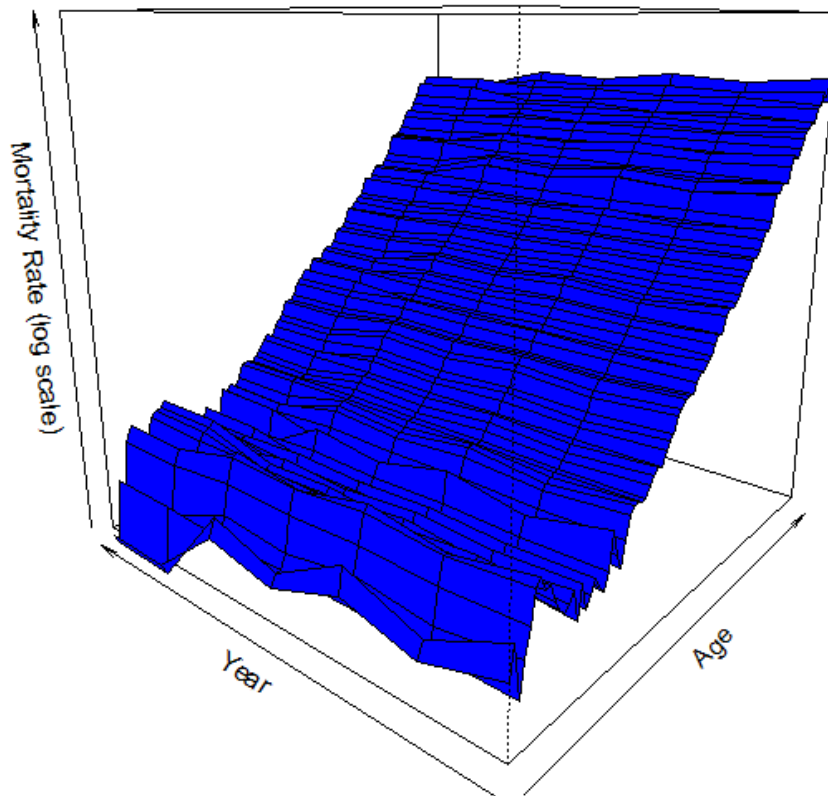


LC Model – Power of Predictability

CIA CSO Data

Lee-Carter Model Prediction based on CIA Data

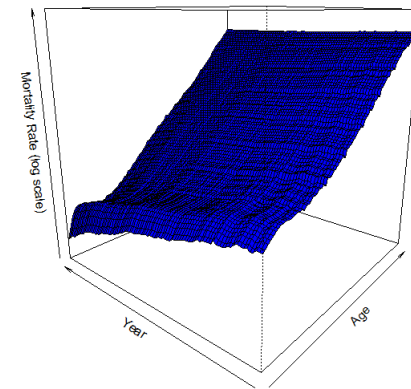
Canada 2005 - 2011



HMD Data

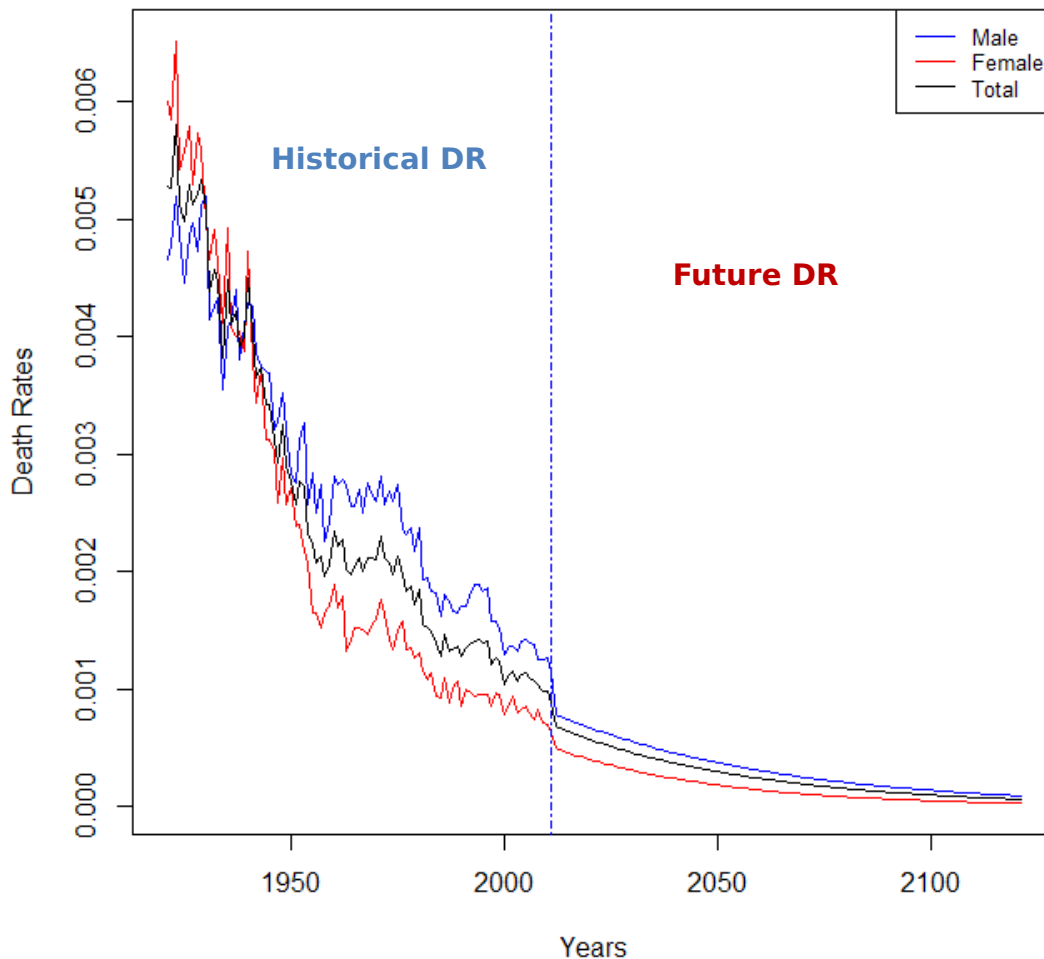
Lee-Carter Model Prediction based on HMD Data

Canada 1921 - 2011



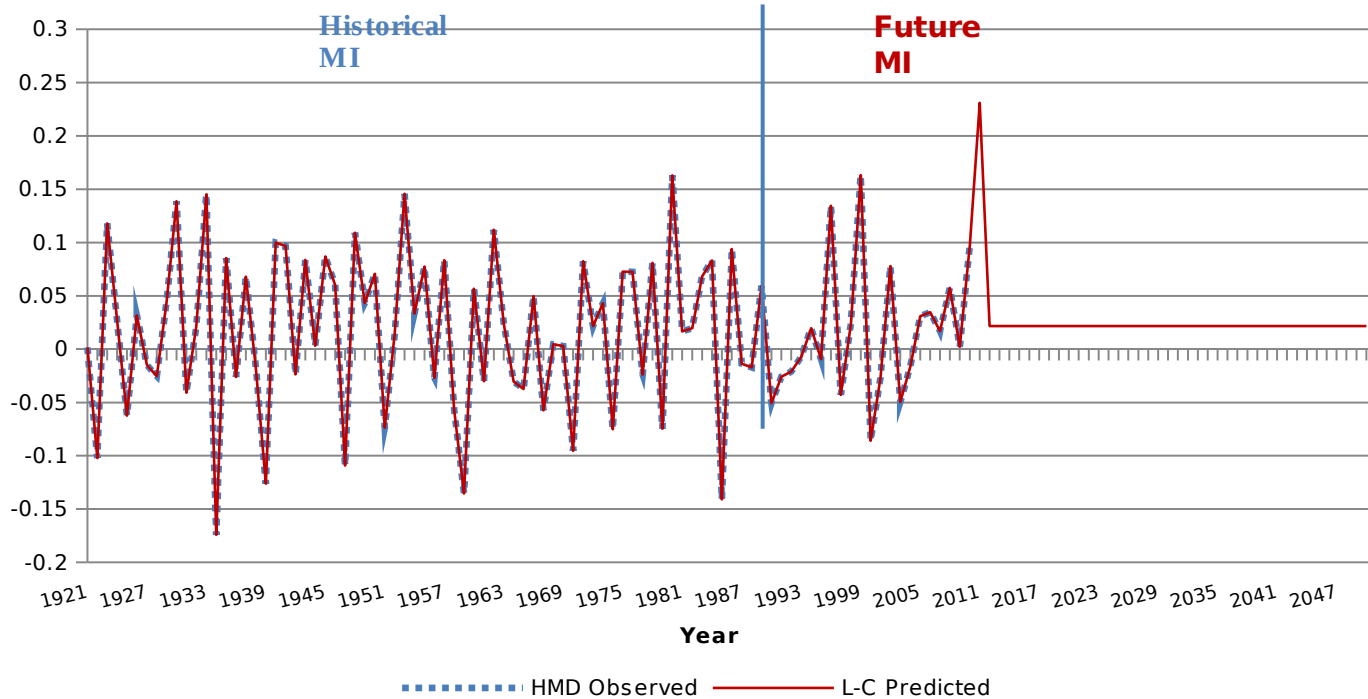
LC Model – DR Illustration

LC HMD model - DR for age 40



LC Model – MI Illustration

Observed v.s. predicted MI for age 40 based on HMD



LC Model – Results

Death rates

DR	'05	'06	'07	'08	'09	'10	'11	Avg
20-25	0.00043	0.00036	0.00041	0.00034	0.00038	0.00027	0.00027	0.000
40-45	0.00060	0.00055	0.00059	0.00053	0.00056	0.00046	0.00046	0.001
60-65	0.00436	0.00419	0.00432	0.00414	0.00424	0.00387	0.00388	0.004
80-85	0.04702	0.04262	0.04591	0.04133	0.04380	0.03502	0.03515	0.042

Mortality improvement rates

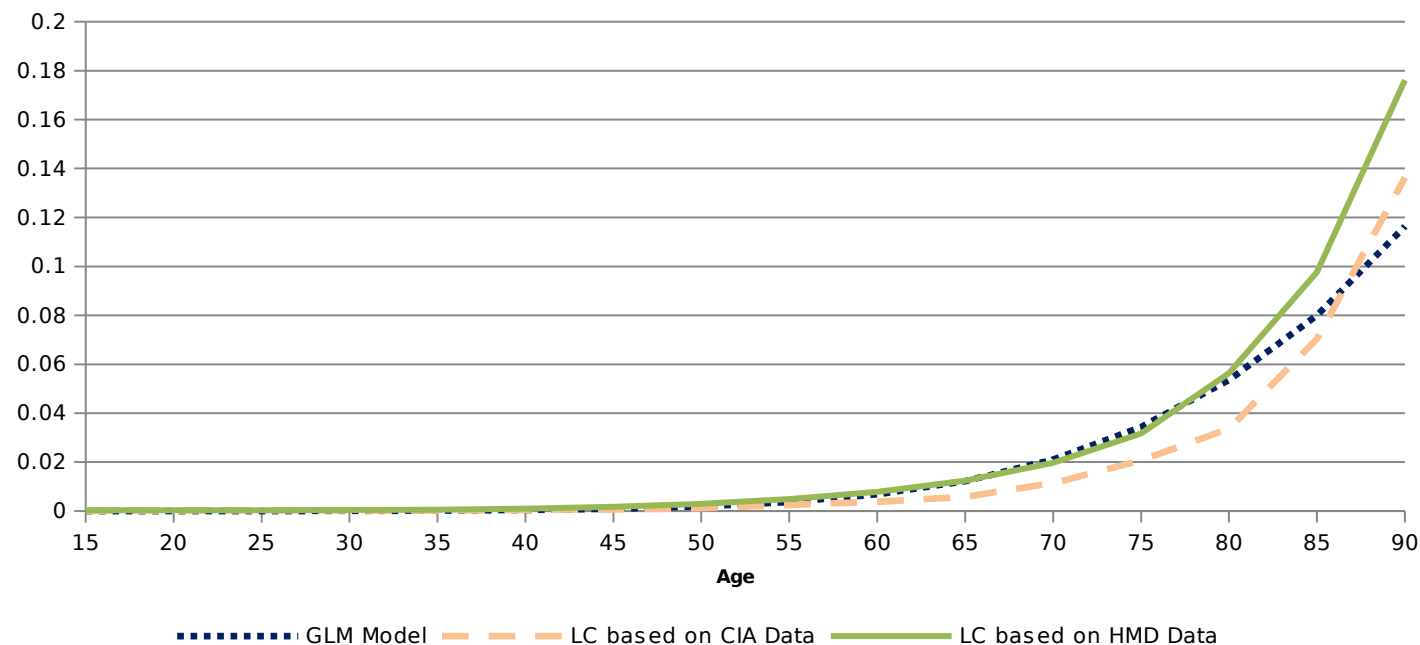
MI	'06	'07	'08	'09	'10	'11	Avg
20-25	15.5%	-13.5%	16.1%	-10.0%	27.9%	-0.4%	6.0%
40-45	8.3%	-6.8%	8.9%	-5.2%	17.9%	-0.3%	3.8%
60-65	3.9%	-3.0%	4.1%	-2.4%	8.6%	-0.1%	1.8%
80-85	9.4%	-7.7%	10.0%	-6.0%	20.0%	-0.4%	4.2%



Comparison of the results

Comparison GLM and LC models

Predicted DR for year 2012



Comparison of the results (cont'd)

Comparison of mortality improvement rates
GLM, L-C and CIA promulgated

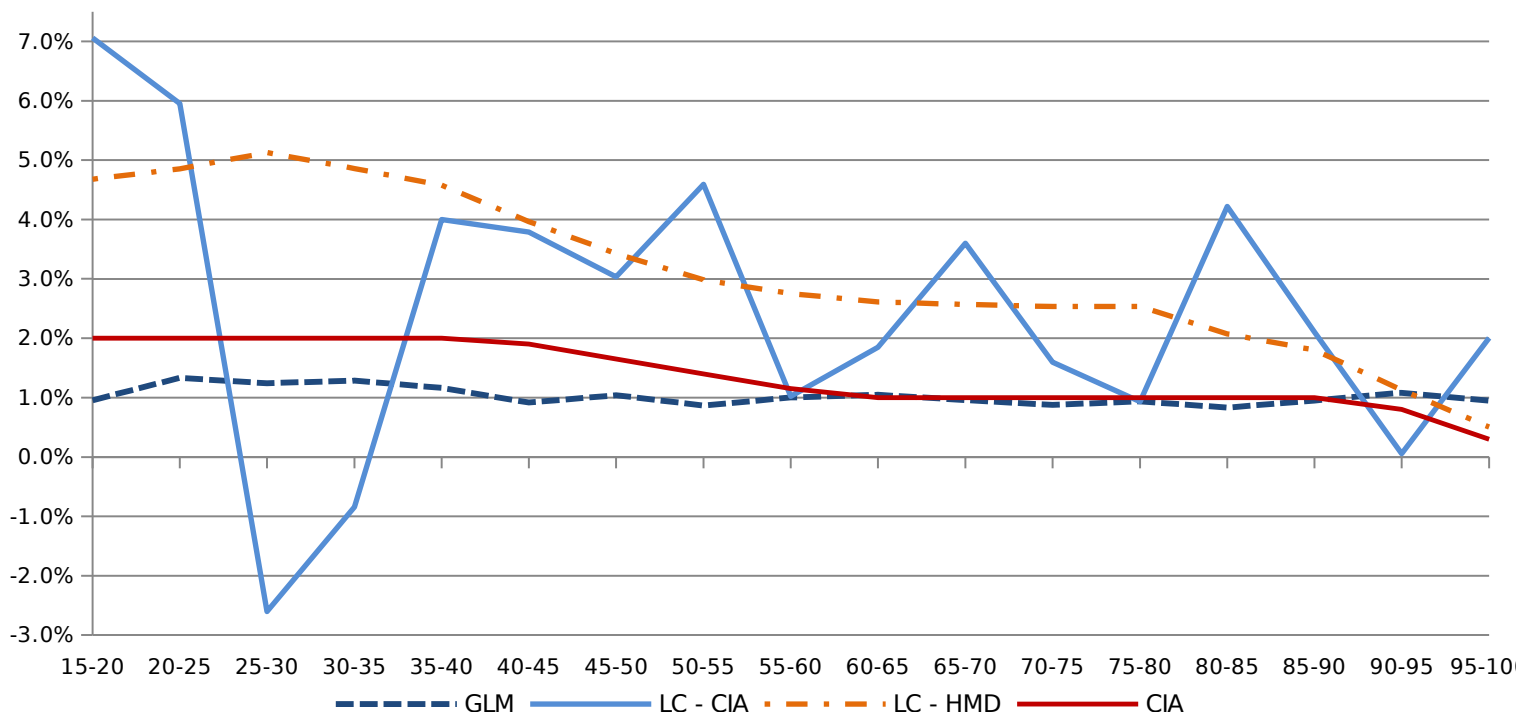
Average MI	Empirical data	GLM	Lee Carter CIA CSO	Lee Carter HMD	CIA
20-25	10.6%	1.3%	6.0%	4.9%	2.0%
30-35	-14.4%	1.3%	-0.8%	4.9%	2.0%
40-45	-4.5%	0.9%	3.8%	4.0%	1.9%
50-55	-7.3%	0.9%	4.6%	3.0%	1.4%
60-65	1.7%	1.0%	1.8%	2.6%	1.0%
70-75	-1.4%	0.9%	1.6%	2.5%	1.0%
80-85	1.7%	0.8%	4.2%	2.1%	1.0%
90-95	-7.0%	1.1%	0.1%	1.1%	0.8%



Comparison of the results (cont'd)

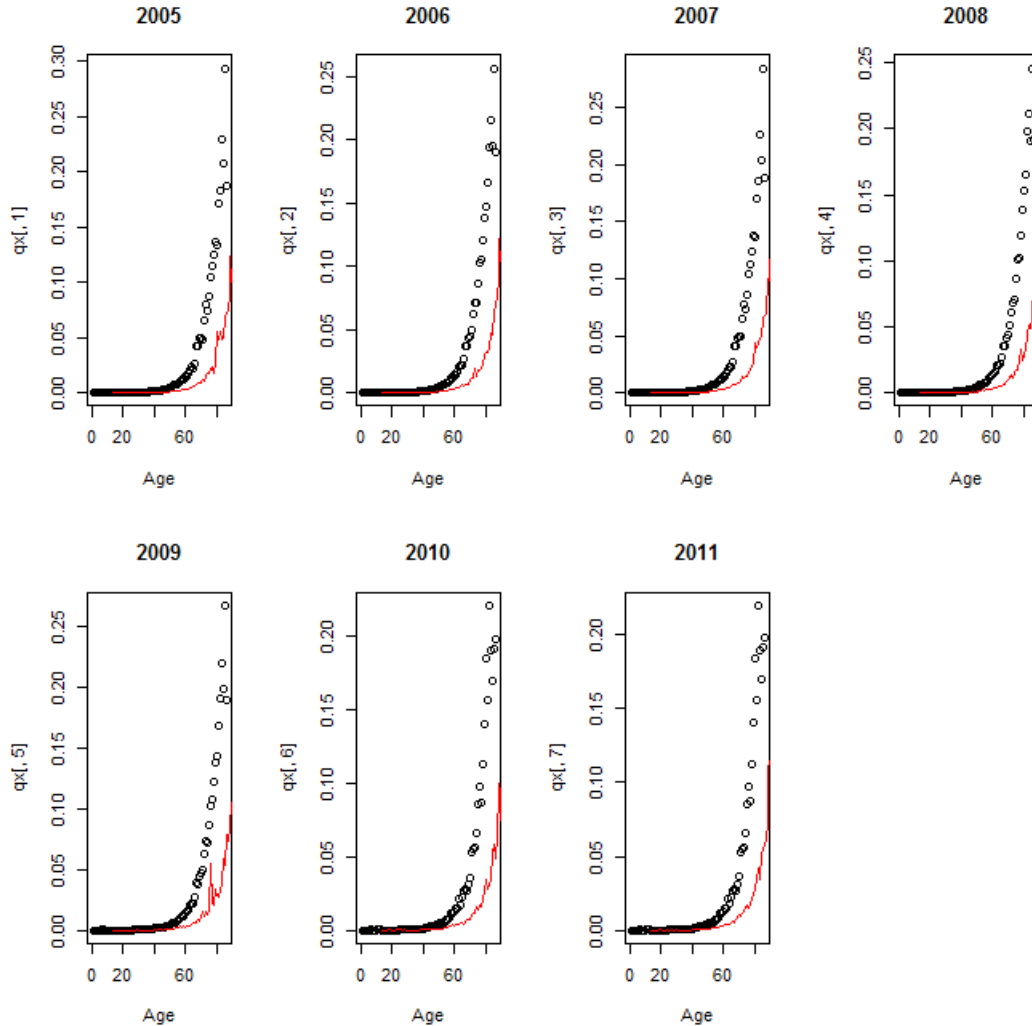
Comparison of mortality improvement rates GLM, L-C and CIA promulgated

Average MI Comparison

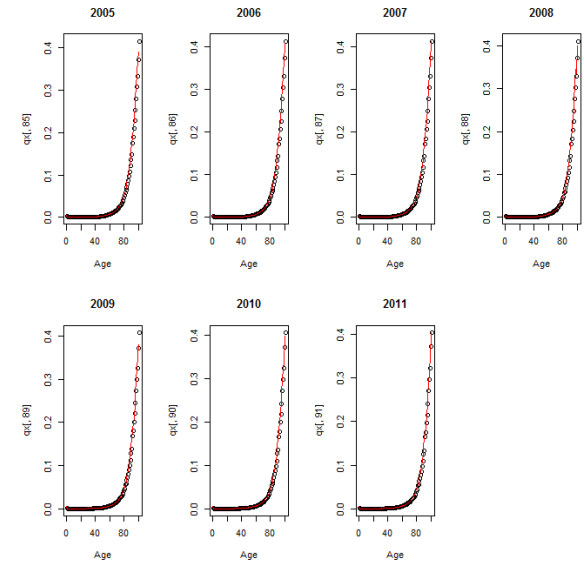


Appendix – LC Model – Fit

CIA CSO Data

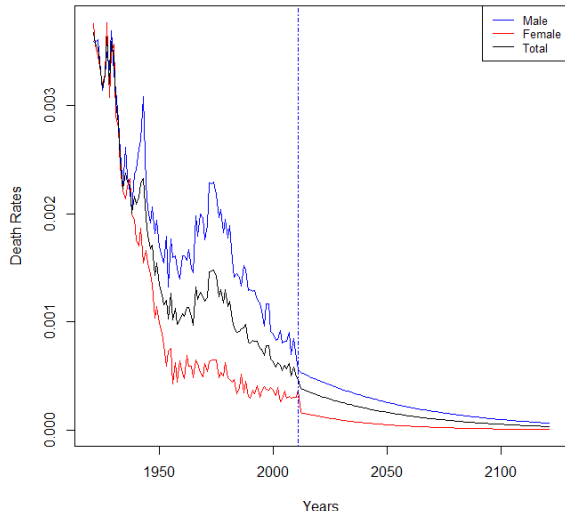


HMD Data

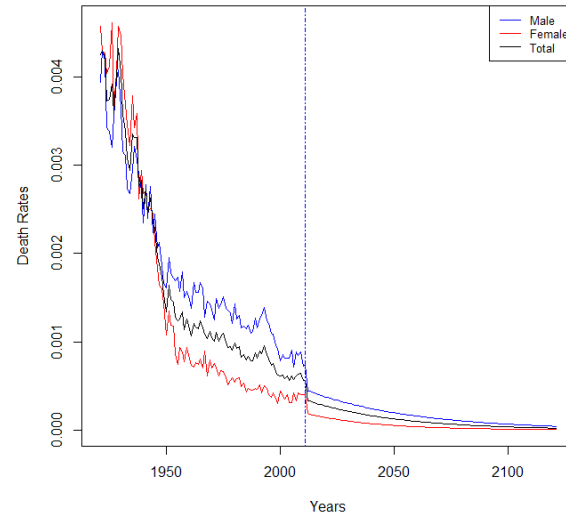


Appendix – LC Model – Illustrations

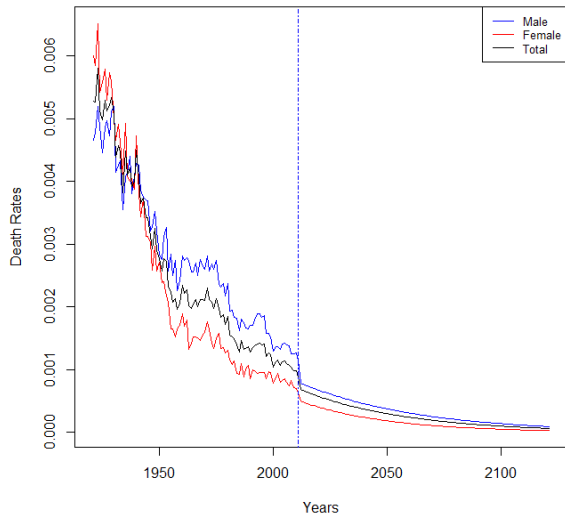
LC HMD model - DR for age 20



LC HMD model - DR for age 30



LC HMD model - DR for age 40



LC HMD model - DR for age 50

