



Institute and Faculty of Actuaries



**The Effect of Model Uncertainty on the Pricing of Critical Illness Insurance**

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Work with

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- Howard Waters
- Andrew Stott
- David Wilkie

Funded by the IFoA

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**Outline:**

1. Critical Illness Insurance (CI)
2. Data & Problems
3. Claim delay distribution modelling
4. CI diagnosis rates under uncertainty
5. CI pricing under uncertainty

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## Critical Illness: Policy description

- Fixed term policy, usually ceasing at age 65
- A fixed sum insured payable on the diagnosis of one of a specified list of illnesses
- Benefit type:
  - Full acceleration (FA): Death is included as a critical illness (88%)
  - Stand alone (SA): Death is not included as a critical illness (12%)
- Covers:
  - Cancer; Death; Heart attack; Stroke; Multiple Sclerosis; Total & permanent disability; Coronary artery bypass graft; Kidney failure; Major organ transplant; Other.

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## Data

CI data for 1999 – 2005 supplied by CMI:

- Details of policies inforce at the start and end of each year
  - 18,000,000 policy-years of exposure
- Details of claims settled in 1999 – 2005
  - 19,000 claims

Covariates in the data:

Covariate	Number of levels
Age	Numerical
Sex	2 (Female = 0)
Smoker status	2 (NS = 0)
Policy duration	Numerical
Office	13
Benefit type	2 (FA = 0 & SA)
Benefit amount	Numerical
Policy type	2 (Single/Joint life = 0)
Settlement year	Numerical
Cause	10

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## Diseases covered

Critical illnesses and percentage of claims in 1999 – 2005

Critical Illness	% claims	Critical Illness	% claims
Cancer	49.0	Total & permanent disability (TPD)	2.6
Death	17.6	Coronary artery bypass graft (CABG)	2.1
Heart attack (HA)	11.6	Kidney failure (KF)	0.6
Stroke	5.4	Major organ transplant (MOT)	0.2
Multiple sclerosis (MS)	4.3	Other causes	6.6
Males	57.3	Non-smokers	73.9
Females	42.7	Smokers	26.1

Source: Continuous Mortality Investigation, UK

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**Institute and Faculty of Actuaries** **Modelling & estimation** **HERIOT WATT UNIVERSITY**

- Estimation & smoothing of CI diagnosis rates
  - how do these depend on **risk factors?**
- Diagnosis is the **insured event and there is a delay between diagnosis and settlement**
  - diagnosis date often not recorded (18%); need to model it
  - does delay also depend on risk factors?
- The **exposure corresponds to claims settled, not to claims diagnosed**; need to adjust it
- Premium **pricing**
- Also take into account uncertainty

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**Institute and Faculty of Actuaries** **Diagnosis to settlement delay** **HERIOT WATT UNIVERSITY**

Mean Delay = 185 days; SD Delay = 263 days

Fit a **claim delay distribution (CDD)**:

$F^{(j)}(u : x; \theta) = \Pr[\text{claim diagnosed age } x, \text{ cause } j, \text{ covariates } \theta, \text{ will be settled within } u \text{ years}]$

- Estimate missing dates of diagnosis as:
  - Date of settlement – median of appropriate CDD
  - (posterior distribution available)
- Use the CDD to adjust the exposure
- Office-specific growth weights introduced

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**Institute and Faculty of Actuaries** **Observed delay & null model fit** **HERIOT WATT UNIVERSITY**

Fig. 1. Histogram of claims settlement delay (in days). (a) Burr distribution. (b) Lognormal distribution.

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### Claim Delay (D) Distribution



Include **risk factors** in GLM setting:

$M_1: D_i \sim \text{LN}(\mu_i, \sigma^2)$

$$E(D_i) = \exp(\eta_i + \sigma^2/2)$$

where

$$\eta_i = \mu_i = \beta_0 + \sum_{j=1}^8 \beta_j z_{ij} + \beta_{9,k} + \beta_{10,l}$$

$M_2: D_i \sim \text{Transformed Gamma}(\alpha, \tau, s_i)$

$$f_D(u) = \frac{\tau(u/s)^{\alpha\tau} \exp(-u/s)^\tau}{u\Gamma(\alpha)}, \quad E(D_i) = \exp(\eta_i)$$

where  $\eta_i$  as above and  $s_i$  given as function of  $\eta_i, \alpha, \tau$ .

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### Claim Delay (D) Distn (cont)



$M_3: D_i \sim \text{Burr}(\alpha, \tau, s_i)$

$$f_D(u) = \frac{\alpha\tau(u/s)^\tau}{u[1+(u/s)^\tau]^{\alpha+1}}, \quad E(D_i) = \exp(\eta_i)$$

with  $s_i$  given as function of  $\eta_i, \alpha, \tau$ .

$M_4: D_i \sim \text{Generalised beta}(\alpha, \tau, \gamma, s_i)$

$$f_D(u) = \frac{\Gamma(\alpha + \gamma)}{\Gamma(\alpha)\Gamma(\gamma)} \frac{\tau(u/s)^{\tau\gamma}}{u[1+(u/s)^\tau]^{\alpha+\gamma}}, \quad E(D_i) = \exp(\eta_i)$$

again, with  $s_i$  given as function of  $\eta_i, \alpha, \tau, \gamma$ .

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### Claim Delay (D) Distn (cont)



Fit the 4 models under a Bayesian framework using *MCMC*:

- Compare fit using *Deviance Information Criterion*:

	LN	Tr. Gamma	Burr	Gen. Beta
DIC	194,356	193,025	191,262	190,992

- Model fit also compared using *posterior predictive checking*.

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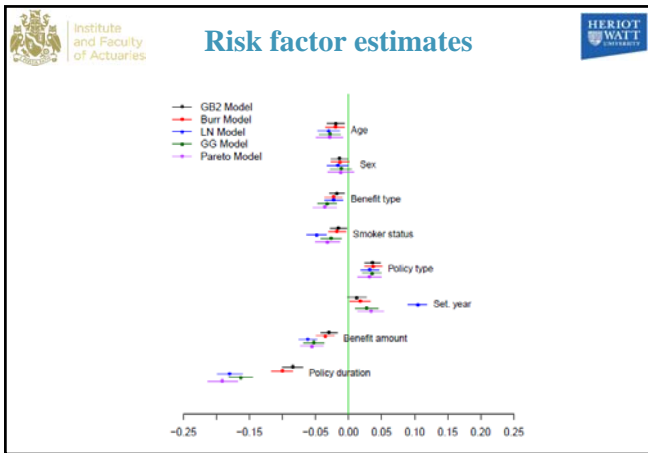
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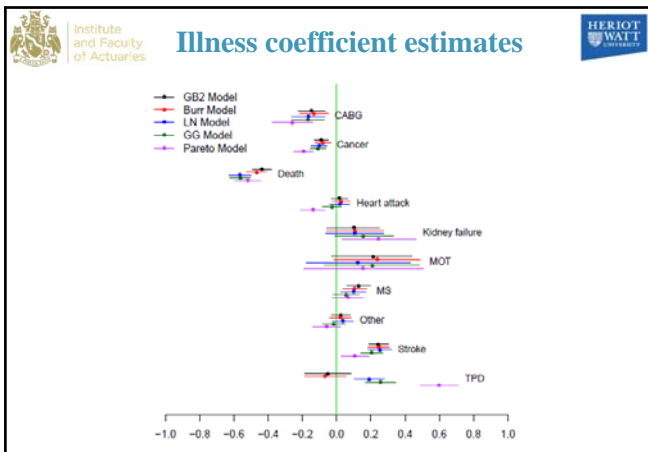
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**Prediction of Claim Delay**

Find 'best' predictive model using Bayesian variable selection – Ozkok et al. (2012a))

Covariate	Scenario				
	1	2	3	4	5
Benefit type	FA	FA	FA	FA	FA
Policy type	JL	JL	JL	JL	JL
Amount (£)	50k	250k	50k	50k	50k
Duration (yrs)	4	4	1	4	4
Office	11	11	11	11	11
Cause	Cancer	Cancer	Cancer	Death	TPD
$E[D]$ (days)	174	156	195	112	217
CI	(167, 182)	(146, 166)	(187, 204)	(106, 119)	(193, 243)

Scenario 1: reference; Red: changes from reference

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**Estimation of CI diagnosis rates**

- Non-recorded diagnosis dates estimated through CDD model
- Suppose Office 1 contributes data for 2000 to 2003. For this office, let:
  - $\theta$  be a set of covariates, including office
  - $\lambda_{x;\theta}^{(j)}$  be the diagnosis inception rate for cause  $j$  at age  $x$  with covariates  $\theta$
  - $E(u : x; \theta)$  be the number of policies (age  $x$ , covariates  $\theta$ ) inforce at time  $u$ ,  $0 \leq u \leq 4$
  - $N^{(j)}(x; \theta)$  be the number of claims (cause  $j$ , age  $x$ , covariates  $\theta$ ) diagnosed and settled in 2000 – 2003

$$N^{(j)}(x; \theta) \sim \text{Poisson} \left( \lambda_{x;\theta}^{(j)} \int_{u=0}^4 E(u : x; \theta) F^{(j)}(4 - u : x; \theta) du \right)$$


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**Model**

For all causes:

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    graph LR
      Healthy1[Healthy] -- λx;θ --> InsuredEvent[Insured event]
    
```

Have also considered:

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    graph TD
      Healthy2[Healthy] -- λx;θCI --> Diagnosed[Diagnosed with a CI]
      Healthy2 -- λx;θD --> Dead[Dead]
    
```

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**Rate smoothing (graduation)**

Gompertz-Makeham-Cox-type Model:

$$\lambda_{x;\theta}^{(j)} = \lambda_1^{(j)}(x) + \exp(\lambda_2^{(j)}(x)) \exp(\beta \mathbf{z}^T)$$

where  $\lambda_i^{(j)}(x)$  is a polynomial function of age only,  $i = 1, 2$

$\lambda_1^{(j)}(x) \equiv 0$  for each cause except death

$\lambda_1^{(j)}(x) \equiv 0 \rightarrow$  log-linear (Cox-type) model for  $\lambda_{x;\theta}^{(j)}$

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**Sensitivity to diagnosis estimates**

Relative rates (divided by rate obtained with Median of CDD)

- Rate using 97.5th percentile of Burr or LN CDD
- CIs derived using parametric bootstrap (Burr v LN)

Non-smokers, Pol durn 0, All causes

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**Premium pricing - derivation**

Annual premium, paid at constant rate,  $n$ -year term

- Can be calculated using

$$\text{Net Premium} = \text{Benefit Amount} \times \frac{\int_{t=0}^n v^t {}_t p_x \lambda_{x+t} dt}{\int_{t=0}^n v^t {}_t p_x dt}$$

where  ${}_t p_x$ : survival probability  
 $\lambda_{x+t}$ : total claim rate at age  $x + t$   
 $v$ : discount factor

- Then bootstrap distribution of  $\lambda$ s used to derive CIs for premiums

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**Premium pricing - comparisons**

Age 40, Pol Durn 0, All causes, Benefit amount £100k,  $i = 3\%$   
 (LN v Burr CDD)

Term	LN CDD		Burr CDD	
	Net premium rate	95% CI	Net premium rate	95% CI
Non-smokers				
5-years	156.05	148.05, 162.73	158.84	150.67, 165.00
25-years	373.69	352.67, 393.23	381.71	360.49, 400.63
Smokers				
5-years	239.70	225.96, 250.89	244.61	230.64, 256.17
25-years	714.03	662.88, 756.37	731.39	680.05, 779.60

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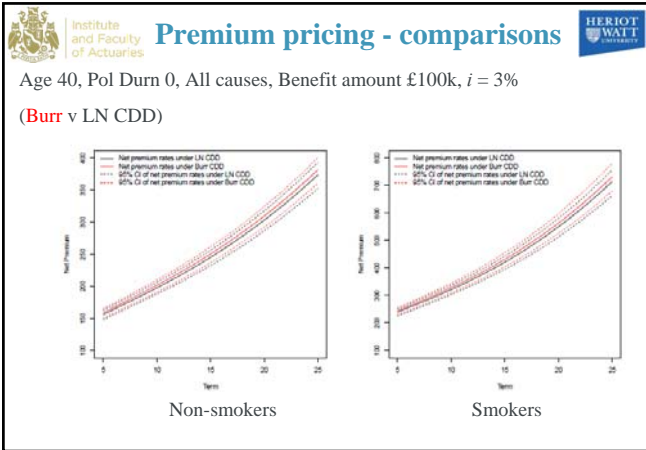
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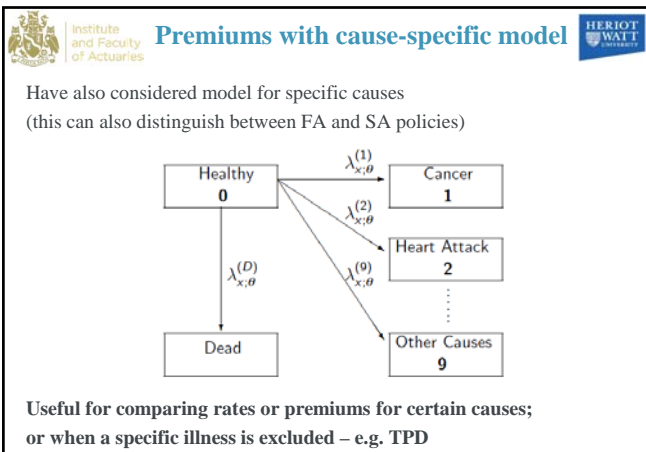
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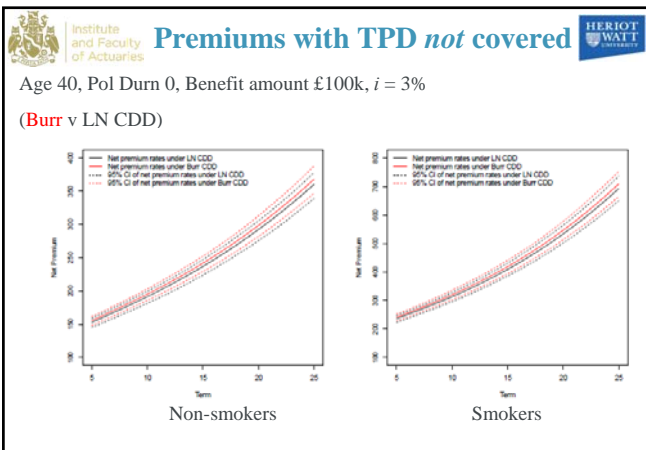
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## Summary



- Delay between diagnosis and settlement in CII is important  
(e.g. IBNR, IBNS)
- Have developed **delay model: depends on risk factors**
- Bayesian analysis **accounts for non-recorded diagnosis dates**
- 4-parameter G.Beta distn fits data best – followed by 3-parameter Burr
- **CII rates and premiums estimated & smoothed**  
– including parameter and model uncertainty
- Estimates of **CDD are model-sensitive**
- **But claim rates and premiums are not**

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## More details in:



Ozkok, E., Streftaris, G., Waters, H.R., and Wilkie, A.D. (2012a)  
Bayesian modelling of the time delay between diagnosis and settlement for Critical Illness Insurance using a Burr generalised-linear-type model. *Insurance: Mathematics & Economics*, 50, 266–279.

Ozkok, E., Streftaris, G., Waters, H.R., and Wilkie, A.D. (2012b)  
Modelling critical illness claim diagnosis rates I: Methodology. *Scandinavian Actuarial Journal*, doi: 10.1080/03461238.2012.728537

Ozkok, E., Streftaris, G., Waters, H.R., and Wilkie, A.D. (2013) Modelling critical illness claim diagnosis rates II: Results. *Scandinavian Actuarial Journal*, DOI:10.1080/03461238.2012.728538.

Dodd, E., Streftaris, G., Waters, H.R. and Stott, A.D. (2014) The effect of model uncertainty on the pricing of critical illness insurance. To appear in *Annals of Actuarial Science*.

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