



Generating Mortality Rate Intensity for Life Insurance Applications through Novel Method of Successive Differencing Under the Parsimonious Generalised Makeham's Framework

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Abstract

Developing and implementing age dependent mortality rates in functional forms have presented critical modelling problems for life insurance and annuity firms; thereby creating serious research gaps. A recurring problem with the existing life tables is that their underlying generating functions are not capable of showing any evidence that mortality at 10 declines. This observed irregularity in the age pattern of mortality is the prime motivation in the search for analytical functions which fully capture the observed variations of mortality with age. To fill the gaps identified, the study aims to develop specific life table function under the Generalised Makeham's framework. The objectives of this study are to (i) compute the mortality rate intensities μ_x (ii) compute the curve of deaths densities $l_x\mu_x$ (iii) compute the probability of deaths q_x and then compare the common domain of definition of these measures. From the method of successive differencing employed to model $GM(2,2)$, the male's ageing parameter value 1.102923606 falls within the globally accepted interval $1.08 \leq C \leq 1.12$ for the $GM(2,2)$. This method is superior to the method of maximum likelihood estimation which mostly violates the permissible interval of validity. Computational evidence from our analysis proves that under the Generalised Makeham's law, the mortality rate intensity declines at 10.

Keywords: Generalised Makeham, curve of deaths, permissible interval, successive differencing

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Introduction

Life insurance which payments are either contingent on death or survival concerns every individual especially the working population who are automatically covered while in service by reason of the government's social security program but who will possibly choose life annuity options at retirement. In actuarial context, it is essential to observe that life tables are used to measure the pattern of deaths, survival rates, life expectancy at varying ages in order to explain changes in the population.

In order to price life insurance and annuity products and ensure the solvency of insurance companies, estimates of future events such as deaths are functionally formulated using mortality models to generate life tables. For life offices, any potential deviation from mortality assumption employed in pricing at the inception of the contract will constitute clear threats to its underwriting performance. It has been observed that many life insurance providers in developing countries such as Nigeria depend on exotic life tables which are incorrectly applied to overcharge or undercharge their policyholders. If they undercharge, many claims are reported and life insurance products are underpriced leading to the risk of insurer's insolvency or if they overcharge, workers will suffer inadequate income streams at retirement.

Developing and implementing age dependent mortality rates in functional forms have presented critical modelling problems for life insurance and annuity firms; thereby creating serious research gaps. A recurring problem with the existing life tables is that their underlying generating functions are not capable of measuring perinatal mortality and there is no evidence that mortality at 10 declines. These observed irregularities in the age pattern of mortality are the prime motivation in the search for analytical functions that fully capture the observed variations of mortality with age. To fill the gaps identified, the study aims to develop specific life table function under the frameworks of Generalised Makeham's force of mortality.

The basic data input required for generating life tables are the age specific death rates enumerated from the information on deaths from vital statistics (using age and sex) and population l_x by age and sex. In most developing countries such as Nigeria, these data do not exist because of dysfunctional vital registration system. The most popular law of mortality stated in Gompertz's law [1] was developed to model old age mortality around 60 - 90 years and assumes that death rates increase exponentially at old age.



Following observation by Debon *et al.* [2], the Generalized Makeham's mortality function in equation (1) is defined to measure mortality from infancy to old ages.

$$\mu_x = GM(m, n) = \sum_{k=1}^m \beta_k x^{k-1} + \exp \sum_{k=m+1}^{m+n} \beta_k x^{k-m-1} \quad (1)$$

where m and n are non-negative integers representing the orders of the polynomials and β_k are mortality

$$\mu_x = GM(0, n) = \sum_{k=1}^0 \beta_k x^{k-1} + \exp \sum_{k=1}^n \beta_k x^{k-1} = \beta_1 + \frac{\beta_0}{x} + \exp \sum_{k=1}^n \beta_k x^{k-1} \quad (2)$$

$$\mu_x = GM(m, 0) = \sum_{k=1}^m \beta_k x^{k-1} + \exp \sum_{k=m+1}^m \beta_k x^{k-m-1} = \sum_{k=1}^m \beta_k x^{k-1} + \exp \left((\beta_{m+1}) + \left(\frac{\beta_m}{x} \right) \right) \quad (3)$$

Consequently, mortality at zero μ_0 is inadmissible. Therefore, in order to absorb the Gompertz's assumption and fall in line with actuarial practice, the undefined terms $\frac{\beta_0}{x}$ and $\frac{\beta_m}{x}$ at age zero are ignored and the equations are modified as follows;

$$\mu_x = GM(m, n) = \begin{cases} \sum_{k=1}^m \beta_k x^{k-1} + \exp \sum_{k=m+1}^{m+n} \beta_k x^{k-m-1} & m \geq 1, n \geq 1 \\ \exp \sum_{k=1}^n \beta_k x^{k-1} & m = 0 \\ \sum_{k=1}^m \beta_k x^{k-1} & n = 0 \end{cases} \quad (4)$$

If $m = 0, n = 2$, we have the Gompertz's law $\mu_x = GM(0, 2) = BC^x$; $60 \leq x \leq 90$ where B is the initial mortality, C is the ageing parameter and x is the age. This mortality law is not capable of capturing infant mortality rates and mortality due to accidents in young adulthood (accidental hump). This is because exponential mortality growth rate is not expected before sexual maturity.

The relatively few parameters in [1] makes it inflexible consequently, the author in [3] assumes a constant additive age-independent constant parameter that accounts for the accidental hump mortality. If $m = 1, n = 2$ in equation(4) the Makeham's mortality function becomes

$$\mu_x = GM(1, 2) = A + BC^x; \quad 20 \leq x \leq 90 \quad (4a)$$

where A defines the background mortality independent of age [4]. The assumption of constant mortality rate is not always true since we all have different mortality exposures due to different life styles, nutrition, job and environment.

To capture mortality at low ages and extreme ages from this law, mortality rates are then extrapolated because of insufficient mortality experience. Since there is an approved range of validity where the mortality function is well behaved, best mortality estimates outside the intended range may not be obtained by extrapolation. In order to avoid the need to extrapolate, we can set $m = 2, n = 2$ in (4) to obtain the Generalized

parameters. Because of the complexity involved in evaluating the parameters as both orders increase, m and n are constrained in practice as $m + n < 5$. This equation has implications for m and n . When $m = 0$ and $n = 0$, we obtain the following equations;

$$\text{Makeham's law } \mu_x = GM(2, 2) = A + Hx + BC^x$$

where $H \in \mathbf{R}$ is a new parameter which explains changes in the background mortality. It is natural for mortality to decline at ten to explain changes in mortality between conception and death. The Heligman-Pollard mortality law of the form

$$\mu_x = \sum_{k=1}^m a_k \exp \left[-b_k (f_k(x) - c_k)^{\alpha_k} \right]$$

was developed to decompose mortality according to three stages of life where the constants $\{a_k, b_k, c_k, \alpha_k\}$ are mortality parameters and $f_k(x) = \ln x$ or $f_k(x) = x$ [2].

However, the major problem currently is that, the generating survival function, the probability of survival function and the hazard rate functions have not been developed as a result of the emerging analytical intractability of the force of mortality function and this accounts for the reason why its associated life insurance monetary functions have not been developed till date.

Since mortality intensity evolves in continuous time, a basic functional representation of mortality risks is the severity to die function which defines the aggregate death severity that a life is exposed to over time. Consequently, the probability that a life survives to an age is the exponentiated negative severity to die function within an arbitrarily closed time interval.

A sound knowledge of mortality intensity could be applied to measure: (i) the severity to die function

$$\int_{\mathbf{R}^+} \mu_{x+t} dt \quad (ii) \text{ the survival probability function } {}_t P_x$$

(iii) the curve of death $l_x \mu_x$ of an insured life all applicable to carry out actuarial valuations in life insurance schemes. Thus all life contingent events represent direct application of numerical techniques in computing life table functions to ensure cost effective approximations.

Actuaries are required to conduct precise mortality estimations using the appropriate numerical technique since valuation of life insurance schemes depend heavily on both mortality rate and interest rate intensities when performing actuarial valuations on life insurance and pension schemes.

The inability to conduct accurate mortality results will expose life insurance companies to unforeseen risk of insolvencies. Since insurance, annuity and pension funds heavily depend on life tables when conducting actuarial valuations for policies and premium computations, there is a dire need for accurate and reliable mortality tables.

Grave concern in actuarial practice is the estimation problem relating to nonlinear mortality intensities that have been functionally formulated to obtain life insurance products. In related studies in [5, 6] it was observed that the uniform distribution of deaths interpolation $l_{x+t} = (1-t)l_x + tl_{x+1}$ has been specifically developed to solve most mortality estimation problems and expedite computational convenience while inadvertently trading off mathematical accuracy and consistencies. As a result, the same number of lives l_x at age x is expected to die periodically and because of this assumption, the authors applied the linear interpolation to estimate non-linear mortality functions in all actuarial computations but were unaware that the survival function $l_x \in C^K$.

From the foregoing, the following observations are evident: (i) Mortality functions based on linear interpolation are not analytically consistent with each other such that same mortality functions would mean different approximations between integral ages. (ii) Because of the identified functional inconsistencies, it can no longer be reasonably assumed that linear interpolation assumption is competent to estimate non-linear mortality functions. (iii) The assumption may not always be acceptable to a larger extent because equal number of lives cannot be assumed to die periodically. (iv) The second and higher derivatives of the survival function l_x would be zero and hence renders the applications of both Taylor's and Euler-Maclaurin's series to mortality estimations inapplicable.

Arguably, further evidence of inconsistent mortality computations emerged in [7, 8] that the uniform distribution of death through linear interpolation assumption over which actuarial computations are currently based may no longer be adequate for many life insurance applications. Unknown in mathematics literature, some workers [9, 10] foresaw this problem in advance and constructed an interpolation resulting in mortality matrices which can be applied to model mortality rate intensity.

The deep investigation in [8] leading to the discovery of key computational deficiencies of linear interpolation was carried out to *expose* the inconsistencies in linear interpolation mortality assumption. The inadequate rate of mortality computations used in life insurance pricing and in mortality assumptions poses a serious challenge to life insurance operations. Unfortunately actuaries do not seem to observe this potential threat ahead and life insurance regulators similarly are being myopic of the possible dire consequences on the industry. The rising waves of research gaps therefore constitute critical concerns for life office to the extent that if they are not

reasonably contained immediately can lead to insolvencies of life offices.

Let l_x be a strictly decreasing function of real age such that if $x < y$ then $l_x > l_y$, the reason for this is that the insured lives must always die at an instant.

(ii) The number of deaths between ages x and $x+1$ is defined as $d_x = l_x - l_{x+1}$. Since l_x is strictly decreasing, $l_\Omega \rightarrow 0$ as no life exists towards the end of mortality table. The radix l_0 is the number of insured surviving at beginning of the mortality table.

Differential equation governing the force of mortality μ_x at an instant

According to [7, 11 – 14], the ratio of $-\frac{dl_{x+\zeta}}{d\zeta}$ to $l_{x+\zeta}$

at age x and time ζ represents the mortality rate intensity. The ratio is the instantaneous force of mortality

$$\mu_x = -\lim_{\Delta \rightarrow 0} \frac{l_x - l_{x+\Delta}}{\Delta} \tag{5}$$

$$\mu_{x+\zeta} = -\frac{1}{l_{x+\zeta}} \frac{dl_{x+\zeta}}{d\zeta} = -\frac{d}{d\zeta} \ln l_{x+\zeta} \tag{6}$$

$$\frac{dl_{x+\zeta}}{d\zeta} = -\mu_{x+\zeta} l_{x+\zeta} \Rightarrow dl_{x+\zeta} = -\mu_{x+\zeta} l_{x+\zeta} d\zeta \tag{7}$$

Integrating (7) from age 0 to age $\Omega - x$, we obtain

$$\int_0^{\Omega-x} dl_{x+\zeta} = - \int_0^{\Omega-x} \mu_{x+\zeta} l_{x+\zeta} d\zeta \tag{8}$$

$$\left[l_{x+\zeta} \right]_0^{\Omega-x} = - \int_0^{\Omega-x} \mu_{x+\zeta} l_{x+\zeta} d\zeta \tag{9}$$

Recall that $l_\Omega \rightarrow 0$ as no live survives at highest age Ω in the mortality table.

$$l_{x+\Omega-x} - l_x = - \int_0^{\Omega-x} \mu_{x+\zeta} l_{x+\zeta} d\zeta \tag{10}$$

The survivor's function at an arbitrary age x is then obtained as

$$l_x = \int_0^{\Omega-x} \mu_{x+\zeta} l_{x+\zeta} d\zeta \tag{11}$$

Now, evaluating (9) from age 0 to age 1 and applying condition (ii) above

$$\left[l_{x+\zeta} \right]_0^1 = - \int_0^1 \mu_{x+\zeta} l_{x+\zeta} d\zeta \Rightarrow l_{x+1} - l_x = - \int_0^1 \mu_{x+\zeta} l_{x+\zeta} d\zeta \tag{12}$$



The number of deaths between ages x and $x+1$ is obtained as

$$d_x = \int_0^1 \mu_{x+\zeta} l_{x+\zeta} d\zeta \tag{13}$$

Again, evaluating (9) from age 0 to age θ we have

$$[l_{x+\zeta}]_0^\theta = -\int_0^\theta \mu_{x+\zeta} l_{x+\zeta} d\zeta \Rightarrow l_{x+\theta} - l_x = -\int_0^\theta \mu_{x+\zeta} l_{x+\zeta} d\zeta \tag{14}$$

The probability that a life aged x dies before reaching age $x+\theta$ years is obtained by dividing (14) by l_x

$${}_\theta q_x = \int_0^\theta \mu_{x+\zeta} \frac{l_{x+\zeta}}{l_x} d\zeta = \int_0^\theta \mu_{x+\zeta} ({}_\zeta p_x) d\zeta \tag{15}$$

The severity to die is the hazard function

$$h_x(\theta) = \int_0^\theta \mu_{x+\zeta} d\zeta \text{ obtained by integrating both sides}$$

of (6) from time 0 to age θ ,

$$\int_0^\theta \mu_{x+\zeta} d\zeta = -\int_0^\theta d \ln l_{x+\zeta} \tag{16}$$

$$\int_0^\theta \mu_{x+\zeta} d\zeta = -(\ln l_{x+\theta} - \ln l_x) = -\ln \frac{l_{x+\theta}}{l_x} \tag{17}$$

$$-\ln({}_\theta p_x) = \int_0^\theta \mu_{x+\zeta} d\zeta \tag{18}$$

$$\begin{aligned} ({}_t p_x) &= \exp\left(-\int_{-x}^\theta \mu_{x+u} du + \int_{-x}^0 \mu_{x+u} du\right) = \exp\left(-\int_{-x}^0 \mu_{x+u} du - \int_0^\theta \mu_{x+u} du + \int_{-x}^0 \mu_{x+u} du\right) \\ &= e^{-\int_0^\theta \mu(x+u) du} \end{aligned} \tag{19b}$$

Differentiating the survival function (19) partially with respect to t , we obtain death density

$$\frac{\partial}{\partial t} ({}_t p_x) = \frac{\partial}{\partial t} e^{-\int_0^t \mu_{x+\zeta} d\zeta} = -\mu_{x+t} e^{-\int_0^t \mu_{x+\zeta} d\zeta} = -\mu_{x+t} ({}_t p_x) = f_{T_x}(t) \tag{20}$$

Therefore, from the arguments above, we state unequivocally the following results

$$\frac{d \ln l_x}{dx} = \lim_{\xi \rightarrow 0} \left(\frac{-\int_0^\xi e^{-\delta s} ({}_s p_x) \mu_{x+s} ds}{\xi} \right) \tag{21}$$

Applying L'Hopital's rule, equation (21) becomes

$$\frac{d \ln l_x}{dx} = -\lim_{\xi \rightarrow 0} \left(\frac{d}{d\xi} \left\{ \int_0^\xi e^{-\delta s} ({}_s p_x) \mu_{x+s} ds \right\} \right) \tag{22}$$

$$\frac{d \ln l_x}{dx} = -\lim_{\xi \rightarrow 0} [e^{-\delta \xi} ({}_\xi p_x) \mu_{x+\xi}] = -\mu_x \tag{23}$$

The probability that a life aged x survives to the next age $(x+\theta)$ years becomes

$$({}_\theta p_x) = \exp\left(-\int_0^\theta \mu_{x+\zeta} d\zeta\right) \tag{19}$$

as $\theta \rightarrow \infty$ in the result above, $({}_t p_x) = 0$ and the

integral $\int_0^\infty \mu_{x+\zeta} d\zeta = \ln 0 = \infty$ while

as $\theta \rightarrow 0$, $({}_t p_x) = 1$ and the integral

$$\int_0^\theta \mu_{x+\zeta} d\zeta = \ln 1 = 0. \text{ Therefore the function } h_x(t)$$

satisfies the conditions that $\lim_{\theta \rightarrow 0} h_x(\theta) = 0$ and

$$\lim_{\theta \rightarrow \infty} h_x(\theta) = \infty.$$

The integral in equation (19) has the property that

$$({}_\theta p_x) = \exp\left(-\int_0^{\theta+x} \mu_\zeta d\zeta + \int_0^x \mu_\zeta d\zeta\right) \tag{19a}$$

Letting $\zeta - x = u$, the survival probability becomes;

The distribution and the complementary functions of

$$T_x \text{ are } F_{T_x}(s) = \int_0^s f_{T_x}(\zeta) d\zeta \text{ and}$$

$$S_{T_x}(s) = 1 - F_{T_x}(s) \tag{24}$$

However, in [15, 16], the continuous death density is in other words defined as

$$f_{T_x}(s) = \frac{d}{ds} ({}_s q_x) \tag{25}$$

The distribution of T_x can be actuarially expressed as

$$F_{T_x}(s) = \mathbf{P}(T_x \leq s) = {}_s q_x \tag{26}$$

and

$$S_{T_x}(s) = \mathbf{P}(T_x > s) \tag{27}$$

Materials and Methods

The method of severity to die function

The severity to die in form integrated hazard function is required to model the probability of survival function. In order to account for increase in background mortality, we set $m = 2$ and $n = 2$ in (4) to obtain the mortality function (28a) where A, B, C, H are mortality parameters satisfying law of parsimony (Occam's razor)

$$\mu_x = A + Hx + BC^x \tag{28a}$$

Integrating both sides of (28a) from zero to an arbitrary age x of the insured to obtain the severity to die function, we have

$$\int_0^x \mu_t dt = \int_0^x (A + Ht + BC^t) dt \tag{28b}$$

$$\int_0^x \mu(t) dt = \left[At + \frac{Ht^2}{2} + \frac{BC^t}{\log_e C} \right]_0^x \tag{28c}$$

$$\int_0^x \mu(t) dt = -(\log_e s)x - (\log_e W)x^2 - \frac{(\log_e C)(\log_e g)C^x}{\log_e C} + \frac{(\log_e C)(\log_e g)}{\log_e C} \tag{28f}$$

Simplifying (28f), we have

$$\int_0^x \mu(t) dt = -(\log_e s)x - (\log_e W)x^2 - (\log_e g)C^x + (\log_e g) \tag{28g}$$

$$\int_0^x \mu(t) dt = -x \log_e s - x^2 \log_e W - C^x \log_e g + \log_e g \tag{28h}$$

$$\int_0^x \mu(t) dt = -\log_e s^x - \log_e W^{x^2} - \log_e g^{C^x} + \log_e g \tag{28i}$$

The survival probability function is given by

$${}_t p_x = \frac{l_{x+t}}{l_x} = \exp\left(-\int_0^t \mu_{x+\theta} d\theta\right) \tag{28j}$$

In order to obtain the exponential function of the LHS in (28i), we set $x = 0$ in (28j) as follows

$${}_t p_0 = \exp\left(-\int_0^t \mu_\theta d\theta\right) \tag{28k}$$

$$l_t = l_0 \exp\left(-\int_0^t \mu_\theta d\theta\right) \tag{28l}$$

The subscript t in (28j) is an arbitrary age and consequently, we replace t by insured's age x

$$l_x = l_0 \exp\left(-\int_0^x \mu_t dt\right) \tag{28m}$$

Substituting (28i) into (28m) and have

$$\int_0^x \mu(t) dt = Ax + \frac{Hx^2}{2} + \frac{BC^x}{\log_e C} - \frac{B}{\log_e C} \tag{28d}$$

In order to simplify (28d), we use the following transformations

$$\left\{ \begin{aligned} A &= \log_e S^{-1} = (-1)\log_e S \\ H &= \log_e W^{-2} = -2\log_e W \\ B &= -(\log_e C)(\log_e g) \\ \log_e g &= \frac{-B}{\log_e C} \\ K &= \frac{l_0}{g} \end{aligned} \right. \tag{28e}$$

Substituting the transformations in (28e) into (28d) we obtain;

$$l_x = l_0 e^{\left[-\log_e s^x - (\log_e W^{x^2}) - \log_e g^{C^x} + \log_e g\right]} = l_0 e^{\left[\log_e s^x + \log_e W^{x^2} + \log_e g^{C^x} - \log_e g\right]} \tag{28n}$$

Simplifying (28n) and obtain

$$l_x = l_0 e^{\left[\log_e \left(\frac{s^x W^{x^2} g^{C^x}}{g}\right)\right]} = l_0 \frac{s^x W^{x^2} g^{C^x}}{g} \tag{28o}$$

The number of lives surviving to age x is given as

$$l_x = \frac{l_0}{g} s^x W^{x^2} g^{C^x} = K s^x W^{x^2} g^{C^x}; \quad K = \frac{l_0}{g} \tag{28p}$$

Taking logarithms of (28p) and have

$$\log_e l_x = \log_e K + \log_e s^x + \log_e W^{x^2} + \log_e g^{C^x} \tag{29}$$

$$\log_e l_x = (\log_e K) + x(\log_e s) + x^2(\log_e W) + C^x(\log_e g) \tag{30}$$



We need to obtain the values of $\phi = \{\bar{A}; \bar{B}; \bar{E}; \bar{F}\}$

first, hence we use another transformation

Let

$$Y = \log_e l_x; \bar{A} = \log_e K; \bar{B} = \log_e s; \bar{E} = \log_e W; \bar{F} = \log_e g \quad (31)$$

Then

$$e^Y = l_x; e^{\bar{A}} = K; e^{\bar{B}} = S; e^{\bar{E}} = W; e^{\bar{F}} = g \quad (32)$$

$$l_x = e^{\bar{A}} e^{\bar{B}x} e^{\bar{E}x^2} e^{\bar{F}C^x} = e^{\bar{A} + \bar{B}x + \bar{E}x^2 + \bar{F}C^x} \quad (33)$$

Putting (31) in (30) and obtain

$$Y = \bar{A} + \bar{B}x + \bar{E}x^2 + \bar{F}C^x \quad (34)$$

substituting arbitrary five ages

x_1, x_2, x_3, x_4, x_5 in(34) and using equal step lengths

$x_2 - x_1 = x_3 - x_2 = x_4 - x_3 = x_5 - x_4 = h$, we have,

$$x_2 = x_1 + h; x_3 = x_1 + 2h; x_4 = x_1 + 3h; x_5 = x_1 + 4h \quad (35)$$

$$Y_1 = \bar{A} + \bar{B}x_1 + \bar{E}x_1^2 + \bar{F}C^{x_1} \quad (36)$$

$$Y_2 = \bar{A} + \bar{B}x_2 + \bar{E}x_2^2 + \bar{F}C^{x_2} \quad (37)$$

$$Y_3 = \bar{A} + \bar{B}x_3 + \bar{E}x_3^2 + \bar{F}C^{x_3} \quad (38)$$

$$Y_4 = \bar{A} + \bar{B}x_4 + \bar{E}x_4^2 + \bar{F}C^{x_4} \quad (39)$$

$$Y_5 = \bar{A} + \bar{B}x_5 + \bar{E}x_5^2 + \bar{F}C^{x_5} \quad (40)$$

Subtracting (36) from (37)

$$\Delta Y_1 = Y_2 - Y_1 = \bar{A} + \bar{B}x_2 + \bar{E}x_2^2 + \bar{F}C^{x_2} - (\bar{A} + \bar{B}x_1 + \bar{E}x_1^2 + \bar{F}C^{x_1}) \quad (41)$$

$$\Delta Y_1 = \bar{A} + \bar{B}x_2 + \bar{E}x_2^2 + \bar{F}C^{x_2} - \bar{A} - \bar{B}x_1 - \bar{E}x_1^2 - \bar{F}C^{x_1} \quad (42)$$

$$\Delta Y_1 = \bar{B}(x_2 - x_1) + \bar{E}(x_2^2 - x_1^2) + \bar{F}(C^{x_2} - C^{x_1}) \quad (43)$$

$$\Delta Y_1 = \bar{B}(x_2 - x_1) + \bar{E}(x_2 - x_1)(x_2 + x_1) + \bar{F}(C^{x_2} - C^{x_1}) \quad (44)$$

Substituting x_2 defined in (35) into (44) we have

$$Y_2 - Y_1 = h\bar{B} + h\bar{E}(2x_1 + h) + \bar{F}C^{x_1}(C^h - 1) \quad (45)$$

Taking second difference between (38) and (37)

$$\Delta Y_2 = Y_3 - Y_2 = \bar{A} + \bar{B}x_3 + \bar{E}x_3^2 + \bar{F}C^{x_3} - (\bar{A} + \bar{B}x_2 + \bar{E}x_2^2 + \bar{F}C^{x_2}) \quad (46)$$

$$\Delta Y_2 = \bar{B}(x_3 - x_2) + \bar{E}(x_3^2 - x_2^2) + \bar{F}(C^{x_3} - C^{x_2}) \quad (47)$$

$$\Delta Y_2 = h\bar{B} + \bar{E}(x_3 - x_2)(x_3 + x_2) + \bar{F}(C^{x_3} - C^{x_2}) \quad (48)$$

Following same procedure as before and substitute for x_3 defined in (35)

$$\Delta Y_2 = h\bar{B} + h\bar{E}(x_1 + 2h + x_1 + h) + \bar{F}(C^{x_1+2h} - C^{x_1+h}) \quad (49)$$

$$Y_3 - Y_2 = h\bar{B} + h\bar{E}(2x_1 + 3h) + \bar{F}C^{x_1+h}(C^h - 1) \quad (50)$$

Taking third difference using (38) and (39) and substitute for x_4

$$\Delta Y_3 = Y_4 - Y_3 = \bar{A} + \bar{B}x_4 + \bar{E}x_4^2 + \bar{F}C^{x_4} - (\bar{A} + \bar{B}x_3 + \bar{E}x_3^2 + \bar{F}C^{x_3}) \quad (51)$$

$$\Delta Y_3 = h\bar{B} + h\bar{E}(x_1 + 3h + x_1 + 2h) + \bar{F}(C^{x_1+3h} - C^{x_1+2h}) \quad (52)$$

$$Y_4 - Y_3 = h\bar{B} + h\bar{E}(2x_1 + 5h) + \bar{F}C^{x_1+2h}(C^h - 1) \quad (53)$$

Taking fourth difference using (39) and (40) and substitute for x_5

$$\Delta Y_4 = Y_5 - Y_4 = \bar{A} + \bar{B}x_5 + \bar{E}x_5^2 + \bar{F}C^{x_5} - (\bar{A} + \bar{B}x_4 + \bar{E}x_4^2 + \bar{F}C^{x_4}) \quad (54)$$

$$\Delta Y_4 = \bar{B}(x_5 - x_4) + \bar{E}(x_5^2 - x_4^2) + \bar{F}(C^{x_5} - C^{x_4}) \quad (55)$$

$$\Delta Y_4 = \bar{B}(x_5 - x_4) + \bar{E}(x_5 - x_4)(x_5 + x_4) + \bar{F}(C^{x_5} - C^{x_4}) \quad (56)$$

$$\Delta Y_4 = h\bar{B} + h\bar{E}(x_1 + 4h + x_1 + 3h) + \bar{F}(C^{x_1+4h} - C^{x_1+3h}) \quad (57)$$

$$Y_5 - Y_4 = h\bar{B} + h\bar{E}(2x_1 + 7h) + \bar{F}C^{x_1+3h}(C^h - 1) \quad (58)$$

subtracting (45) from (50) and obtain

$$Y_3 - Y_2 - (Y_2 - Y_1) = h\bar{B} + h\bar{E}(2x_1 + 3h) + \bar{F}C^{x_1+h}(C^h - 1) - \left[\begin{matrix} h\bar{B} + h\bar{E}(2x_1 + h) \\ + \bar{F}C^{x_1}(C^h - 1) \end{matrix} \right] \quad (59)$$

$$Y_3 - Y_2 - Y_2 + Y_1 = h\bar{B} + h\bar{E}(2x_1 + 3h) + \bar{F}C^{x_1+h}(C^h - 1) - h\bar{B} - h\bar{E}(2x_1 + h) - \bar{F}C^{x_1}(C^h - 1) \quad (60)$$

$$Y_3 - 2Y_2 + Y_1 = h\bar{B} - h\bar{B} + h\bar{E}(2x_1 + 3h) - h\bar{E}(2x_1 + h) + \bar{F}C^{x_1+h}(C^h - 1) - \bar{F}C^{x_1}(C^h - 1) \quad (61)$$

$$Y_3 - 2Y_2 + Y_1 = 2h^2\bar{E} + \bar{F}C^{x_1}(C^h - 1)^2 \quad (62)$$

Subtracting (50) from (53) and have

$$Y_4 - Y_3 - (Y_3 - Y_2) = h\bar{B} + h\bar{E}(2x_1 + 5h) + \bar{F}C^{x_1+2h}(C^h - 1) - \left[\begin{matrix} h\bar{B} + h\bar{E}(2x_1 + 3h) \\ + \bar{F}C^{x_1+h}(C^h - 1) \end{matrix} \right] \quad (63)$$

$$Y_4 - Y_3 - (Y_3 - Y_2) = h\bar{B} + h\bar{E}(2x_1 + 5h) + \bar{F}C^{x_1+2h}(C^h - 1) - h\bar{B} - h\bar{E}(2x_1 + 3h) - \bar{F}C^{x_1+h}(C^h - 1) \quad (64)$$

$$Y_4 - 2Y_3 + Y_2 = h\bar{B} - h\bar{B} + h\bar{E}(2x_1 + 5h) - h\bar{E}(2x_1 + 3h) + \bar{F}C^{x_1+2h}(C^h - 1) - \bar{F}C^{x_1+h}(C^h - 1) \quad (65)$$

$$Y_4 - 2Y_3 + Y_2 = 2h\bar{E}x_1 + 5h^2\bar{E} - 2h\bar{E}x_1 - 3h^2\bar{E} + \bar{F}C^{x_1+h}(C^h - 1)^2 \quad (66)$$

$$Y_4 - 2Y_3 + Y_2 = 2h^2\bar{E} + \bar{F}C^{x_1+h}(C^h - 1)^2 \quad (67)$$

Subtract (53) from (58) and have

$$Y_5 - Y_4 - (Y_4 - Y_3) = h\bar{B} + h\bar{E}(2x_1 + 7h) + \bar{F}C^{x_1+3h}(C^h - 1) - \left[\begin{matrix} h\bar{B} + h\bar{E}(2x_1 + 5h) \\ + \bar{F}C^{x_1+2h}(C^h - 1) \end{matrix} \right] \quad (68)$$

$$Y_5 - Y_4 - (Y_4 - Y_3) = h\bar{B} + h\bar{E}(2x_1 + 7h) + \bar{F}C^{x_1+3h}(C^h - 1) - h\bar{B} - h\bar{E}(2x_1 + 5h) - \bar{F}C^{x_1+2h}(C^h - 1) \quad (69)$$

$$Y_5 - Y_4 - (Y_4 - Y_3) = h\bar{B} - h\bar{B} + h\bar{E}(2x_1 + 7h) - h\bar{E}(2x_1 + 5h) + \bar{F}C^{x_1+3h}(C^h - 1) - \bar{F}C^{x_1+2h}(C^h - 1) \quad (70)$$

$$Y_5 - 2Y_4 + Y_3 = 2h\bar{E}x_1 + 7h^2\bar{E} - 2h\bar{E}x_1 - 5h^2\bar{E} + \bar{F}C^{x_1+2h}(C^h - 1)^2 \quad (71)$$

$$Y_5 - 2Y_4 + Y_3 = 2h^2\bar{E} + \bar{F}C^{x_1+2h}(C^h - 1)^2 \quad (72)$$

Subtract (62) from (67) and have

$$Y_4 - 2Y_3 + Y_2 - (Y_3 - 2Y_2 + Y_1) = 2h^2\bar{E} + \bar{F}C^{x_1+h}(C^h - 1)^2 - (2h^2\bar{E} + \bar{F}C^{x_1}(C^h - 1)^2) \quad (73)$$

$$Y_4 - 2Y_3 + Y_2 - Y_3 + 2Y_2 - Y_1 = 2h^2\bar{E} + \bar{F}C^{x_1+h}(C^h - 1)^2 - 2h^2\bar{E} - \bar{F}C^{x_1}(C^h - 1)^2 \quad (74)$$

$$Y_4 - 3Y_3 + 3Y_2 - Y_1 = \bar{F}C^{x_1}(C^h - 1)^3 \quad (75)$$

Subtract (67) from (72) and obtain

$$Y_5 - 2Y_4 + Y_3 - (Y_4 - 2Y_3 + Y_2) = 2h^2\bar{E} + \bar{F}C^{x_1+2h}(C^h - 1)^2 - [2h^2\bar{E} + \bar{F}C^{x_1+h}(C^h - 1)^2] \quad (76)$$

$$Y_5 - 2Y_4 + Y_3 - (Y_4 - 2Y_3 + Y_2) = 2h^2\bar{E} + \bar{F}C^{x_1+2h}(C^h - 1)^2 - 2h^2\bar{E} - \bar{F}C^{x_1+h}(C^h - 1)^2 \quad (77)$$

$$Y_5 - 2Y_4 + Y_3 - Y_4 + 2Y_3 - Y_2 = 2h^2\bar{E} + \bar{F}C^{x_1+2h}(C^h - 1)^2 - 2h^2\bar{E} - \bar{F}C^{x_1+h}(C^h - 1)^2 \quad (78)$$

$$Y_5 - 3Y_4 + 3Y_3 - Y_2 = \bar{F}C^{x_1+h}(C^h - 1)^3 \quad (79)$$



combining equations (75) and (79) together again and have

$$\begin{cases} Y_4 - 3Y_3 + 3Y_2 - Y_1 = \bar{F}C^{x_1} (C^h - 1)^3 \\ Y_5 - 3Y_4 + 3Y_3 - Y_2 = \bar{F}C^{x_1+h} (C^h - 1)^3 \end{cases} \quad (80)$$

Dividing the lower equation by the upper in system (80) to obtain the ageing parameter C

$$C^h = \frac{Y_5 - 3Y_4 + 3Y_3 - Y_2}{Y_4 - 3Y_3 + 3Y_2 - Y_1} \quad (81)$$

Setting (81) in (79) and have

$$Y_5 - 3Y_4 + 3Y_3 - Y_2 = \bar{F} \left[\left(\frac{Y_5 - 3Y_4 + 3Y_3 - Y_2}{Y_4 - 3Y_3 + 3Y_2 - Y_1} \right)^{\frac{x_1+h}{h}} \right] \left[\frac{Y_5 - 3Y_4 + 3Y_3 - Y_2}{Y_4 - 3Y_3 + 3Y_2 - Y_1} - 1 \right]^3 \quad (82)$$

$$\bar{F} \left[\left(\frac{Y_5 - 3Y_4 + 3Y_3 - Y_2}{Y_4 - 3Y_3 + 3Y_2 - Y_1} \right)^{\frac{x_1+h}{h}} \right] \left[\frac{Y_5 - 3Y_4 + 3Y_3 - Y_2 - Y_4 + 3Y_3 - 3Y_2 + Y_1}{Y_4 - 3Y_3 + 3Y_2 - Y_1} \right]^3 = Y_5 - 3Y_4 + 3Y_3 - Y_2 \quad (83)$$

$$\bar{F} \left[\left(\frac{Y_5 - 3Y_4 + 3Y_3 - Y_2}{Y_4 - 3Y_3 + 3Y_2 - Y_1} \right)^{\frac{x_1+h}{h}} \right] \left[\frac{Y_5 - 4Y_4 + 6Y_3 - 4Y_2 + Y_1}{Y_4 - 3Y_3 + 3Y_2 - Y_1} \right]^3 = Y_5 - 3Y_4 + 3Y_3 - Y_2 \quad (84)$$

$$\bar{F} = \frac{Y_5 - 3Y_4 + 3Y_3 - Y_2}{\left[\left(\frac{Y_5 - 3Y_4 + 3Y_3 - Y_2}{Y_4 - 3Y_3 + 3Y_2 - Y_1} \right)^{\frac{x_1+h}{h}} \right] \left[\frac{Y_5 - 4Y_4 + 6Y_3 - 4Y_2 + Y_1}{Y_4 - 3Y_3 + 3Y_2 - Y_1} \right]^3} \quad (85)$$

Using the first equation in (62), we obtain

$$\bar{E} = \frac{Y_3 - 2Y_2 + Y_1 - \bar{F}C^{x_1} (C^h - 1)^2}{2h^2} \quad (86)$$

Recall from (45)

$$\bar{B} = \frac{Y_2 - Y_1 - h\bar{E}(2x_1 + h) - \bar{F}C^{x_1} (C^h - 1)}{h} \quad (87)$$

Using equation (36), we have

$$\bar{A} = Y_1 - \bar{B}x_1 - \bar{E}x_1^2 - \bar{F}C^{x_1} \quad (88)$$



It is on this argument that this study is based on age dependent mortality intensities satisfying Occam's razor and moreover, life insurance schemes issued for protection and pension purposes are defined on long or short term durations.

The data set employed in this study is a single year of age population data set l_x taken from the published DAV 2008 German survival data for male. German data is considered because the country has proven mortality data collection records. Moreover Germany has death rate pattern similar to that of Nigeria. Germany has death rate of 12.3% while Nigeria has death rate of 13.1%. The ages considered in the analysis are in single year within the age range 0 – 120. Our goal in this study is to compute the values of instantaneous mortality, curve of death, probability of death and probability of survival for a life insurance contract offered to a life aged x .

Generalised Makeham's mortality life table

In the Table that follows, x represents the age, l_x is the decreasing survival function, μ_x is the mortality rate intensity, $l_x\mu_x$ is the curve of death. p_x is the probability that a life aged x survives to the next age $x+1$, q_x is the probability that a life aged x will die before reaching age $x+1$, The mortality table below was generated using equations (28a), (81), (86), (87), (88).

Results and Discussion

The study focuses on extended mathematical modelling and implementations of life table functions using Generalised Makeham's law of mortality. The findings are discussed along the male gender disparity.

In Table 1, the neonatal and infant mortality from birth to the first birthday is high but the mortality rate intensity falls rapidly within $0 \leq x \leq 7$ and becomes relatively stable within $8 \leq x \leq 9$ before attaining a minimum risk around age 10. The relative stability in this interval can be due to the predictable trend's gradual reduction in mortality over time or it can be due to idiosyncratic conditions neither due to shocks nor trends. This is an empirical evidence showing that the mortality rate generated through $GM(2,2)$ actually declines at 10. The parameters of $GM(2,2)$ is,

$$(A, B, H, C)_{MALE} = (0.003012821, 4.07194 \times 10^{-05}, -0.000100466, 1.102923606) \quad (90)$$

From the method of successive differencing employed to model $GM(2,2)$, the male ageing parameter values 1.102923606 fall within the globally accepted interval $1.08 \leq C \leq 1.12$ for the $GM(m,n)$ family. This method is superior to the method of maximum

likelihood estimation adopted in [17] where the ageing C parameter is estimated as 1.024738. The authors' method violated the permissible interval.

The mortality rate intensity further declines within $11 \leq x \leq 33$ and finally increases in the interval $34 \leq x \leq 120$. The possible medical intervention in the public health system to manage young adulthood mortality could be observed as the male's mortality rates have reduced. The decline in mortality rates within the former interval has significant impact on life annuities and on life insurers. However, the increase in mortality rates with age in the latter interval could be associated with various factors.

In men specifically, the risk of prostate cancer increases with age around 65 and beyond until it is peaked between 78 and 79. Irrespective of sex, the probability of contracting a chronic sickness or disability increases with age while immunity reduces thus exposing lives more vulnerable to health risk and consequently increasing the risk of mortality.

As a result of deaths occurring over time, the survival curves for male in Figs 1 and 2 exhibit a consistent decrease in the number of lives surviving out of the initial 1000,000 newborns and consequently, their trajectories across ages display negative slope in line

with equation, $-\frac{d}{dx}l_x = \mu_x$. This accounts for the

reason why the curve l_x is relatively flat for male with almost zero gradient. The curve exhibits a sudden sharp decline in the number of survivors around 80 for male signifying that mortality rates are relatively low up till age 80 before subsequently experiencing a sharp increase as observed in Fig. 1. The implication is that in Figs 1 and 2 the points at which the survival curve l_x inflexes is around 80 for both sexes. Furthermore, the curve of death in Fig. 1 tapers to a Mesokurtic curve. The mode of death intensity curve moves towards advanced ages and the concentration of death around the mode is increasing. Consequently, the survival curve l_x is noted to exhibit a rectangularization pattern such that there is an observed progressively concentrated and increased mortality rates during senescent ages. As a result, the survival function assumes the observed rectangular shape. The accidental deaths at young ages are also seen to be increasing. These observed phenomena have pervasive socio-economic implications on the increasing number of lives attaining the retirement age as well as the period of extension during which the life annuity providers will pay out the benefits. Consequently, robust mortality rate estimation has a direct effect on the actuarial present value of future liabilities and the associated level of reserves that life office paying benefits holds. From age 106 in Table 1, accurate estimation of mortality could be challenging due to errors in age heaping reporting from vital statistics and by the low number of lives and of deaths.



Particularly, the interval $107 \leq x \leq 120$ has the following implications. (i) Mortality data are scanty and are not reliable and as a result, the asymptotic behaviour of the underlying parsimonious function of mortality towards the end of mortality table is inconsistent. (ii) It seems many deaths were recorded but do not necessarily represent the point where l_x inflexes.

Table 1: GM (2,2) male mortality table

x	l_x	μ_x	$l_x \mu_x$	q_x	p_x
0	1000000	0.00305354	3054	0.00300100	0.99699900
1	996999	0.00295727	2948	0.00290472	0.99709528
2	994103	0.00286142	2845	0.00280957	0.99719043
3	991310	0.00276605	2742	0.00271560	0.99728440
4	988618	0.00267121	2641	0.00261982	0.99738018
5	986028	0.00257695	2541	0.00252731	0.99747269
6	983536	0.00248332	2442	0.00243407	0.99756593
7	981142	0.00239040	2345	0.00234115	0.99765885
8	978845	0.00229825	2250	0.00225061	0.99774939
9	976642	0.00220696	2155	0.00215842	0.99784158
10	974534	0.00211662	2063	0.00207073	0.99792927
11	972516	0.00202731	1972	0.00198043	0.99801957
12	970590	0.00193916	1882	0.00189369	0.99810631
13	968752	0.00185227	1794	0.00180851	0.99819149
14	967000	0.00176678	1708	0.00172285	0.99827715
15	965334	0.00168283	1624	0.00163985	0.99836015
16	963751	0.00160059	1543	0.00155953	0.99844047
17	962248	0.00152021	1463	0.00147987	0.99852013
18	960824	0.00144191	1385	0.00140296	0.99859704
19	959476	0.00136588	1311	0.00132781	0.99867219
20	958202	0.00129238	1238	0.00125548	0.99874452
21	956999	0.00122164	1169	0.00118704	0.99881296
22	955863	0.00115397	1103	0.00112150	0.99887850
23	954791	0.00108967	1040	0.00105782	0.99894218
24	953781	0.00102909	982	0.00100023	0.99899977
25	952827	0.00097262	927	0.00094561	0.99905439
26	951926	0.00092068	876	0.00089608	0.99910392
27	951073	0.00087373	831	0.00085272	0.99914728
28	950262	0.00083229	791	0.00081346	0.99918654
29	949489	0.00079693	757	0.00078147	0.99921853
30	948747	0.00076826	729	0.00075679	0.99924321
31	948029	0.00074699	708	0.00073943	0.99926057
32	947328	0.00073386	695	0.00073153	0.99926847
33	946635	0.00072973	691	0.00073101	0.99926899
34	945943	0.00073551	696	0.00074317	0.99925683
35	945240	0.00075223	711	0.00076489	0.99923511
36	944517	0.00078100	738	0.00080041	0.99919959
37	943761	0.00082308	777	0.00084979	0.99915021
38	942959	0.00087983	830	0.00091520	0.99908480
39	942096	0.00095276	898	0.00099565	0.99900435
40	941158	0.00104354	982	0.00109652	0.99890348
41	940126	0.00115400	1085	0.00121792	0.99878208
42	938981	0.00128618	1208	0.00136105	0.99863895
43	937703	0.00144229	1352	0.00152927	0.99847073
44	936269	0.00162481	1521	0.00172707	0.99827293
45	934652	0.00183646	1716	0.00195367	0.99804633
46	932826	0.00208024	1941	0.00221478	0.99778522
47	930760	0.00235944	2196	0.00251193	0.99748807
48	928422	0.00267772	2486	0.00285000	0.99715000
49	925776	0.00303910	2814	0.00323404	0.99676596
50	922782	0.00344802	3182	0.00366826	0.99633174
51	919397	0.00390936	3594	0.00415490	0.99584510
52	915577	0.00442852	4055	0.00470305	0.99529695
53	911271	0.00501146	4567	0.00531785	0.99468215
54	906425	0.00566474	5135	0.00600491	0.99399509
55	900982	0.00639560	5762	0.00677372	0.99322628
56	894879	0.00721202	6454	0.00763008	0.99236992
57	888051	0.00812281	7213	0.00858397	0.99141603
58	880428	0.00913768	8045	0.00964531	0.99035469
59	871936	0.01026734	8952	0.01082534	0.98917466
60	862497	0.01152362	9939	0.01213569	0.98786431
61	852030	0.01291953	11008	0.01358755	0.98641245
62	840453	0.01446946	12161	0.01519776	0.98480224
63	827680	0.01618926	13400	0.01698120	0.98301880
64	813625	0.01809640	14724	0.01895345	0.98104655
65	798204	0.02021017	16132	0.02113370	0.97886630
66	781335	0.02255183	17621	0.02354560	0.97645440
67	762938	0.02514485	19184	0.02620527	0.97379473
68	742945	0.02801510	20814	0.02914348	0.97085652
69	721293	0.03119110	22498	0.03238212	0.96761788
70	697936	0.03470433	24221	0.03595315	0.96404685
71	672843	0.03858949	25965	0.03988300	0.96011700
72	646008	0.04288486	27704	0.04421307	0.95578693
73	617446	0.04763267	29411	0.04897270	0.95102730
74	587208	0.05287949	31051	0.05420566	0.94579434
75	555378	0.05867666	32588	0.05995196	0.94004804
76	522082	0.06508084	33978	0.06626162	0.93373838
77	487488	0.07215451	35174	0.07318129	0.92681871
78	451813	0.07996656	36130	0.08075908	0.91924092
79	415325	0.08859299	36795	0.08906037	0.91093963
80	378336	0.09811763	37121	0.09813235	0.90186765
81	341209	0.10863292	37067	0.10804522	0.89195478
82	304343	0.12024083	36594	0.11886260	0.88113740
83	268168	0.13305379	35681	0.13064944	0.86935056
84	233132	0.14719586	34316	0.14347666	0.85652334
85	199683	0.16280382	32509	0.15740949	0.84259051
86	168251	0.18002855	30290	0.17252201	0.82747799
87	139224	0.19903645	27711	0.18888985	0.81111015
88	112926	0.22001105	24845	0.20657776	0.79342224
89	89598	0.24315477	21786	0.22563004	0.77436996
90	69382	0.26869087	18642	0.24614453	0.75385547
91	52304	0.29686557	15527	0.26812481	0.73187519
92	38280	0.32795046	12554	0.29164054	0.70835946
93	27116	0.36224506	9823	0.31675026	0.68324974
94	18527	0.40007972	7412	0.34333675	0.65666325
95	12166	0.44181880	5375	0.37152721	0.62847279
96	7646	0.48786416	3730	0.40125556	0.59874444
97	4578	0.53865901	2466	0.43228484	0.56771516
98	2599	0.59469220	1546	0.46479415	0.53520585
99	1391	0.65650286	913	0.49820273	0.50179727
100	698	0.72468563	506	0.53295129	0.46704871
101	326	0.79989637	261	0.56748466	0.43251534
102	141	0.88285840	124	0.60283688	0.39716312
103	56	0.97436953	55	0.64285714	0.35714286
104	20	1.07530965	22	0.70000000	0.30000000
105	6	1.18664924	7	0.66666667	0.33333333
106	2	1.30945863	3	1.00000000	0.00000000
107	0	1.44491836	0	0.00000000	1.00000000
108	0	1.59433042	0	0.00000000	1.00000000
109	0	1.75913086	0	0.00000000	1.00000000
110	0	1.94090349	0	0.00000000	1.00000000
111	0	2.14139515	0	0.00000000	1.00000000
112	0	2.36253248	0	0.00000000	1.00000000
113	0	2.60644040	0	0.00000000	1.00000000
114	0	2.87546255	0	0.00000000	1.00000000
115	0	3.17218377	0	0.00000000	1.00000000
116	0	3.49945494	0	0.00000000	1.00000000
117	0	3.86042038	0	0.00000000	1.00000000
118	0	4.25854803	0	0.00000000	1.00000000
119	0	4.69766275	0	0.00000000	1.00000000
120	0	5.18198308	0	0.00000000	1.00000000

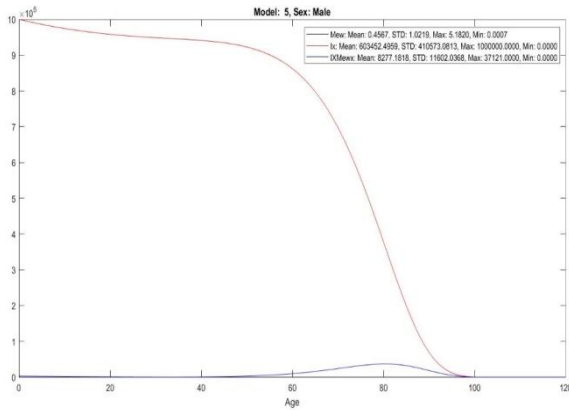


Figure 1: GM (2,2) male survival and curve of death functions

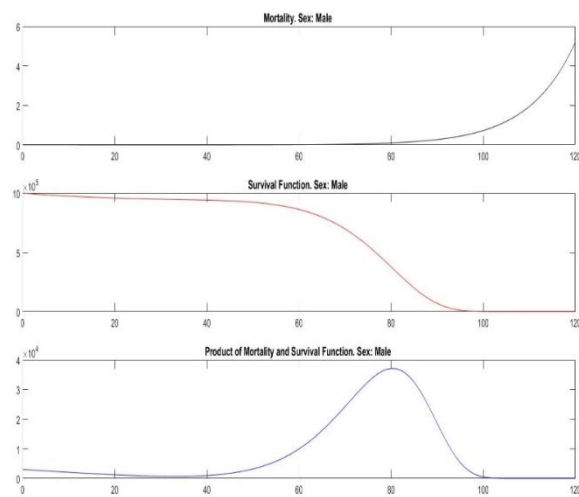


Figure 2: GM (2,2) male's mortality, survival and curve of death functions

The survival l_x function in Figs. 1 and 2 representing the expected number of lives surviving to age x out of an initial group of 1000000 lives clearly forms asymptote on x axis.

In Table 1 and within $0 < x < 1$, the neonatal mortality rate intensity seems to be very high but declines more sharply for male. The probability density function of the distribution of deaths represents an important mortality statistics since it is an immediate indication of key longevity measures describing how long a population will live on the average and the extent of variability of ages at death. In Table 1 with respect to male, the curve of death function $l_x \mu_x$ describes the expected density of deaths at age x with respect to the population of lives surviving to age x and coincidentally for the male, there is a local minimum of $l_x \mu_x$ around 10 where mortality declines. The local

extreme points of $l_x \mu_x$ corresponds to points of inflexion of l_x following the observations below,

$$\frac{d}{dx}(l_x \times \mu_x) = -\frac{d}{dx}\left(l_x \times \frac{1}{l_x} \frac{d}{dx} l_x\right) = \frac{d}{dx}\left(-\frac{d}{dx} l_x\right) = -\frac{d^2}{dx^2} l_x \quad (92)$$

$$\frac{d}{dx} l_x \mu_x = \mu_x \frac{d}{dx} l_x + l_x \frac{d}{dx} \mu_x \quad (93)$$

Observing that $l'_x = -\mu_x l_x$, we have

$$\frac{d}{dx} l_x \mu_x = \mu_x (-\mu_x l_x) + l_x (\mu'_x) \quad (94)$$

The point at which l_x inflexes becomes

$$\frac{d}{dx} l_x \mu_x = \mu_x (-\mu_x l_x) + l_x (\mu'_x) = 0 \quad (95)$$

So that

$$\mu_x^2 = \frac{d}{dx} \mu_x \quad (96)$$

Since the functions $\{l_x, \mu_x, (l_x \mu_x)\}$ are functionally related, their behaviors are jointly examined through mortality surface and three dimensional plots as displayed in Fig. 3.

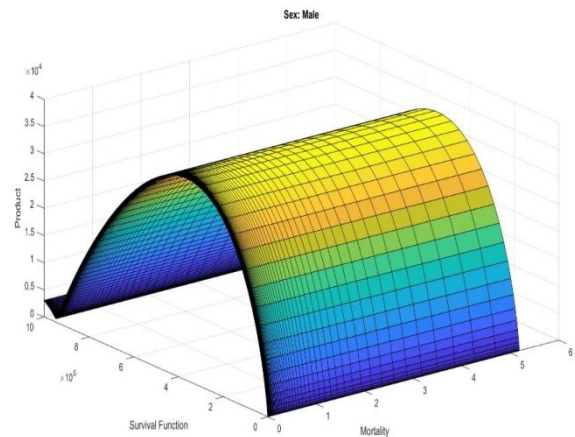


Figure 3: GM (2,2) surface plot for male's mortality, survival and curve of death functions

Notably, the function $\mu_x = GM(2,2) > 0$ is continuous for all $x \geq 0$ and satisfies the condition,

$$\int_0^{\infty} \mu_x dx = \infty \quad (97)$$

Although in Table 1, the probability of death satisfies $0 < q_x \leq 1$, the mortality rate intensity is increasing and $\mu_x > 1$ towards the end of the mortality table. The implication in Table 1 is that there is a mortality risk of having a higher rate of death than expected. The high rate of mortality (jumps) may have occurred as a result of the sudden occurrence of pandemics or war. This is because q_x is a probability whereas μ_x is a rate. Consequently at an infinitesimally small time Δ in

Table 1 during senescence, the mortality intensity is high and becomes,

$$\mu_x = -\frac{1}{l_x} \lim_{\Delta \rightarrow 0} \frac{l_{x+\Delta} - l_x}{\Delta} = \lim_{\Delta \rightarrow 0} \frac{l_x - l_{x+\Delta}}{\Delta} = \lim_{\Delta \rightarrow 0} \frac{\Delta q_x}{\Delta} \quad (98)$$

where $0 < \Delta < 1$ and consequently, the above argument explains clearly the mathematical distinction between the mortality rate intensity μ_x and the probability of death q_x .

For instance if $\Delta = \frac{1}{3}$, then $\mu_x = 3 \left(\frac{1}{3} q_x \right)$ and

therefore, $\mu_x > 1$

The highest age in the mortality table is given by $\Omega = \text{Sup} \left\{ \zeta \in \mathbf{R}^+ \mid F_{T_x}(\zeta) \leq 1 \right\}$. Consequently, it is numerically determined from the estimated l_x to be the first age where $q_x = 1$ whereas $\Omega_x(M) = 106$ for the males. To study the behaviour of mortality pattern beyond age 106, the male's intensity is extended to age 120 where the mortality rates at extreme age still exhibit exponential increase. This behaviour may not be valid in practice.

The omega age represents the age after which lives rarely survive because for all $x > \Omega$, $l_x = l_{x+1} = \dots = 0$ and consequently,

$$\lim_{x \rightarrow 0} q_x = \lim_{x \rightarrow 0} \left(\frac{l_x - l_{x+1}}{l_x} \right) = \lim_{x \rightarrow 0} (1 - p_x) = 0. \text{ But the}$$

probability of death $q_x = 1$. These two arguments explain clear inconsistency of mortality behaviour in death probability function at extreme ages q_x in Table 1. For the male in Table 1, the trajectories of l_x is observed to decline steeply at perinatality till infancy and exhibiting at least a point of inflexion to x axis as seen in Figure 1.

The argument that no life exists after age 106 is supported as follows (i) The set Ω containing the domain of mortality validity is assumed to have no limit point (ii) then we can show that Ω is closed. Ω is closed iff it contains all its limit points. Certainly by (i), Ω has no limit point by the hypothesis of the theorem, consequently, there are no limit point of Ω which are outside of Ω (which are not contained in Ω), so Ω contains all its limit points. Therefore Ω is closed.

In Table 1, for male and within the age intervals $0 \leq x \leq 31$, $81 \leq x \leq 120$, we observe that $\mu_x > q_x$ however, within the interval $33 \leq x \leq 79$, $\mu_x < q_x$. By reason of equation (15),

$$l_x q_x = \int_0^1 l_{x+\xi} \mu_{x+\xi} d\xi. \text{ If } l_{x+\xi} \mu_{x+\xi} \text{ were increasing,}$$

then $\frac{d}{d\xi} l_{x+\xi} \mu_{x+\xi} > 0$ and at the beginning of the

interval $0 < x < 1$, $l_x q_x > l_x \mu_x \Rightarrow q_x > \mu_x$ provided the curve $l_x \mu_x$ is increasing. By definition,

$$-\frac{d}{dx} l_x = l_x \mu_x, \text{ it then follows that when } l_x \mu_x \text{ is}$$

increasing, then clearly $\frac{d}{dx} l_x$ is decreasing and the

gradient of the tangent to the curve l_x will be decreasing. The survival function l_x will then be

concave to the age axis. Consequently, if the survival function l_x is concave to the age axis, the condition

$q_x > \mu_x$ is satisfied. However, if the survival function l_x is convex to the age axis $q_x < \mu_x$.

In Table 1 for male, $\mu_x \cong q_x$ almost at ages $\{32, 80\}$. The observation that $\mu_x \cong q_x$ in Table 1 occurs whenever

the survival function l_x is almost linear in the form

$$\frac{l_x - l_{x+\xi}}{l_x - l_{x+1}} = \frac{x + \xi - x}{x + 1 - x} = \xi \quad (99)$$

for $0 < \xi < 1$ and

$$l_{x+\xi} = (1 - \xi) l_x + \xi l_{x+1} \quad (100)$$

$$\mu_{x+\xi} = \frac{1}{l_{x+\xi}} \frac{dl_{x+\xi}}{d(x+\xi)} = \frac{1}{l_{x+\xi}} \frac{(l_{x+1} - l_x)}{d(x+\xi)} = \frac{(l_{x+1} - l_x)}{l_{x+\xi}} \times \frac{l_x}{l_{x+\xi}} \quad (101)$$

$$\mu_{x+\xi} = \frac{(l_{x+1} - l_x)}{l_x} \times \frac{1}{\frac{l_{x+\xi}}{l_x}} = \frac{(l_{x+1} - l_x)}{l_x} \times \frac{1}{\xi p_x} = \frac{q_x}{1 - \xi \times q_x} \quad (102)$$

Consequently at time $\xi = 0$, $q_x \cong \mu_x$.

The progressive increase in the mortality rate intensity for male is displayed in Fig. 1. Most life tables such as Commissioner's Standard Ordinary table (CSO) type and many published works do not account for μ_x based on the governing mortality intensities because of the computational intractability associated with their estimations. The male's ageing parameter our model is confirmed to satisfy the globally accepted ageing interval $1.08 < C < 1.12$. An interesting point in the modelling of $GM(2,2)$ is the novelty of the estimation technique through successive differencing applied. The technique gives better estimation for the ageing parameter than the maximum likelihood method (MLE) commonly used in most estimations.



The MLE technique does not often satisfy the globally accepted interval of validity and hence may be doomed in mortality analysis. When a life buys life insurance, the life is suspected to have a substandard health issues and hence such a life will have substandard mortality. As a result, life insurers will be interested in the functional relationship between standard and substandard mortality during the period of insurance. We assume here that the difference between standard and substandard mortality is the extra mortality risk which can be constant if $\Delta\mu_{x+s} - \Delta\mu_{x+s+1} = 0$, decreasing when $\Delta\mu_{x+s} - \Delta\mu_{x+s+1} > 0$ and increasing if $\Delta\mu_{x+s} - \Delta\mu_{x+s+1} < 0$

Conclusion

In this study, we have investigated different age dependent mortality functions to generate with higher precision mortality rate intensities and life insurance products. The rationale behind the use of successive differencing approach is to ensure that mortality rate estimation is technically precise and appropriate to the mortality risk evaluation. Secondly in measuring the mortality intensities, the central problem is not concerned about the choice of the appropriate functional form which seems very important to our problem areas but rather a combination of analytical functions and actuarial assumptions that are critical in creating the pay-off space for the life insurance claim contingencies and life annuity benefits together with developing analytically robust life table models in order to solve the estimation problem of the non-linear mortality intensities so as to produce good underwriting results. The different numerical estimation methods and assumptions adopted in this study evolve from varying underwriting experiences emanating from: (i) various forms of life insurance valuation and underwriting techniques, (ii) the extent of their complexities and (iii) the level to which actuary usually advises life insurers on the methods to be applied when carrying out actuarial valuations.

Finally, our results suggest that the underlying functions applied such as successive differencing is well suited to the estimation of non-linear parsimonious mortality functions which satisfies the Occam's razor. If life offices do not apply accurate mortality table based on correct estimations, they will be vulnerable to the risk of paying much more death benefits than expected. Consequently life offices are therefore advised to reserve for mortality risks to mitigate against being insolvent.

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