

# Stochastic Mortality Modeling with Lévy processes

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2010/9/10

# ABSTRACT

- In the classical Lee-Carter model, the mortality indices  $kt$  assumed to be a random walk model with drift is normally distributed.
- The assumption of normality is stacked against the hard fact.
- So, we adopt two infinitely divisible distributions—**Generalized Hyperbolic (GH)** and **Classical Tempered Stable (CTS) distributions**—to model the mortality indices.
- We show that GH or CTS distribution assumptions are better than normal distribution assumption.

# Agenda

- Introduction
- The Lee-Carter Model with GH and CTS Innovations
  - The Lee-Carter model
  - Normality Test for Mortality Indices
  - Modeling Mortality indices as GH and CTS Processes
- Empirical Analysis
  - Mortality Data
  - Model Comparison
- Conclusions

# Introduction

- Lee and Carter (1992)...**popular**.
- Brouhns, Denuit, and Vermunt (2002), Renshaw and Haberman (2003), Denuit, Devolder, and Goderniaux (2007), and Li and Chan (2007)...**various modifications of the Lee-Carter model**.
- Cairns, Blake, and Dowd (2006), and Cairns et al. (2009)...**CBD model and its extension**.
- Haberman and Renshaw (2009)... **forecast age-period-cohort mortality and test for robustness**.
- The above papers *don't* include mortality jumps.

# Introduction

- Biffis (2005)... model asset prices and mortality dynamics.
- Luciano and Vigna (2005)... for Italian mortality data...add a jump component and fit better.
- Cox, Lin and Wang (2006)... for US and UK ...geometric Brownian motion and compound Poisson ...Swiss Re bond.
- Lin and Cox (2008)... Brownian motion and a discrete Markov chain with log-normal jump size distribution... incomplete market framework.
- Chen and Cox (2009)... a jump process into the Lee-Carter model.
- The above papers use diffusion processes with jump components, *finite activity* Lévy processes, to describe the dynamics of mortality rates.

# Introduction

- A pure jump Lévy process can display either finite activity or infinite activity.
- The classical example of a **finite-activity** jump process is the compound Poisson jump process.
- **Variance Gamma process** ... infinite activity ... finite variation ... a special case of **CTS** process.
- **Normal Inverse Gaussian process** ... infinite activity ... infinite variation ... a special case of **GH** process.

# Introduction

- Hainaut and Devolder (2008) first applies  $\alpha$ -stable subordinators—infinite-activity strictly positive Lévy processes—to model the **mortality hazard rates**.
- However, in the Lee-Carter model, the first difference of mortality indices may be negative to reflect the mortality improvement.
- Unlike Hainaut and Devolder (2008), in this paper we incorporate the GH and CTS processes into the original Lee–Carter model, and use it to fit and forecast the mortality rates.

# Introduction

- We select the ninety-five-year mortality data of ten countries—**Denmark, Finland, France, Iceland, Italy, Netherlands, Norway, Spain, Sweden and Switzerland**—as the observed data.
- We **fit** the model to the mortality rates from **1912 to 2001** and **forecast** the development of the mortality curve for the consequent **five years**.

# The Lee-Carter Model with GH and CTS Innovations

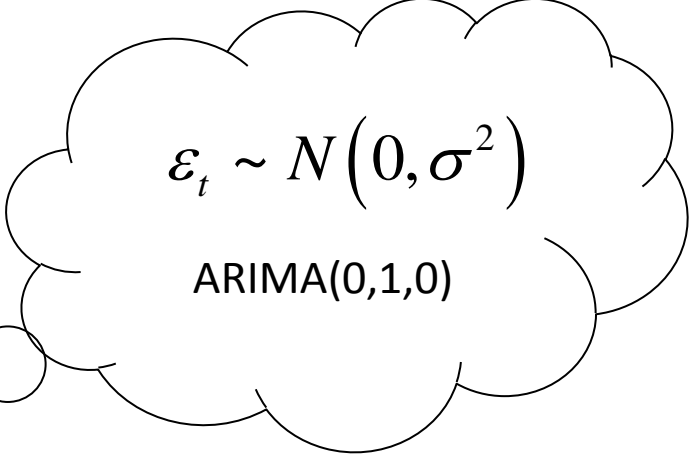
- The Lee-Carter model
- Normality Test for Mortality Indices
- Modeling Mortality indices as GH and CTS Processes

# The Lee-Carter model

- Lee and Carter (1992)

$$\ln(m_{x,t}) = \alpha_x + \beta_x k_t + e_{x,t}$$


$$\sum_t k_t = 0 \quad \text{and} \quad \sum_x \beta_x = 1$$

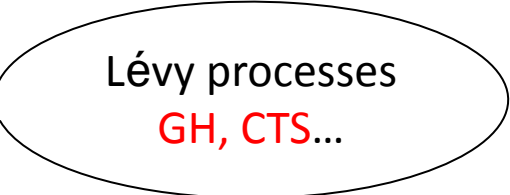

$$\varepsilon_t \sim N(0, \sigma^2)$$

ARIMA(0,1,0)

$$k_t - k_{t-1} = \gamma + \varepsilon_t$$

Our key point


$$k_t^L - k_{t-1}^L = L_1$$



Lévy processes  
GH, CTS...

# Normality Test for Mortality Indices

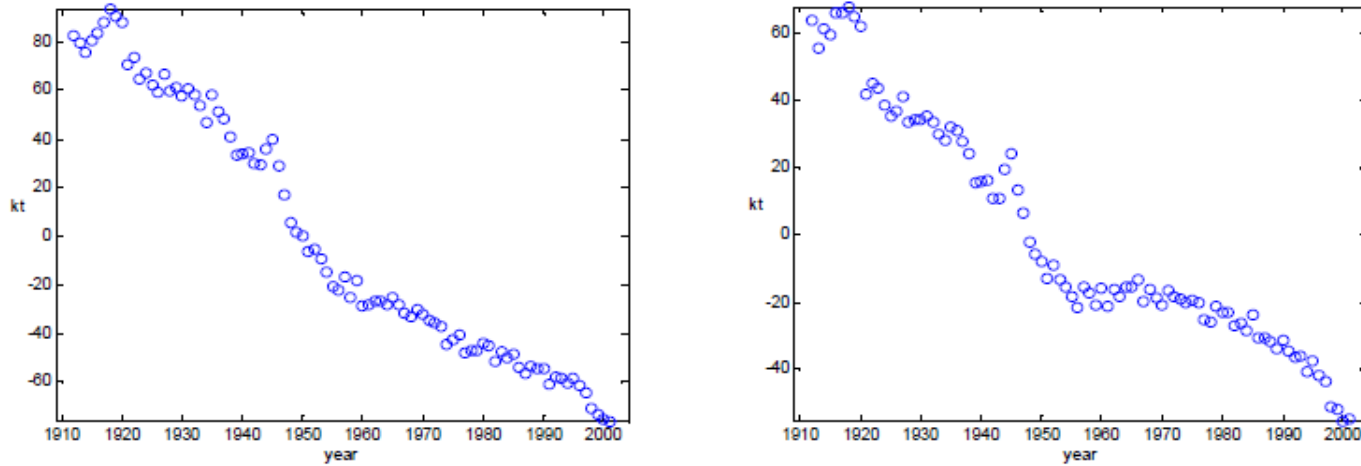


Figure 1A. Demark Female (Left) and Male (Right).

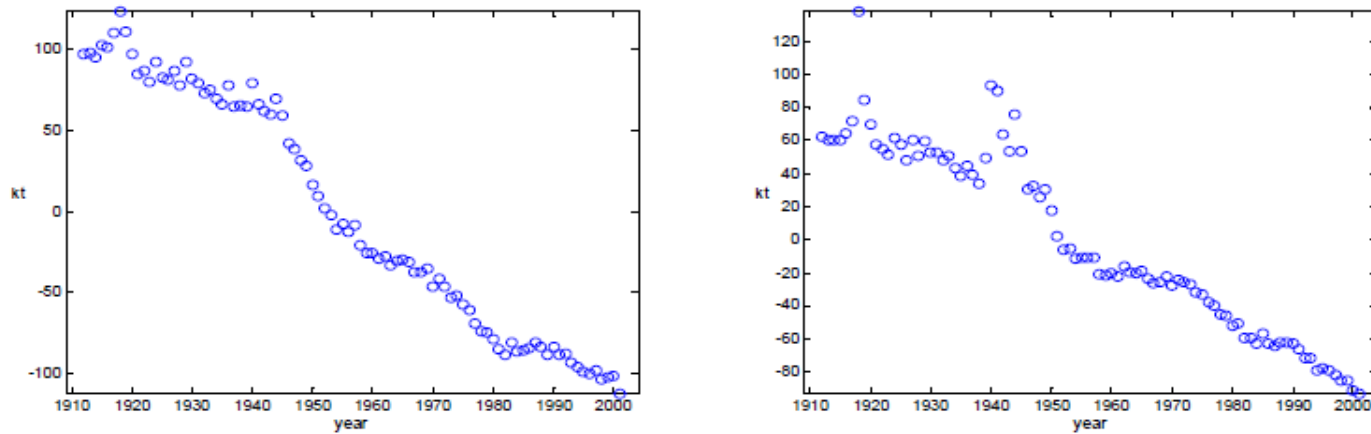


Figure 1B. Finland Female (Left) and Male (Right).

# Normality Test for Mortality Indices

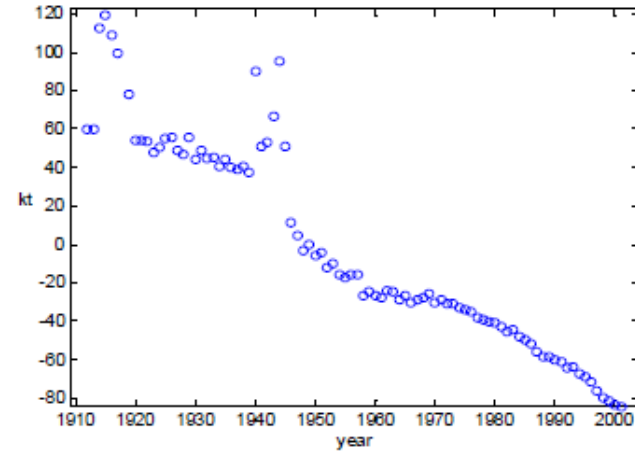
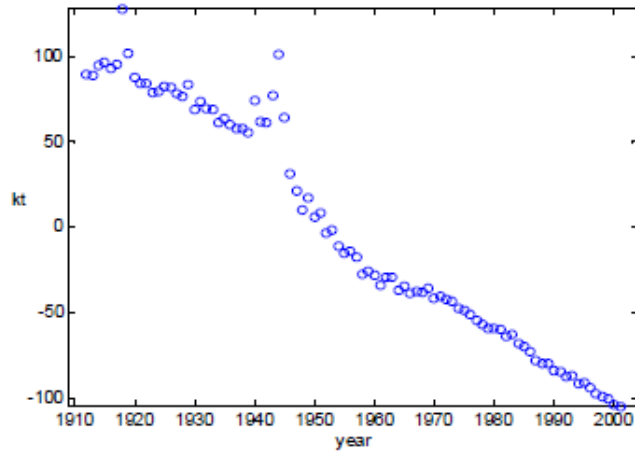


Figure 1C. France Female (Left) and Male (Right).

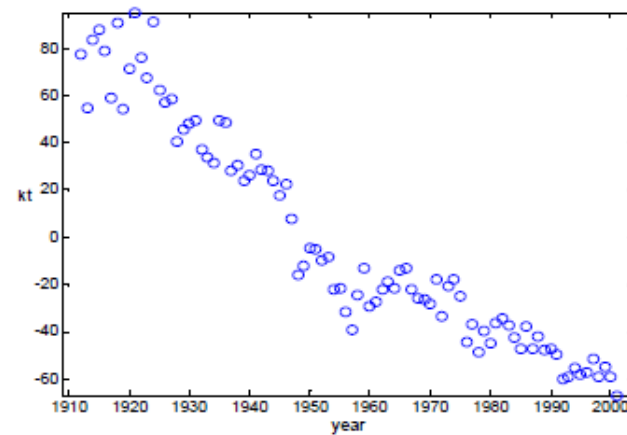
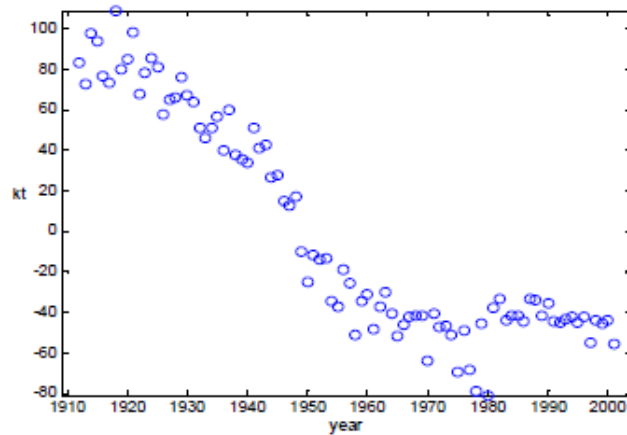


Figure 1D. Iceland Female (Left) and Male (Right).

# Normality Test for Mortality Indices

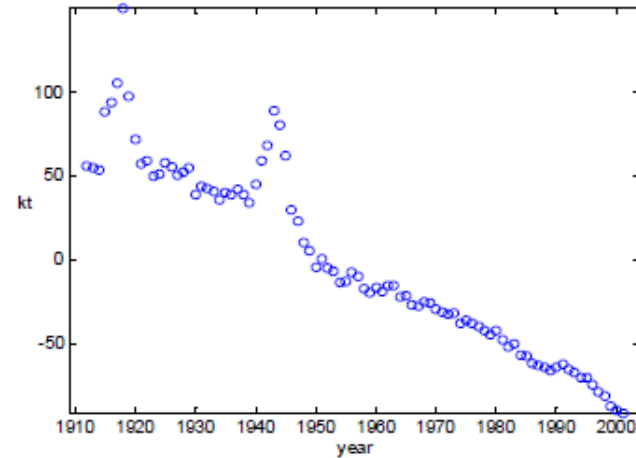
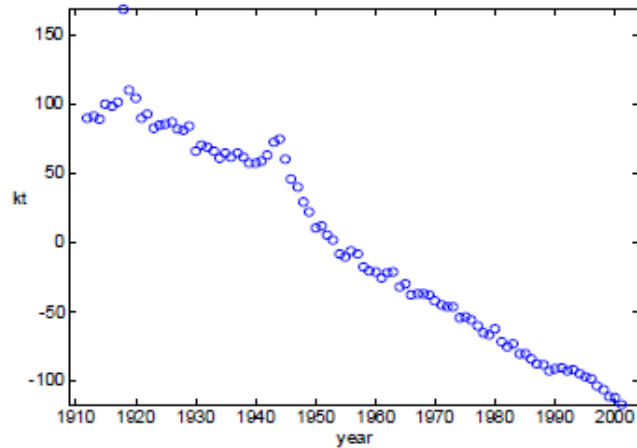


Figure 1E. Italy Female (Left) and Male (Right).

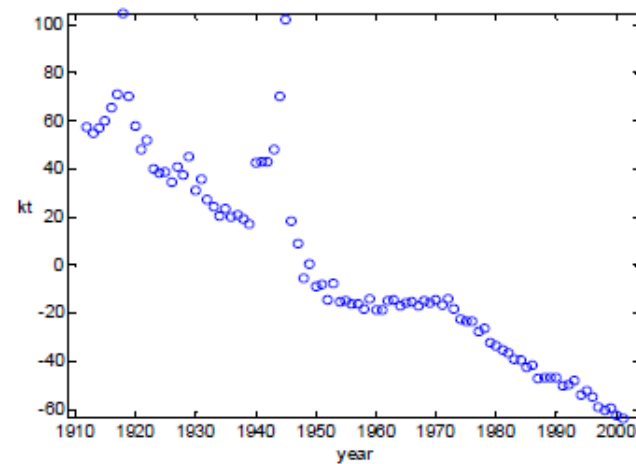
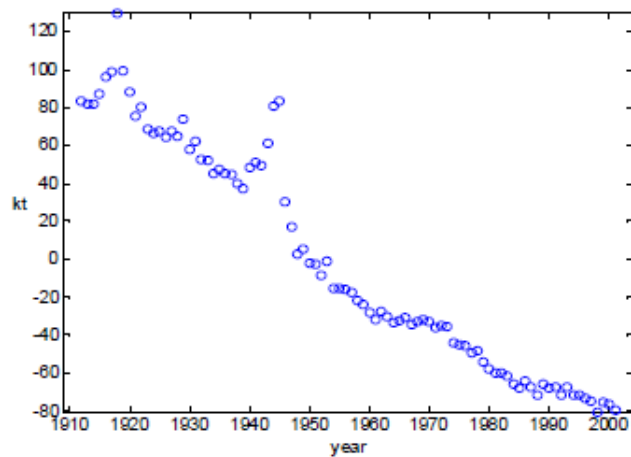


Figure 1F. Netherlands Female (Left) and Male (Right).

# Normality Test for Mortality Indices

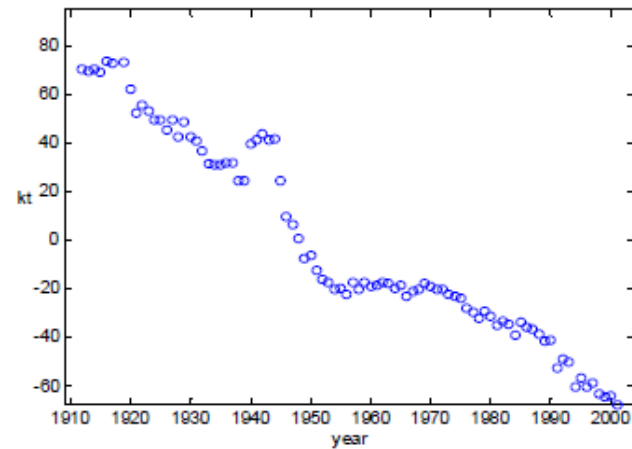
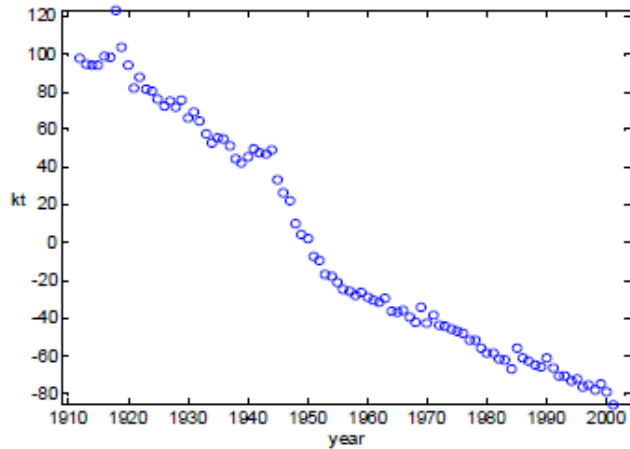


Figure 1G. Female (Left) and Male (Right).

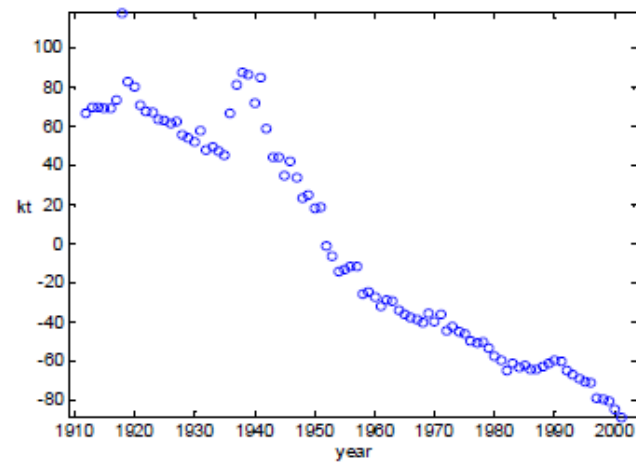
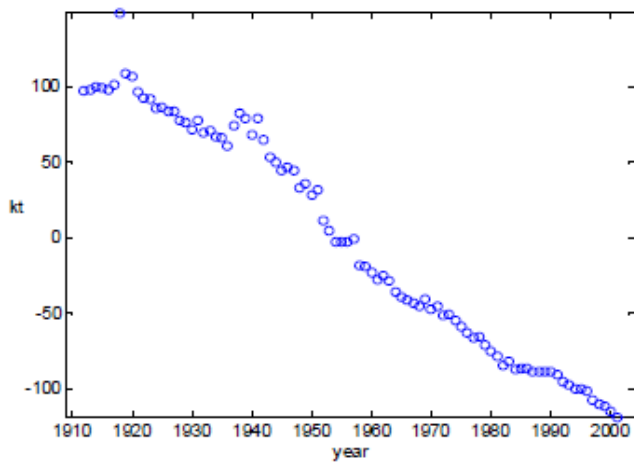


Figure 1H. Spain Female (Left) and Male (Right).

# Normality Test for Mortality Indices

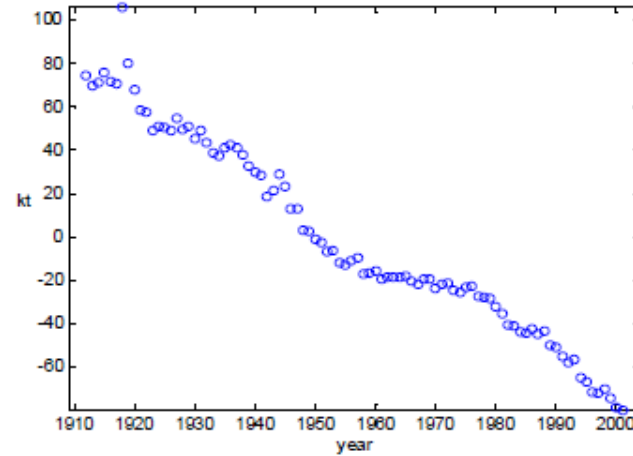
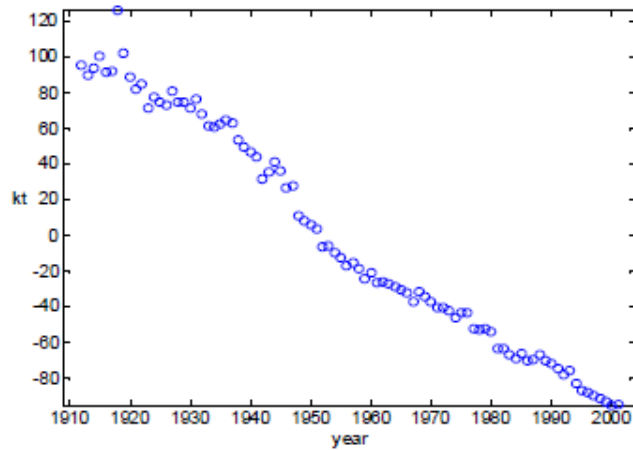


Figure 1I. Sweden Female (Left) and Male (Right).

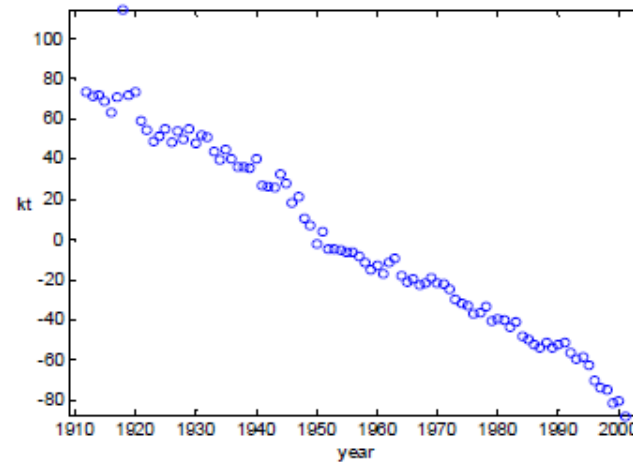
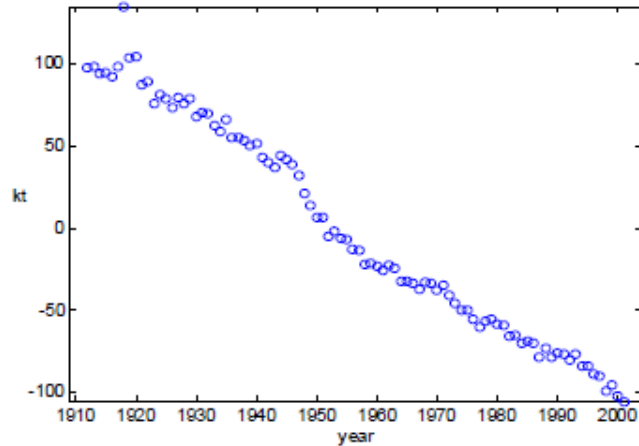


Figure 1J. Switzerland Female (Left) and Male (Right).

# Normality Test for Mortality Indices

$H_0$  : The data are from a normal distribution.

- The JB test statistic (Jarque and Bera, 1980), a goodness-of-fit measure of departure from normality, as follows:

$$JB = n \left[ \frac{s^2}{6} + \frac{(k-3)^2}{24} \right]$$

where  $n$  is the sample size;  $s$  is the sample skewness;  $k$  and is the sample kurtosis.

# Normality Test for Mortality Indices

Country	Gender	P-value	Skewness	Excess Kurtosis
Denmark	Female	0.248	-0.302	0.426
	Male	<0.001	-0.738	2.783
Finland	Female	0.284	0.352	0.063
	Male	<0.001	1.476	13.001
France	Female	<0.001	-0.219	6.612
	Male	<0.001	0.462	8.203
Iceland	Female	0.083	0.327	0.776
	Male	0.063	0.025	1.095
Italy	Female	<0.001	1.418	27.125
	Male	<0.001	-0.120	9.555
Netherlands	Female	<0.001	-1.670	12.193
	Male	<0.001	-3.037	23.809
Norway	Female	<0.001	0.904	6.299
	Male	<0.001	0.126	4.906
Spain	Female	<0.001	1.312	17.559
	Male	<0.001	0.962	10.622
Sweden	Female	<0.001	1.354	11.542
	Male	<0.001	1.815	17.930
Switzerland	Female	<0.001	1.049	12.520
	Male	<0.001	0.615	19.881

# Modeling Mortality indices as GH and CTS Processes

$$k_t - k_{t-1} = L_1$$

where  $L_1$  is the Lévy process with unit time scale. When  $L_1$  is the GH (CTS) process, the mortality index  $k_t$  follows GH (CTS) distribution.

# Modeling Mortality indices as GH and CTS Processes

Processes	Characteristic Function $\phi(\omega)$
Generalized Hyperbolic model $GH(\alpha, \beta, \delta, \lambda, \mu)$	$e^{i\mu u} \left( \frac{\alpha^2 - \beta^2}{\alpha^2 - (\beta + iu)^2} \right)^{\lambda/2} \frac{K_\lambda \left( \delta \sqrt{\lambda^2 - (\beta + iu)^2} \right)}{K_\lambda \left( \delta \sqrt{\alpha^2 - \beta^2} \right)}$
Classical Tempered Stable $CTS(\alpha, C_+, C_-, \lambda_+, \lambda_-, m)$	$\exp \left\{ ium + \Gamma(1-\alpha) (C_+ \lambda_+^{\alpha-1} - C_- \lambda_-^{\alpha-1}) \right. \\ \left. + C_+ \Gamma(-\alpha) \left( (\lambda_+ - iu)^\alpha - \lambda_+^\alpha \right) \right. \\ \left. + C_- \Gamma(-\alpha) \left( (\lambda_- + iu)^\alpha - \lambda_-^\alpha \right) \right\}$

Note:  $K$  is the modified Bessel function of the second kind, and  $\Gamma$  is a gamma function.

# Modeling Mortality indices as GH and CTS Processes

- The GH distribution includes the NIG distribution, the hyperbolic distribution, the **normal distribution**, Student  $t$  distribution, the skew hyperbolic- $t$  distribution, and the variance gamma distribution.
- The CTS distribution includes the truncated Lévy flight distribution, the CGMY distribution, the Variance Gamma distribution, the **normal distribution**.
- Therefore, we provides a more **flexible** and **abundant** mortality index models with GH and CTS innovations than that with normal innovation.

# Empirical Analysis

- Mortality Data
- Model Comparison

# Mortality Data

- Human Mortality Database (HMD)
- Ten countries: Denmark, Finland, France, Iceland, Italy, Netherlands, Norway, Spain, Sweden and Switzerland.
- Age: 0 to 100.
- Fitting period: 1912 to 2001.
- Forecasting period: 2002 to 2006.

# Model Comparison

- Log likelihood function (LLF)
- Akaike Information Criterion (AIC)  
$$\text{AIC} = (-2 * \text{LLF}) + (2 * \text{NPS})$$
where NPS is the effective number of parameters being estimated.
- Bayesian information criteria (BIC)  
$$\text{BIC} = (-2 * \text{LLF}) + (\text{NPS} * \log(\text{NOS}))$$
where NOS is the number of observations.
- Kolmogorov–Smirnov test  
 $H_0$  : The sample is drawn from the reference distribution.

# Good-of-Fit Measures for Female

Female		LLF	AIC	BIC	KS p-value
Denmark	Normal	-281.6312	567.2624	572.2620	0.0143
	GH	-266.0407	542.0813	554.5804	<b>0.9954</b>
	CTS	<b>-266.0381</b>	<b>542.0762</b>	<b>554.5752</b>	0.9942
Finland	Normal	-317.3149	638.6298	643.6294	0.0175
	GH	<b>-292.8436</b>	<b>595.6872</b>	<b>608.1862</b>	<b>0.6544</b>
	CTS	-292.8464	595.6929	608.1919	0.6493
France	Normal	-311.0207	626.0413	631.0410	0.0027
	GH	<b>-289.2394</b>	<b>588.4787</b>	<b>600.9778</b>	<b>0.9225</b>
	CTS	-291.3043	592.6087	605.1077	0.6013
Iceland	Normal	<b>-323.9160</b>	<b>651.8320</b>	<b>656.8316</b>	0.5365
	GH	-362.3137	734.6274	747.1264	<b>0.9489</b>
	CTS	-362.3246	734.6492	747.1483	0.9385
Italy	Normal	-319.6590	643.3180	648.3176	0.0002
	GH	<b>-284.8698</b>	<b>579.7396</b>	<b>592.2387</b>	<b>0.8639</b>
	CTS	-287.3213	584.6427	597.1417	0.3910
Netherlands	Normal	-314.2988	632.5976	637.5973	0.0013
	GH	<b>-293.1537</b>	<b>596.3074</b>	<b>608.8064</b>	<b>0.9261</b>
	CTS	-293.5052	597.0104	609.5095	0.7154
Norway	Normal	-272.4171	548.8343	553.8339	0.1826
	GH	<b>-265.6993</b>	<b>541.3985</b>	<b>553.8976</b>	<b>0.9990</b>
	CTS	-265.7121	541.4241	553.9232	0.9984
Spain	Normal	-307.0227	618.0454	623.0451	0.0058
	GH	<b>-278.6876</b>	<b>567.3751</b>	<b>579.8742</b>	0.8670
	CTS	-279.8252	569.6504	582.1494	<b>0.9079</b>
Sweden	Normal	-300.1478	604.2955	609.2951	0.0830
	GH	<b>-271.8271</b>	<b>553.6542</b>	<b>566.1533</b>	0.8476
	CTS	-273.0809	556.1618	568.6609	<b>0.9467</b>
Switzerland	Normal	-301.6165	607.2331	612.2327	0.1811
	GH	<b>-279.7489</b>	<b>569.4978</b>	<b>581.9968</b>	<b>0.9985</b>
	CTS	-279.8568	569.7137	582.2127	0.9929

# Good-of-Fit Measures for Male

Male		LLF	AIC	BIC	KS p-value
Denmark	Normal	-259.6505	523.3010	528.3006	0.0425
	GH	-253.6852	517.3704	529.8695	0.7456
	CTS	<b>-253.1297</b>	<b>516.2595</b>	<b>528.7585</b>	<b>0.7683</b>
Finland	Normal	-322.9167	649.8334	654.8330	0.0012
	GH	<b>-312.1082</b>	<b>634.2164</b>	<b>646.7155</b>	<b>0.9880</b>
	CTS	-313.7963	637.5927	650.0917	0.6842
France	Normal	-323.1906	650.3813	655.3809	0.0000
	GH	-297.9508	605.9016	618.4006	<b>0.9966</b>
	CTS	<b>-291.7267</b>	<b>593.4535</b>	<b>605.9525</b>	0.2486
Iceland	Normal	-323.3071	650.6143	655.6139	0.5617
	GH	<b>-342.3913</b>	<b>694.7827</b>	<b>707.2817</b>	<b>0.9946</b>
	CTS	-342.4289	694.8578	707.3568	0.9931
Italy	Normal	-320.0279	644.0558	649.0554	0.0022
	GH	-301.0792	612.1584	624.6575	0.8997
	CTS	<b>-286.8498</b>	<b>583.6996</b>	<b>596.1986</b>	<b>0.9018</b>
Netherlands	Normal	-322.3932	648.7863	653.7859	0.0002
	GH	<b>-293.6083</b>	<b>597.2166</b>	<b>609.7157</b>	<b>0.8486</b>
	CTS	-295.6467	601.2934	613.7925	0.8130
Norway	Normal	-270.8644	545.7288	550.7284	0.0022
	GH	<b>-261.6509</b>	<b>533.3017</b>	<b>545.8008</b>	<b>0.9998</b>
	CTS	-262.0067	534.0134	546.5124	0.9880
Spain	Normal	-309.9867	623.9735	628.9731	0.0039
	GH	<b>-286.4151</b>	<b>582.8302</b>	<b>595.3292</b>	<b>0.9921</b>
	CTS	-286.9939	583.9879	596.4869	0.8771
Sweden	Normal	-275.6614	555.3228	560.3224	0.0166
	GH	<b>-257.7139</b>	<b>525.4279</b>	<b>537.9269</b>	<b>0.8817</b>
	CTS	-258.2076	526.4151	538.9141	0.6671
Switzerland	Normal	-303.0029	610.0058	615.0054	0.0531
	GH	<b>-275.4430</b>	<b>560.8861</b>	<b>573.3851</b>	<b>0.9761</b>
	CTS	-276.5020	563.0040	575.5030	0.8912

# Model Comparison

- The definition of the Mean Absolute Percentage Error (MAPE) is calculated as follows:

$$MAPE = 100\% \times \frac{1}{n} \sum_{i=1}^n \left| \frac{A_i - F_i}{A_i} \right|$$

where  $A_i$  is historical mortality rate and  $F_i$  is the forecast mortality rate.

- Forecasting period: 2002 to 2006 ... **five years**.
- Simulate 100,000 paths.

## Percentile of MAPE of Mortality Projection for Female (Unit: %)

Female		5%	10%	Medium	90%	95%
Denmark	Normal	<b>3.4661</b>	<b>3.4848</b>	<b>3.5996</b>	<b>3.9596</b>	<b>4.0840</b>
	GH	3.4821	3.5034	3.7137	4.4009	4.6574
	CTS	3.4778	3.5005	3.7161	4.2933	4.5534
Finland	Normal	7.1603	7.2002	7.6322	8.5164	8.8146
	GH	<b>7.1345</b>	<b>7.1602</b>	<b>7.4194</b>	<b>8.0261</b>	<b>8.2470</b>
	CTS	7.1348	7.1723	7.4431	8.0841	8.3508
France	Normal	3.5614	3.5894	3.8911	5.0650	5.4992
	GH	<b>3.5217</b>	<b>3.5361</b>	<b>3.7007</b>	<b>4.8471</b>	<b>5.4722</b>
	CTS	3.5319	3.5536	3.7987	5.3770	6.3538
Iceland	Normal	10.5264	10.5812	11.0350	12.5004	13.0243
	GH	10.5170	<b>10.5588</b>	11.0394	12.6986	13.5475
	CTS	<b>10.5136</b>	10.5598	<b>10.9045</b>	<b>12.4920</b>	<b>13.2356</b>
Italy	Normal	4.1625	4.1903	4.5423	6.0183	6.6042
	GH	<b>4.1252</b>	<b>4.1372</b>	<b>4.2982</b>	<b>5.3761</b>	<b>5.9318</b>
	CTS	4.1316	4.1423	4.3559	5.8481	6.7074
Netherlands	Normal	2.9293	2.9699	3.3566	4.6307	<b>5.0872</b>
	GH	<b>2.8961</b>	<b>2.9166</b>	<b>3.1286</b>	<b>4.5533</b>	5.7357
	CTS	2.9043	2.9279	3.2173	4.7998	5.8568
Norway	Normal	<b>5.4941</b>	<b>5.5092</b>	5.5938	<b>5.8619</b>	<b>5.9884</b>
	GH	5.4961	5.5112	5.5928	5.8806	6.0660
	CTS	5.5019	5.5119	<b>5.5906</b>	5.8833	6.0096
Spain	Normal	4.2810	4.3192	4.5913	5.4238	5.8462
	GH	<b>4.2544</b>	<b>4.2728</b>	<b>4.4257</b>	<b>5.1299</b>	<b>5.7671</b>
	CTS	4.2554	4.2823	4.4859	5.4550	6.0277
Sweden	Normal	3.8128	3.8275	3.9756	<b>4.4927</b>	<b>4.6687</b>
	GH	<b>3.7951</b>	<b>3.8049</b>	3.9123	4.5567	4.9596
	CTS	3.7963	3.8056	<b>3.9106</b>	4.5457	4.8950
Switzerland	Normal	4.0601	4.0793	4.2413	4.8233	<b>5.0412</b>
	GH	<b>4.0472</b>	<b>4.0574</b>	4.1814	<b>4.7487</b>	5.1421
	CTS	4.0487	4.0617	<b>4.1806</b>	4.7575	5.2766

## Percentile of MAPE of Mortality Projection for Male (Unit: %)

<b>Male</b>		5%	10%	Medium	90%	95%
Denmark	Normal	4.8640	4.8773	5.0096	<b>5.3456</b>	<b>5.4648</b>
	GH	4.8661	4.8789	4.9964	5.3920	5.5433
	CTS	<b>4.8629</b>	<b>4.8761</b>	<b>4.9925</b>	5.4102	5.6029
Finland	Normal	8.0899	8.1501	8.6171	10.3698	11.1702
	GH	8.0403	<b>8.0758</b>	<b>8.3922</b>	<b>9.9201</b>	11.3050
	CTS	<b>8.0382</b>	8.0908	8.4744	9.9820	<b>11.0628</b>
France	Normal	5.0559	5.1106	5.7243	7.9078	8.7535
	GH	<b>4.9889</b>	<b>5.0164</b>	<b>5.3147</b>	<b>6.8770</b>	<b>8.0444</b>
	CTS	4.9996	5.0433	5.5282	8.2538	9.3840
Iceland	Normal	11.2200	11.2890	11.6577	<b>12.8096</b>	<b>13.4348</b>
	GH	<b>11.2156</b>	11.2708	<b>11.6262</b>	12.8255	13.5292
	CTS	11.2196	<b>11.2807</b>	11.6611	12.9423	13.6455
Italy	Normal	5.2765	5.3187	5.7539	7.2911	<b>7.8877</b>
	GH	<b>5.2503</b>	<b>5.2719</b>	<b>5.5599</b>	7.4513	8.4238
	CTS	5.2513	5.2732	5.6115	<b>7.1392</b>	8.0599
Netherlands	Normal	4.0499	4.1410	4.8670	6.7219	7.3716
	GH	<b>3.9520</b>	<b>3.9871</b>	<b>4.3069</b>	<b>6.0929</b>	<b>7.3205</b>
	CTS	3.9620	4.0003	4.4576	6.8600	8.5175
Norway	Normal	6.4322	6.4506	6.5525	6.7766	<b>6.8710</b>
	GH	<b>6.4361</b>	<b>6.4458</b>	<b>6.5123</b>	<b>6.7655</b>	7.0372
	CTS	6.4327	6.4460	6.5198	6.8094	7.0229
Spain	Normal	3.7542	3.7951	4.1770	5.3553	<b>5.7795</b>
	GH	<b>3.7197</b>	<b>3.7376</b>	<b>3.9743</b>	<b>5.3256</b>	6.4212
	CTS	3.7199	3.7417	3.9897	5.4643	6.1843
Sweden	Normal	5.1920	5.2041	5.2909	5.6212	5.7940
	GH	<b>5.1787</b>	<b>5.1858</b>	<b>5.2427</b>	<b>5.5158</b>	<b>5.7149</b>
	CTS	5.1810	5.1889	5.2590	5.6478	5.9990
Switzerland	Normal	5.4998	5.5220	5.7623	6.5570	6.8685
	GH	<b>5.4769</b>	<b>5.4857</b>	<b>5.5864</b>	<b>6.1664</b>	<b>6.5336</b>
	CTS	5.4783	5.4886	5.6292	6.4505	7.2256

# Conclusions

- We make the first attempt to incorporate two infinity-activity pure jump Lévy processes — **Generalized Hyperbolic (GH)** and **Classical Tempered Stable (CTS) Processes** — into Lee-Carter model respectively.
- The results **consistently** indicate a preference for the mortality indices with GH or CTS distributions over those with normal distribution.
- In this paper, we do not provide the valuation and hedging of mortality securities. Using the Esscher transform, future research may focus on how to price and hedge the mortality-linked security such as Swiss Re mortality bond issued in 2003.

Thank you for your attention.