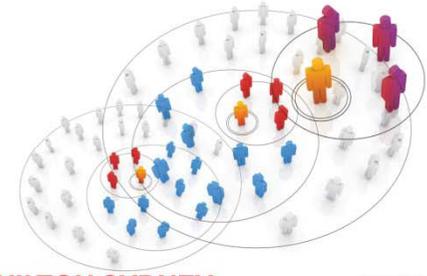




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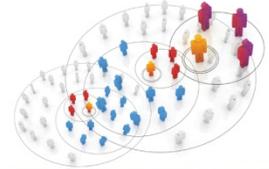
Heterogeneity of Australian Population Mortality, and Implications for a Viable Life Annuity Market

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University of New South Wales

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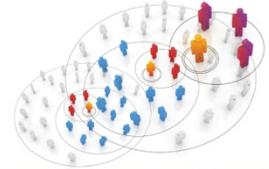
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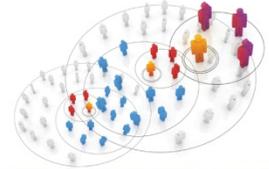
Agenda

- Introduction
- Model Specification
 - Frailty Model
 - Markov Aging Model
- Data
- Results
 - Model Fitting
 - Impact on Annuity Rates
- Conclusion



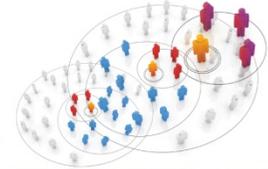
Background

- Risks in an insured portfolio are heterogeneous
 - Ignoring heterogeneity: adverse selection
 - Risk rating using risk factors
- Life Annuity
 - Limited use of rating factors
 - Demand for annuity products
 - Understanding heterogeneity
 - Pricing and adverse selection in annuity business
 - Underestimation of mortality improvement at higher ages



Research Objective

- Quantifying mortality heterogeneity of Australian population
 - Frailty model (Vaupel et al. 1979)
 - Markov aging model (Lin and Liu 2007)
- Projection of mortality rates at higher ages, taking into account heterogeneity
- Impact of allowing for heterogeneity on life annuity pricing
 - Difference in annuity prices between heterogeneous lives
 - Impact on annuity business



Frailty Factor

- An unobserved mortality risk factor, fixed at birth
- Mathematically defined in terms of force of mortality:

$$\mu(x, z) = z \cdot \mu(x, 1)$$

- Assumed form of standard force of mortality and frailty distribution

- Standard force of mortality $\mu(x, 1) = \alpha \cdot e^{\beta x}$

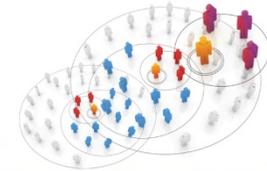
- Frailty distribution

- Gamma

$$f_Z(z) = \frac{\lambda^k}{\Gamma(k)} \cdot z^{k-1} \cdot e^{-\lambda \cdot z}$$

- Inverse Gaussian

$$f_Z(z) = \left(\frac{\delta}{\pi}\right)^{\frac{1}{2}} \cdot e^{\sqrt{4\delta\theta}} \cdot z^{-\frac{3}{2}} \cdot e^{-\theta z - \frac{\delta}{z}}$$



Frailty Model

- Distribution of frailty at age x

- Gamma distribution

$$f_{Z|X}(z|X = x) = \frac{(\lambda(x))^k}{\Gamma(k)} \cdot z^{k-1} \cdot e^{-\lambda(x) \cdot z}$$

with

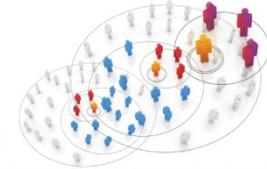
$$\lambda(x) = \lambda + H(x, 1), \text{ and } E[z] = \frac{k}{\lambda(x)}, \quad \text{Var}[z] = \frac{k}{(\lambda(x))^2}$$

- Inverse Gaussian distribution

$$f_{Z|X}(z|X = x) = \left(\frac{\delta}{\pi}\right)^{\frac{1}{2}} \cdot e^{\sqrt{4\delta\theta(x)}} \cdot z^{-\frac{3}{2}} \cdot e^{-\theta(x) \cdot z - \frac{\delta}{z}}$$

with

$$\theta(x) = \theta + H(x, 1), \text{ and } E[z] = \left(\frac{\delta}{\theta(x)}\right)^{\frac{1}{2}}, \quad \text{Var}[z] = \frac{1}{2} \sqrt{\frac{\delta}{(\theta(x))^3}}$$



Maximum Likelihood Estimation

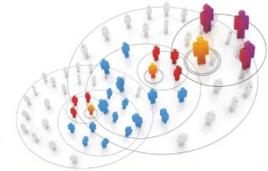
- Mean frailty approach
 - Assumes the average force of mortality is the cohort force of mortality
- Does not take into account the impact of heterogeneity on the variability
- Normal approximation for sample mean mortality rates
 - The observed cohort is a sample of size E_x of the population
 - According to CLT, the observed sample mean force of mortality is normally distributed

- Gamma:

$$E[\hat{\mu}_x] = \mu(x, 1) \cdot \frac{k}{\lambda + H(x, 1)}, \quad \text{Var}[\hat{\mu}_x] = \frac{(\mu(x, 1))^2 \cdot k}{E_x \cdot (k + H(x, 1))^2}$$

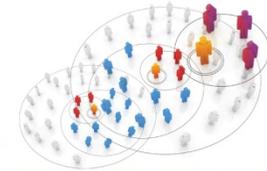
- Inverse Gaussian

$$E[\hat{\mu}_x] = \mu(x, 1) \cdot \left(\frac{\delta}{\theta + H(x, 1)} \right)^{\frac{1}{2}}, \quad \text{Var}[\hat{\mu}_x] = \frac{(\mu(x, 1))^2}{2 \cdot E_x} \sqrt{\frac{\delta}{(\theta + H(x, 1))^3}}$$



Markov Aging Model

- Aging process can be modeled in terms of changes in physiological functions
- Studies in human body functions reveal that functional variables decline roughly linearly after age 30
- Physiological age: represent the degree of aging in human body
 - Change in physiological age represents the decline in human body function
 - High physiological age represents higher probability of dying
 - Mathematically, transition is random in nature



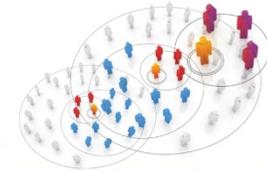
Markov Aging Model

- Markov process with n transient states and 1 absorbing death state, describing the aging process of human beings

$$\Lambda = \begin{pmatrix} -(\lambda_1 + q_1) & \cdots & 0 & 0 & \cdots & 0 \\ \cdots & \cdots & -(\lambda_k + q_k) & \lambda_k & \cdots & 0 \\ 0 & \cdots & 0 & -(\lambda + \gamma + \alpha e^{\beta(k+1)}) & \cdots & 0 \\ 0 & \cdots & 0 & 0 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & 0 & 0 & \cdots & -(\alpha + e^{\beta n}) \end{pmatrix}$$

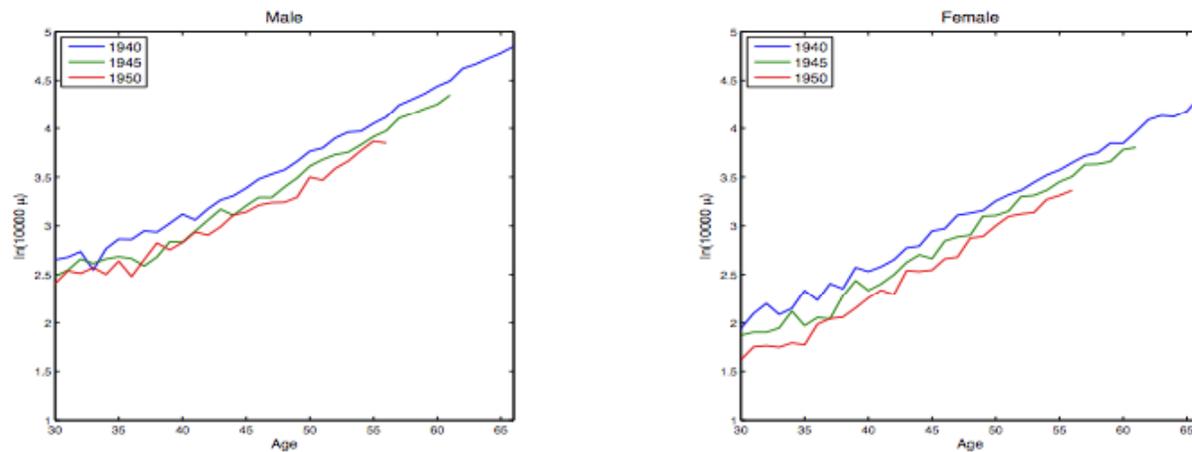
- $\lambda_i = \lambda$ for $i = 5, 6, \dots, n$
- Death rates for $i = 5, 6, \dots, n - 1$

$$q_i = \begin{cases} \gamma + \gamma_1 + \alpha e^{\beta i} & : \text{ for } i_1 < i < i_2 \\ \gamma + \alpha e^{\beta i} & : \text{ otherwise} \end{cases}$$
- Time to death follows phase-type distribution with $\hat{S}(t) = \alpha \exp(\Lambda t) e$
- $\hat{q}_x = \frac{\hat{S}_x - \hat{S}_{x+1}}{\hat{S}_x}$
- Weighted least square estimation: $\sum_x (q_x - \hat{q}_x)^2 \cdot w_x$

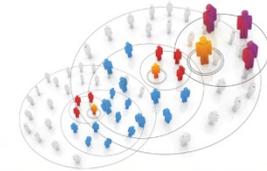


Data for Frailty Model

Figure: Observed Cohort Force of Mortality: Log Transform

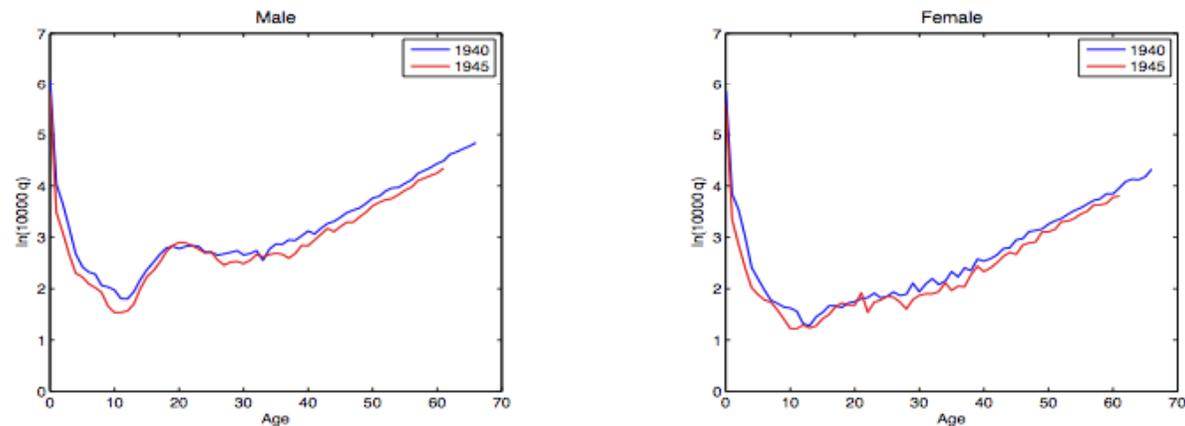


Note: Cohort force of mortality is estimated by the central death rate for birth cohort 1940, 1945, and 1950, both male and female

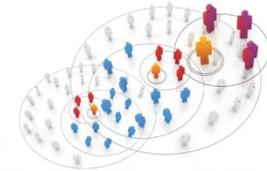


Data for Markov Aging Model

Figure: Observed Cohort Death Probability: Log Transform

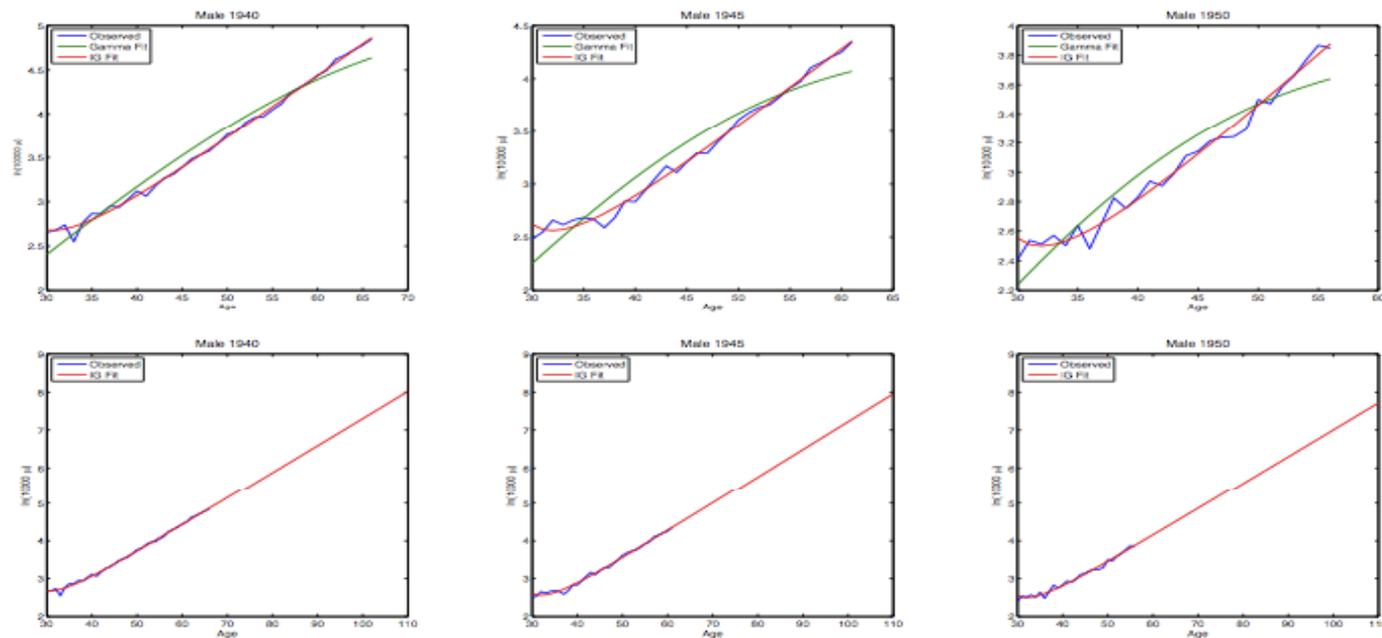


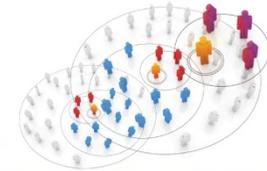
Note: Observed death probability is estimated from central death rate $q_x = \frac{m_x}{1+1/2m_x}$ for birth cohort 1940, and 1945, both male and female



Fitting for Frailty Model

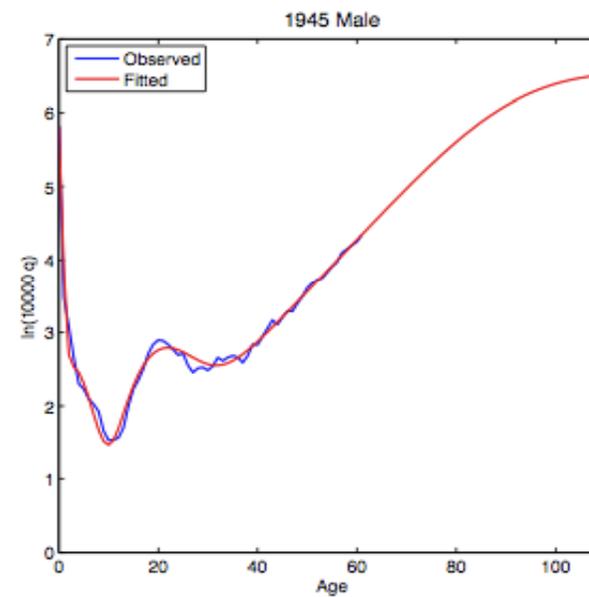
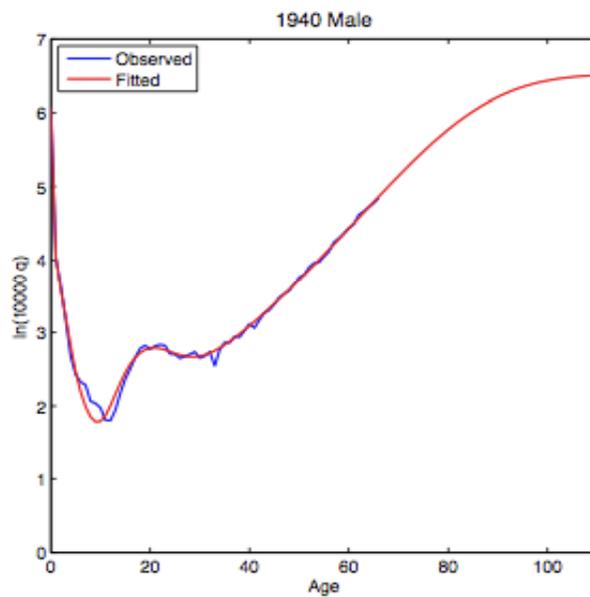
Figure: Observed v.s. Fitted Cohort Average Force of Mortality: Male

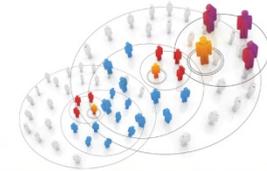




Fitting for Markov Aging Model

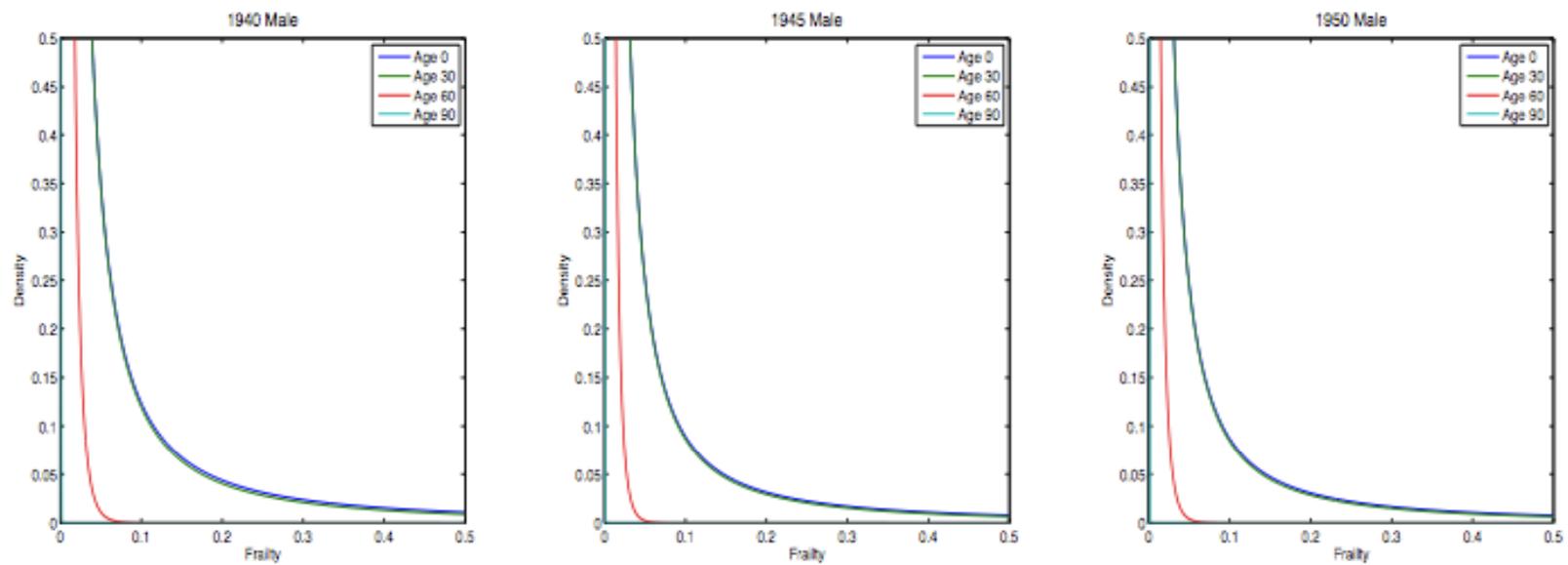
Figure: Observed v.s. Fitted Death Probability with Projection at Higher Ages

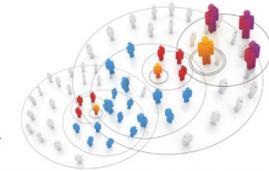




Distribution of Frailty

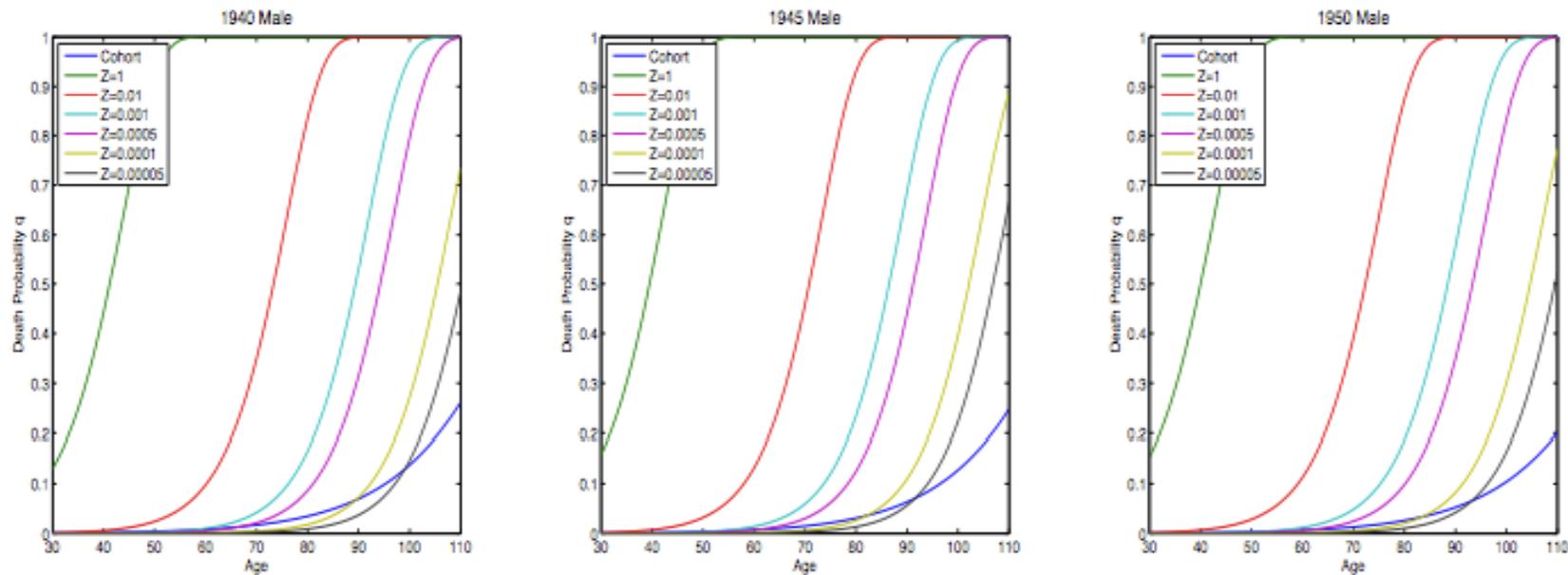
Figure: Distribution of Frailty at different ages

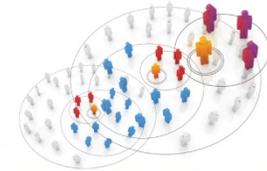




Mortality Rates of Individuals with Different Frailty

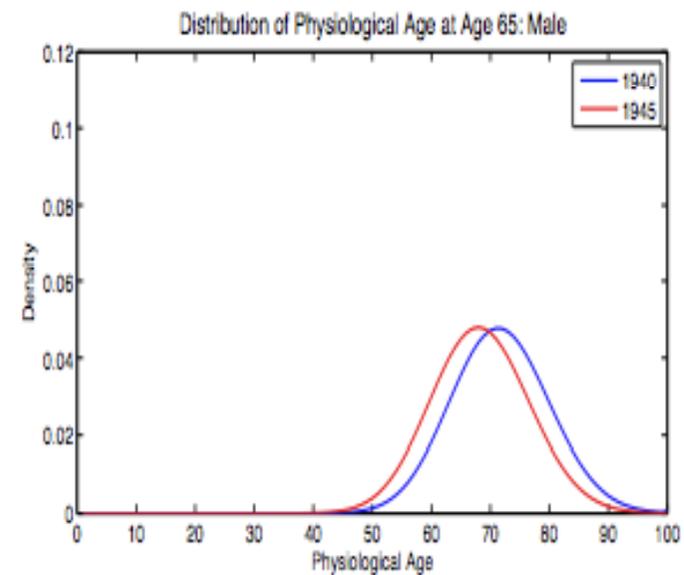
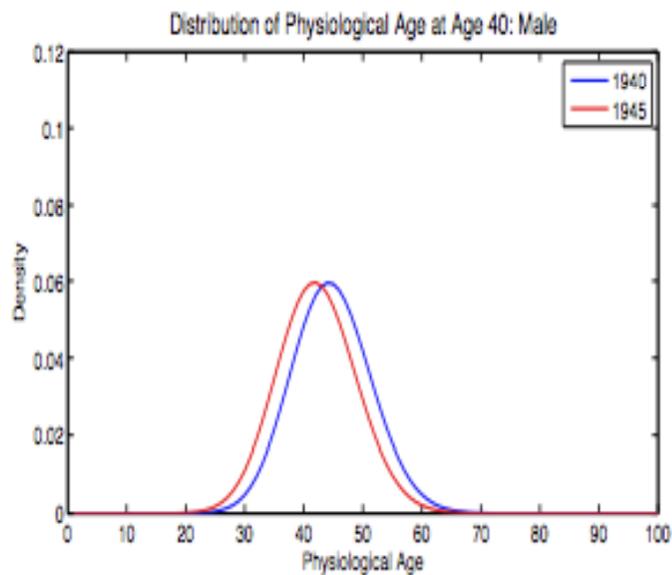
Figure: Mortality Rates of Individuals with Different Frailty: Male

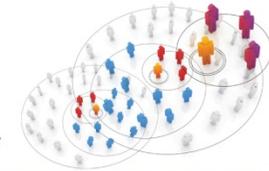




Distribution of Physiological Age at Each Age

Figure: Distribution of Physiological Age at Different Ages: Male





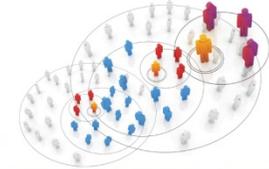
Annuity Rates for Heterogeneous Individuals

Table: Annuity Rate for Individuals with Different Frailty

Frailty	Cohort	0.00005	0.0001	0.0002	0.0005	0.001	0.01
q_{65}	0.012	0.001	0.002	0.004	0.010	0.021	0.189
Whole Life Annuity	\$14.31	\$18.12	\$16.36	\$14.38	\$11.49	\$9.18	\$2.59
Deferred Life Annuity	\$2.36	\$4.32	\$2.94	\$1.66	\$0.47	\$0.08	\$0.00
F(z)		19.40%	38.26%	56.95%	76.49%	86.70%	99.59%

Table: Annuity Rates for Individuals with Different Physiological Age

Physiological Age j	64	68	73	77	81	94
q_{65}	0.006	0.008	0.011	0.015	0.019	0.047
Whole Life	\$19.44	\$18.31	\$16.83	\$15.63	\$14.44	\$11.15
Deferred Life	\$5.34	\$4.55	\$3.63	\$2.99	\$2.46	\$1.50
F(j)	19.47%	35.49%	59.01%	75.81%	87.83%	99.60%



Summary of Results

- Both models indicate that mortality heterogeneity for Australian population is significant, but the heterogeneity structure and estimated results are different
- Frailty model
 - Distribution of frailty is heavily skewed
 - Small proportion of high risks will die first: comply with observed causes of death pattern
 - Heavily depend on assumptions
 - Less practical since frailty factor is unobserved, hard to link it to real world observation
- Markov Aging Model
 - Heterogeneity increases with age
 - More practical since physiological age is easier to be linked with observed health conditions