Renewal and stability in populations structured by remaining years of life

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Abstract

A unisex model of population renewal for populations structured by remaining years of life is presented in both continuous and discrete form. Various stable and transient properties populations are compared and related when viewed by chronological versus thanatological age. Results from the two age perspectives are conformable if the starting population is stable, but are otherwise divergent.

*This is a work in progress. All findings are preliminary at this time, so please don't cite without permission of the author.

Introduction

In this paper, I explore some formal demographic consequences of a particular redefinition of age. Instead of counting age as the time passed since birth, consider age as the amount of time left until death¹. The first kind of age is called chronological age, and the second kind of age is called thanatological age. Individuals in this case move in the same direction along imaginary life lines, but the reference point for thanatological age is at the end of the life line instead of at the beginning. For individuals, the timing of events in life is exactly identified using either of the two definitions of age, but aggregate patterns differ. Demography is built upon a small set of empirical regularities with respect to chronological age patterns. The models and methods of demography are in varying degrees prefaced on such regularity. There is evidence that many demographic phenomena are best described in the aggregatate as a function of time since birth, and others of time until death, while others can be a function of both to some degree (Riffe et al. 2015, Wolf et al.

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¹This quantity is referred to as residual lifetime in the reliability literaure, or just time-to-death (TTD) elsewhere.

2015). This observation ought to motivate the formal demographic base to incorporate this aspect of time in into models.

I describe the Lotka-Euler-Leslie renewal model when structured by thanatological age. The renewal model relates fertility, mortality, and population structure in a parsimonious form, revealing the long term consequences of a particular set of conditions. All diverse lines of contemporary demographic research are relatable to the renewal model, and by extention to stable population theory. To present the renewal model based on thanatological age is not necessarily a suggestion for a better way to model or project population. Rather, the thanatological renewal model represents an alternative and coherent way to describe the process of population renewal by relating renewal to a death flow rather than to a birth flow. I have described a limited case of this perspective on renewal for the case of stationary populations (Riffe 2015), and here provide a broader treatment allowing for growth in the population. The chronological-age classified renewal model and the thanatological renewal model both describe the movement of population stocks over time, but from opposite vantage points. Specifying a model of thanatological renewal expands the toolbox available to approach complex temporal population structure.

The next section of this article describes the lifetable method of transforming age-classified data into thanatological age classes. I then derive the continuous version of the thanatological renewal model, and some other stable properties. This is followed by the matrix version of the model, which is explained alongside the lifecycle graph.

A lifetable approximation of thanatological age

In general, to make a lifetable-based projective statement about the remaining time distribution of a current stock of population, it is best to to account for population subgroups with different mortality and to account for uncertainty about future changes in mortality. The present inquiry approximates than atological age using lifetable-projected remaining lifetime, which is intended to approximate remaining lifetime under the strong assumptions of homogenous populations und fixed vital rates. Under these assumptions, each birth cohort undergoes the same pattern of attrition. In a period setting, this means that we can distribute the remaining lifespans of the population in each age class using the same lifetable.

The basic transformation to than atological age proceeds by defining a conditional deaths density function, f(y|a). For chronological age zero, f(y|0) is identical to d(a), the lifetable deaths distribution (based on a unity radix). For ages higher than 0, simply condition on survival:²

$$f(y|a) = \mu(a+y)\frac{\ell(a+y)}{\ell(a)} \qquad . \tag{1}$$

²In the reliability literature, f(y|a) is denoted by X - t|X > t, where X is the total lifespan and t is the current chronological age attained.

A convenient discretization of (1) is to simply work with the d_a column of the lifetable (here with single ages):

$$f_{y,a} = \frac{d_{a+y} + d_{a+y+1}}{L_a + L_{a+1}} \qquad , (2)$$

where L_a is lifetable exposure.³ One produces in this way a probability density function over future death times for each chronological age, a. The f(y|a)- weighted average of y for a given a returns the familiar remaining life expectancy column of the lifetable, e(a). If P(a) is a chronological age-classified population count, we can derive P(y), a thanatological age-classified population count using⁴

$$P(y) = \int_0^\infty P(a)f(y|a) \, da \qquad . \tag{3}$$

Since $1 = \int_0^\infty f(y|a) \, \mathrm{d}y$ for each a, this density is useful for decomposing data classified by chronological age, such as the population pyramid, into thanatological age classes. Figure 1a shows this exercise for 2010 US data (Human Mortality Database), assuming the 2010 period mortality schedule remains constant. When looking from the chronological age perspective, this decomposition reveals thanatological heterogeneity, but when viewed from the thanatological perspective (Figure 1a) it reveals chronological age heterogeneity and the overall thanatological age profile, the future death flow of the current population. Figure 1b shows this recomposition for the same 2010 US data. The outer profile of Figure 1b is P(y), under our assumptions.

Brouard (1986) applied this method to highlight regional differences in French age structure, at times combining this projective method with historical stocks. Brouard (1989), and Vaupel (2009), showed that in stationary populations the marginal chronological and thanatological age structures are identical, and this provides a limited basis for comparison between the two perspectives. These results also follow from the following specification of a thanatological renewal model.

Lifetable-derived thanatological age structure is probabilistic until after death, since it refers to the future according to an assumed mortality pattern; in this case a period mortality pattern is used. As such, Figure 1b is projective in nature, a forward-looking glance at population attrition given the population stock and lifetable of a particular moment, whereas Figure 1a is reflective in nature, since a population's chronological age-structure is mostly the fruit of past fertility, but also to a lesser degree migration and attrition. Any count classified by chronological age can be reclassified in this way given a pertinent lifetable. To derive thanatological fertility rates under these assumptions, $\gamma(y)$, apply (3) to chronological age-classified birth counts, B(a), to get B(y) and again to exposure-to-risk, E(a), to calcualte E(y) and then divide in the usual way:

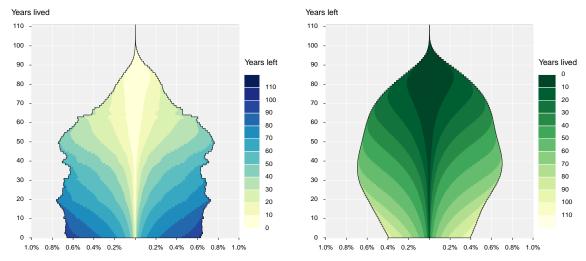
³Equation (2) will incur some error depending on the method used to derive L_a . To remove this error, rescale over the a margin so that each distribution conditioned on a sums to 1. Working with ℓ_a rather than L_a eliminates error and yields similar results, but is not necessarily better.

⁴An different-looking but identical definition is found in Brouard (1989).

Figure 1: 2010 US population structure

(a) Chronological age-structure (years lived) decomposed by thanatological age (years left).

(b) Thanatological age-structure (years left) decomposed by chronological age (years lived).



*Population and mortality data from the HMD.

$$\gamma(y) = \frac{B(y)}{E(y)} \qquad . \tag{4}$$

Figure 2b represents the full variety in female than atological period fertility rates that can be found for all years of data that overlap in the HFD and HMD.⁵ I show $\gamma(y)$ calculated in this way because it enters into the renewal model here studied. Figure 2a gives ASFR for the same populations and years as a more familiar reference, but note that both the x and y axes are different. The shape of than atological fertility curves is speculative and due to the above-mentioned strong assumptions. Nonetheless it is best not to include such rates into a model blindly, so we offer this schematic overview of some of their characteristics.

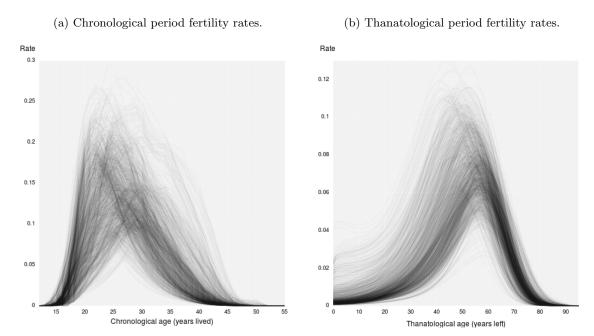
Period than atological fertility rates estimated in this way have a characteristic shape. The distribution is wider than that of chronological fertility rates. The range of post-childbearing lifespans stretches from zero, due e.g., to maternal mortality, to at least 100.⁶ This spaghetti plot includes both fertility booms and busts, as well as some mortality crises (1918, WWII). Some patterns in the data (not highlighted in the figure) include that the left tail (a gauge of how orphan-prone a population is) has tended to fall over time.

⁵There are as of this writing 1933 population-years of overlap between the HMD and HFD, including a wide variety of fertility and mortality combinations.

⁶For instance, Jeanne Calment lived for another 100 years after giving birth to her daughter.

All populations have shown an increase in both the mean and mode of thanatological age at childbearing over time. Several contemporary populations now show modes of over 60 years. Period thanatological total fertility rates track standard period total fertility rates closely, but tend to be somewhat higher (not shown).

Figure 2: Chronological and thantological fertility rates, all 1933 country-year combinations present in both the HFD and HMD.* Note different x and y scales.



*AUT, 1951 - 2010; BGR, 1947 - 2009; BLR, 1964 - 2012; CAN, 1921 - 2009; CHE, 1932 - 2011; CHL, 1992 - 2005; CZE, 1950 - 2011; DEUTE, 1956 - 2011; DEUTNP, 1990 - 2011; DEUTW, 1956 - 2011; ESP, 1922 - 2012; EST, 1959 - 2011; FIN, 1939 - 2009; FRATNP, 1946 - 2013; GBR_NIR, 1974 - 2011; GBR_NP, 1974 - 2011; GBR_SCO, 1945 - 2011; GBRTENW, 1938 - 2011; HUN, 1950 - 2009; IRL, 1955 - 2009; ISL, 1963 - 2010; JPN, 1947 - 2012; LTU, 1959 - 2011; NLD, 1950 - 2009; NOR, 1967 - 2009; PRT, 1940 - 2012; RUS, 1959 - 2010; SVK, 1950 - 2009; SVN, 1983 - 2009; SWE, 1891 - 2011; TWN, 1976 - 2010; UKR, 1959 - 2009; USA, 1933 - 2010

In general, than atological fertility rates will not be useful for purposes of projection, unless their empirical regularity is stronger than typical ASFR. This is a question I do not seek to answer in the present treatment, but in any case such rates are not susceptible to direct measurement. There are alternative ways of defining than atological fertility rates, for instance by transforming the birth and population vector from a stable population. As proven in appendix B, than atological fertility rates derived from a stable population imply a certain symmetry with stable populations from the chronological model. First I define the basic model.

The thanatological renewal model

The renewal model that follows uses than atological fertility rates, $\gamma(y)$, the form of which is suggested in Figure 2b, and we refer to a homogenous, closed, unisex population based on female vital rates. Assuming constant vital rates, the births for the present year are given by

$$B(t) = \int_0^\infty P(y, t)\gamma(y) \, \mathrm{d}y = \int_0^\infty P(a, t)m(a) \, \mathrm{d}a \qquad , \tag{5}$$

where a indexes chronological age, y indexes than atological age, P(a), P(y) are population counts, and $\gamma(y), m(a)$ are exact fertility probabilities (rates), best imagined as the fertility rates of females born to females. The than atological (first) integral can be broken down back in terms of chronological age:

$$= \int_{y=0}^{\infty} \int_{a=0}^{\infty} \gamma(y) P(a,t) f(y|a) \, da \, dy \qquad , \tag{6}$$

where f(y|a) is defined per equation (1), so $d(a)/\ell(0)$. We can relate the present population to past births with $P(a,t) = B(t-a)\ell(a)$:

$$= \int_{y=0}^{\infty} \int_{a=0}^{\infty} \gamma(y)B(t-a)d(a+y) \, \mathrm{d}a \, \mathrm{d}y \qquad . \tag{7}$$

Eventually strong ergodicity will assert itself, and B(t) will be related to B(t-a) according to a constant factor e^{ra} , where r is the familiar intrinsic rate of growth:

$$= \int_{u=0}^{\infty} \int_{a=0}^{\infty} \gamma(y)B(t)e^{-ra}d(a+y) \, da \, dy \qquad . \tag{8}$$

Divide out B(t) to get back a familiar-looking renewal equation:

$$1 = \int_{y=0}^{\infty} \int_{a=0}^{\infty} \gamma(y) d(a+y) e^{-ra} \, da \, dy \qquad . \tag{9}$$

Now compare this to Lotka's chronological formulation:

$$1 = \int_{a=0}^{\infty} m(a)l(a)e^{-ra} da \qquad , \tag{10}$$

which can be expressed as well as

$$= \int_{a=0}^{\infty} \int_{y=0}^{\infty} m(a)d(a+y)e^{-ra} dy da , \qquad (11)$$

and note that (11) and (9) are really quite similar, since $\int_{y=0}^{\infty} d(a+y) da = l(a)$. For intuition, notice that l(a) is here decomposed into d(a). Imagine a matrix of this decomposition, where one margin is chronological age and the other margin is thanatological age. For the chronological case (10), we multiply chronological age-specific fertility rates, m(a), over the chronological age margin, and for the thanatological case we multiply $\gamma(y)$ over the thanatological age margin.

There are gaps in the above line of development, since the jump from (7) to (8) (strong ergodicity) is assumed for the moment.⁷ The step has some intuition, due to the much greater density of connections within the model; persons from nearly any thanatological age can produce offspring that can have any other thanatological age.⁸ In this sense, the smoothing mechanism at play as the population passes through time must be much stronger than that for chronological age, at least in most cases. The mechanisms at play unfold in the same way as those so intuitively described by Arthur (1982), and said proof may apply here without further modification. Later in this paper I note how strong ergodicity is guaranteed given the properties of the thanatological projection matrix. A proof of the uniqueness of the solution to (9) is given in Appendix A. A proof that the chronological and thanatological renewal models must yield the same r if the starting population is already stable is given in Appendix B. A fast-converging iterative method for finding r from equation (9) is given in Appendix C

If the than atological fertility rates do not derive from a stable population, as will typically be the case, then the two age perspectives will, unless by coincidence, yield divergent stable models. There is a strong parallel here with the case of two-sex age-specific fertility rates. As in the case of divergence between male and female single-sex models, there will virtually always be divergence between single-sex models under chronological versus than atological age, unless the starting population is stable. Even though the modeled population stocks are in a way commensurable, the rates used here, f(a) versus $\gamma(y)$ are calculated on the basis of differently distributed denominators E(a) versus E(y).

Once one finds r from (9), other familiar stable population parameters can be calculated. For instance, we may calculate the mean than atological generation time, T^* :

$$T^* = \frac{\int_{y=0}^{\infty} \int_{a=0}^{\infty} y e^{-ra} d(a+y) \gamma(y) \, da \, dy}{\int_{y=0}^{\infty} \int_{a=0}^{\infty} e^{-ra} d(a+y) \gamma(y) \, da \, dy}$$
 (12)

Literally, this is the mean of the remaining lifespans of new mothers in the stable population. The thanato-

⁷The step to exponential stability is proven for the discrete version of the model.

⁸I say *nearly* because the very highest thanatological ages tend to be composed of pre-menarchical females, so these can be thought of as structural zeros.

logical net reproduction rate, R_0^{\star} is related by, e.g.,

$$R_0^{\star} = e^{rT^{\star}} \tag{13}$$

The birth rate, b, is given by

$$b = \frac{1}{\int_{u=0}^{\infty} \int_{a=0}^{\infty} e^{-ra} d(a+y) \, da \, dy} \qquad , \tag{14}$$

the denominator of which is the inside of equation (9) less $\gamma(y)$. This simplifies to

$$=\frac{1}{\int_{a=0}^{\infty} e^{-ra} l(a) \, \mathrm{d}a} \qquad , \tag{15}$$

which is exactly the same as that valid for chronological age in the Lotka-Euler setup, except that r here comes from the thanatological renewal model. If the stable age structure, $c^*(y)$, is known, then we may take the sum of hypothetical births over thanatological age:

$$= \int_{y=0}^{\infty} \gamma(y)c^{\star}(y) \, \mathrm{d}y \qquad . \tag{16}$$

The stable than atological age structure, $c^*(y)$, is the proportion of the stable population with remaining years to live y

$$c^{\star}(y) = b \int_{a=0}^{\infty} e^{-ra} d(a+y) da$$
 , (17)

or, given the stable chronological age structure, c(a), and a lifetable,

$$= \int_{a=0}^{\infty} c(a)f(y|a) \, \mathrm{d}a \tag{18}$$

$$= \frac{\int_{a=0}^{\infty} d(a+y)e^{-ra} da}{\int_{a=0}^{\infty} l(a)e^{-ra} da}$$
 (19)

It has previously been shown that when r=0 and y=a, that $c(a)=c^*(y)$ (Brouard 1989, Vaupel 2009), and now we can arrive at the same conclusion directly by equating the chronological and than tological renewal equations. For small non-zero magnitude values of r and survival patterns typical of humans, it is also the case that $c^*(y|r) \approx c(a|-r)$. This could make a useful heuristic, and the relationship is asymptotically exact as r approaches zero, but the exact relationship is to be found in equation (18). Further

stable population quantities may be estimated by similarly translating the various common definitions (e.g., in the glossary of Coale (1972)) to the present perspective. I focus on the main model rather than on these.

The thanatological projection matrix

These descriptions can be made more explicit, and in some ways more tractable, by hashing out the projection matrix that corresponds to the thanatological renewal model. As with the chronological age-structured Leslie matrix, the thanatological projection matrix, \mathbf{Y} , is square and of dimension $y \times y$, where y is the number of equal-interval thanatological age classes into which the population is structured. The matrix contains elements for survival and elements for fertility. Unlike Leslie matrices, \mathbf{Y} is not sparse, but is populated primarily with positive entries greater than zero. In the following, I illustrate using a 6×6 matrix.

Mortality occurs in only the population class with zero remaining years of life. The population of than a tological age 1 in year t moves to 0 in year t+1 and dies in the course of that year. Thus, rather than in the subdiagonal, survival elements are located in the superdiagonal. All survival values are 1, since there is no decrement until after living through age 0. The survival component of \mathbf{Y} is organized as in Matrix 0.1. To imagine a lifecycle graph, each age decrements by one through each lower age in successive order until being absorbed by death after completing than atological age zero.

Matrix 0.1: Survival component of unisex than atological projection matrix, Y

Fertility inputs to the matrix are derived from single-sex thanatological fertility and the lifetable d_a distribution, where a indexes chronological age and is equated to y remaining years of life for members just born. It is simpler to imagine d_a in this case as indexing the future lifespan distribution of newborns (Riffe 2015). Fertility in a thanatologically structured human population occurs in all but the very highest thanatological age classes, which can be thought of as structural zeros for simplicity, as they only contain pre-menarchical females that will have very long lives.

For our 6×6 example, say that fertility is observed in thanatological classes 0-4, while the final class has no fertility, where γ_y indicates the fertility probability for class y in the year t entering population (in the matrix columns). Each γ_y is then distributed over rows according to d_a . The fertility entry in row m and column n of \mathbf{Y} will therefore be $\gamma_n \cdot d_m$. One can assume that those dying over the course of year t (the first

column) are exposed to fertility for half of the year, 9 and so discount the fertility entry accordingly.

Further, infant mortality, $\gamma_y \cdot d_0$, located in the first row, must also be discounted, since part of the mortality will occur in the same year t and the rest in year t+1. The first row of fertility must be further discounted by a factor, λ , in order to account for the fact that infant mortality is higher in the lower Lexis triangle than in the upper. Of those infants who die in the first year of life, a proportion equal to λ do not survive to December 31st of the calendar year in which they were born.¹⁰ The fertility component of **Y** is then composed as in Matrix 0.2.

Matrix 0.2: Fertility component of unisex than atological projection matrix, Y

e_y	0_t	1_t	2_t	3_t	4_t	5_t
0_{t+1}	$\int (1-\lambda)\frac{\gamma_0 d_0}{2}$	$(1-\lambda)\gamma_1 d_0$	$(1-\lambda)\gamma_2 d_0$	$(1-\lambda)\gamma_3d_0$	$(1-\lambda)\gamma_4 d_0$	0
1_{t+1}	$rac{\gamma_0 d_1}{2}$	$\gamma_1 d_1$	$\gamma_2 d_1$	$\gamma_3 d_1$	$\gamma_4 d_1$	0
2_{t+1}	$\frac{\gamma_0 d_2}{2}$	$\gamma_1 d_2$	$\gamma_2 d_2$	$\gamma_3 d_2$	$\gamma_4 d_2$	0
3_{t+1}	$\frac{\gamma_0 d_3}{2}$	$\gamma_1 d_3$	$\gamma_2 d_3$	$\gamma_3 d_3$	$\gamma_4 d_3$	0
4_{t+1}	$rac{\gamma_0 d_4}{2}$	$\gamma_1 d_4$	$\gamma_2 d_4$	$\gamma_3 d_4$	$\gamma_4 d_4$	0
5_{t+1}	$\frac{\gamma_0 d_5}{2}$	$\gamma_1 d_5$	$\gamma_2 d_5$	$\gamma_3 d_5$	$\gamma_4 d_5$	0

The survival and fertility components of \mathbf{Y} add together elementwise, thus the full 6×6 matrix is composed as in Matrix 0.3.

Figure 3 depicts the projection matrix \mathbf{Y} in the form of a lifecycle graph. In this graph, each thanatological age is a node, and death is the absorbing state, denoted with an "X". Population members are born into any of the thanatological age classes, and then follow clockwise around the graph on the thick blue paths in order until finally being absorbed into death (the node labeled "X", not included in the projection matrix). The light gray paths represent fertility. These originate from any age class that has a fertility rate greater than zero, which for this toy matrix are all age classes except the 5th (highest). The \mathbf{y}^{th} age class is subject to the fertility rate $\gamma(y)$, and these births distribute out to all other age classes according to the lifespan distribution from the lifetable, d_a . Hence, the rate intensity of a fertility path is determined by the fertility rate of the origin age and probability of having a lifespan equal to the order of the destination node. A fraction of the fertility destined for infant mortality equal to λ leads straight to the absorbing state of death. Fertility paths from the 4th thanatological age class are highlighted. These correspond to the fifth column

 $^{^{9}}$ One might be tempted to not allow for fertility at all for females dying in year t, but recall that fertility is measured in the moment of birth, and not conception.

 $^{^{10}\}lambda$ can be derived directly from death counts data classified by Lexis triangles. If the demographer does not have information to derive λ directly, ad hoc or semidirect methods may be used to assign a reasonable proportion, such as 0.9 for contemporary low mortality populations.

Matrix 0.3: A full unisex than atological projection matrix, Y

$$\mathbf{Y} = \begin{bmatrix} e_y & 0_t & 1_t & 2_t & 3_t & 4_t & 5_t \\ 0_{t+1} & (1-\lambda)\frac{\gamma_0 d_0}{2} & (1-\lambda)\gamma_1 d_0 + 1 & (1-\lambda)\gamma_2 d_0 & (1-\lambda)\gamma_3 d_0 & (1-\lambda)\gamma_4 d_0 & 0 \\ 1_{t+1} & \frac{\gamma_0 d_1}{2} & \gamma_1 d_1 & \gamma_2 d_1 + 1 & \gamma_3 d_1 & \gamma_4 d_1 & 0 \\ 2_{t+1} & \frac{\gamma_0 d_2}{2} & \gamma_1 d_2 & \gamma_2 d_2 & \gamma_3 d_2 + 1 & \gamma_4 d_2 & 0 \\ 3_{t+1} & \frac{\gamma_0 d_3}{2} & \gamma_1 d_3 & \gamma_2 d_3 & \gamma_3 d_3 & \gamma_4 d_3 + 1 & 0 \\ 4_{t+1} & \frac{\gamma_0 d_4}{2} & \gamma_1 d_4 & \gamma_2 d_4 & \gamma_3 d_4 & \gamma_4 d_4 & 1 \\ 5_{t+1} & \frac{\gamma_0 d_5}{2} & \gamma_1 d_5 & \gamma_2 d_5 & \gamma_3 d_5 & \gamma_4 d_5 & 0 \end{bmatrix}$$

of the example matrix \mathbf{Y} , plus an additional path direct to death for the fraction of infant mortality prior to December 31st in year t. Paths that loop back into the same age class simply indicate those offspring that are destined to die in the same year as their mother. It is perhaps clearer to see from the density of this graph and by recalling that than atological fertility distributions are relatively wide (see Figure 2b) that there is a high degree of mixing of than atological age within populations via reproduction – a strong tendency toward stability.

Thanatological age-classes will ideally terminate at the highest value permitted by data. For the data used in this paper, there are 111 total age classes, which translate to 111 total remaining-years classes (0-110+). In practice \mathbf{Y} becomes a 111×111 matrix, with most entries non-zero (nearly complete connectivity, such as in Figure 3). Construction may appear tedious for this reason. However, note that the bulk of fertility entries can be derived as the outer (tensor) product $d_a \otimes f_y$, leaving only the first row and first column mortality discounting followed by the addition of the survival superdiagonal. In most statistical programming languages constructing \mathbf{Y} entails only a couple more lines of code than constructing a Leslie matrix.

Thanatological projection matrices may be manipulated using standard matrix techniques applicable to the Leslie matrix. Where \mathbf{p} is a population vector classified by thanatological age, projection proceeds by multiplying \mathbf{Y} from the left: $\mathbf{p}(t+1) = \mathbf{Y}\mathbf{p}(t)$.

For rate schedules typical of human populations, than atological projection matrices are filled with non-negative values, mostly greater than zero. Raising the matrix to some not-very-large power, k, will make all entries greater than zero. This observation, or else by noting that the lifecylce graph is strongly connected, lets us conclude that the matrix is irreducible and primitive. By the Perron-Frobenius theorem, the matrix will always have a unique, dominant, positive real eigenvalue, the natural log of which is the intrinsic growth rate, r, and strong ergodicity is assured. In other words, given a fixed matrix and enough time, the population will conform to some stable than atological age distribution. Strong ergodicity was taken for

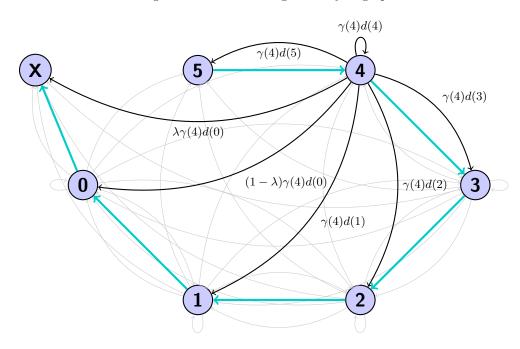


Figure 3: The thanatological lifecycle graph

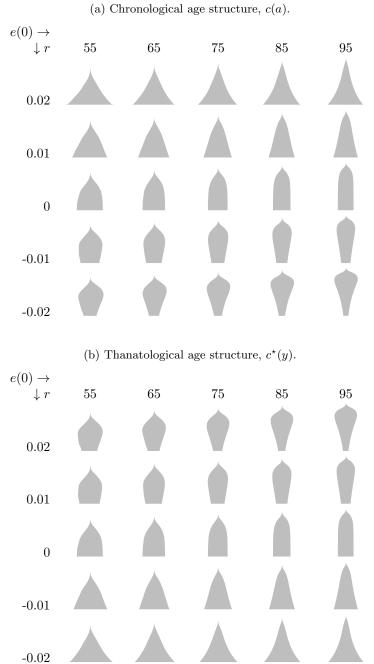
granted in jumping from equation (7) to (8), but the presently described properties of the matrix model should, albeit unrigorously, satisfy lingering uncertainty on that step. The eigenvector corresponding to the dominant eigenvalue gives this stable than atological age structure. If the fertility rates placed in the matrix are from the stable population, then the growth rate from the thanatological projection matrix is equal to that of the standard Leslie matrix. 11 The square of the ratio of the 1st to the 2nd eigenvalues is the damping ratio, an indicator of the springiness of the stable population structure, $c^{\star}(y)$, if perturbed from the state of stability.¹² A higher damping ratio means that the population structure oscillates back to its stable state faster, i.e., that oscillations decrease in size more rapidly. Thanatological damping ratios are usually higher than chronological damping ratios in the data used elsewhere in this paper.

Stable age structure

stable age structure is

¹¹Some care needs to be taken in order to demonstrate this point, since common approximations and adjustments used in matrix construction can throw things off. R code is available from the author to demonstrate this point. See also Appendix B for the corresponding proof from the continuous model. ¹²See Caswell (2001, p.86-87).

Figure 4: Stable age structures by life expectancy, e(0), and growth rate, r.*



*Underlying mortality pattern based on 2010 USA (HMD)

Discussion

The thanatological renewal model is valid, but this perspective may take some time to gestate before accepting that it is also a sound description of how populations renew. What should one imagine under the model of thanatological population renewal? A useful mnemonic bases itself on Figures 1a and 1b. In the chronological age-structured model, new generations appear at the bottom of the pyramid, and move up one rung per year. All age-classes are subject to attrition, which is spread out over ages and not readily visible in the pyramid. In the thanatological leaf, each birth cohort increments to the population over the whole range of thanatological age according to d(a), as seen in (9), becoming the shaded layers seen in Figure 1b. Each horizontal step is a death cohort, and these move one step down the pyramid each year without any decrement (indeed incrementing due to births) until reaching the very bottom. In short, the locations of increment and decrement, and the direction of movement (when so visualized) are all switched. The chronological and thanatological renewal models are almost perfectly opposite descriptions of the same process.

A Unique solution for thanatological r

This appendix contains a brief proof that the real solution for the intrinsic growth rate, r, is unique for the case of the thanatological renewal model, (9), and can be proven so in essentially the same fashion as those in existence for the Lotka-Euler model, (10). This proof follows that given in Pressat (1973). Define a convenience function, I(r), for the integrand of (9) for a given r and fixed $\gamma(y)$ and d(a):

$$I(r) = \int_{y=0}^{\infty} \int_{a=0}^{\infty} \gamma(y) d(a+y) e^{-ra} \, da \, dy$$
 (20)

Since the death distribution function, d(a), and fertility function, $\gamma(y)$, are continuous and non-negative, $\lim_{r\to+\infty} I(r) = 0$ and $\lim_{r\to-\infty} I(r) = \infty$. If $r_2 > r_1$, then $I(r_1) > I(r_2)$. I() is therefore a continuous and monotonically decreasing function of r with boundaries that include the value 1 of (9), and necessarily only obtain this value once. As with the Lotka-Euler equation, there will be more complex conjugate solutions for r in the thanatological model, and these are not explored in this paper.

B Equivalence in stability

This appendix contains a proof that the thanatological renewal equation implies the same intrinsic growth rate, r, as the Lotka-Euler model if the starting population is stable, given a particular set of thanatological fertility rates. This is identical to claiming that the right side of the renewal equation is equal given the same r. Since there is a unique real solution, demonstrated in the previous appendix, then it is sufficient to

show that the two integral equations are equal under these conditions. The basic relationship is:

$$\int_{a=0}^{\infty} \ell(a)e^{-ra}f(a) da = \int_{y=0}^{\infty} \int_{a=0}^{\infty} d(a+y)e^{-ra}\gamma(y) da dy$$
 (21)

when

$$\gamma(y) = \frac{b^{\star}(y)}{c^{\star}(y)} \tag{22}$$

$$b^*(y) = \int_{a=0}^{\infty} b(a)\mu(a+y) \frac{\ell(a+y)}{\ell(a)} da$$
 (23)

$$c^*(y) = \int_{a=0}^{\infty} c(a)\mu(a+y)\frac{\ell(a+y)}{\ell(a)} da$$
 (24)

Note that:

$$c(a) = \frac{\ell(a)e^{-ra}}{\int \ell(a)e^{-ra} \, \mathrm{d}a} \tag{25}$$

$$b(a) = f(a)c(a) \tag{26}$$

Replacing $\gamma(y)$ with the full expanded expression and plugging into the than atological renewal equation gives:

$$= \int_{y=0}^{\infty} \int_{a=0}^{\infty} d(a+y)e^{-ra} \frac{\int_{t=0}^{\infty} f(t)\ell(t)e^{-rt}\mu(a+y)\frac{\ell(t+y)}{\ell(t)} dt}{\int_{t=0}^{\infty} \ell(t)e^{-rt} dt} dt} \frac{\int_{t=0}^{\infty} f(t)e^{-rt}\mu(t+y)\frac{\ell(t+y)}{\ell(t)} dt}{\int_{t=0}^{\infty} \ell(t)e^{-rt} dt} dt} da dy$$
(27)

The denominator of c(a) cancels out:

$$= \int_{y=0}^{\infty} \int_{a=0}^{\infty} d(a+y)e^{-ra} \frac{\int_{t=0}^{\infty} f(t)\ell(t)e^{-rt}\mu(t+y)\frac{\ell(t+y)}{\ell(t)}}{\int_{t=0}^{\infty} \ell(t)e^{-rt}\mu(t+y)\frac{\ell(t+y)}{\ell(t)}} dt da dy$$
 (28)

Some l(a)'s also cancel out:

$$= \int_{y=0}^{\infty} \int_{a=0}^{\infty} d(a+y)e^{-ra} \frac{\int_{t=0}^{\infty} f(t)e^{-rt}\mu(t+y)\ell(t+y) dt}{\int_{t=0}^{\infty} e^{-rt}\mu(t+y)\ell(t+y) dt} da dy$$
 (29)

Now $d(a+y) = \mu(a+y)\ell(a+y)$, so we can cancel the inner integral with the denominator:

$$= \int_{u=0}^{\infty} \int_{a=0}^{\infty} f(a)e^{-ra}\mu(a+y)\ell(a+y) \,da \,dy$$
 (30)

Note also that $l(a) = \int_{=a}^{\infty} d(x) dx$ (present livings are future deaths), which brings us back to the chronological formulation:

$$= \int_{a=0}^{\infty} e^{-ra} \ell(a) f(a) \, \mathrm{d}a \tag{31}$$

This finding essentially proves that in the stable population there is at least one than atological fertility rate schedule that will satisfy the constraints of the lifespan distribution and r. However, it is not necessary to derive the than atological fertility distribution in the way prescribed here, and indeed an infinite number of such fertility distributions would satisfy the same constraints. Imagine these fertility rates cross-classified in both age dimensions, where two marginal sums give the than atological and chronological rates, as they would be calculated over the corresponding cross-classified exposures. The shape of this rate surface can be shifted around in infinitely many ways that would preserve a given growth rate and lifespan distribution, and the above proof shows one such mapping.

C An iterative method to find the thanatological r

Coale (1957) proposed a fast-converging iterative approach to estimate the intrinsic growth rate for the Lotka-Euler equation. For the thanatological renewal model, a similar approach may be taken, with some slight modifications to Coale's original. The following steps can be followed to estimate r from Equation (9):

1. Derive a first rough estimate of the mean remaining years of life at reproduction, \widehat{T}^* , akin to Lotka's mean generation time, T. To start, a good-enough guess is to just assume r = 0:¹³

$$\widehat{T}^{\star} = \frac{\int_{y=0}^{\infty} \int_{a=0}^{\infty} y d(a+y) \gamma(y) \, da \, dy}{\int_{y=0}^{\infty} \int_{a=0}^{\infty} d(a+y) \gamma(y) \, da \, dy}$$
(32)

2. A first rough guess at the net reproduction rate, R_0^{\star} is given by

$$R_0^{\star} = \int_{y=0}^{\infty} \int_{a=0}^{\infty} d(a+y)\gamma(y) \, \mathrm{d}a \, \mathrm{d}y$$
 (33)

¹³ Using the Brouard-Carey equality, the denominator could also be: $\int_{y=0}^{\infty} \ell(y) \gamma(y) dy$

3. A first rough estimate of r, r^0 , is given by

$$r^0 = \frac{\ln(R_0^{\star})}{\widehat{T}^{\star}} \tag{34}$$

- 4. Plug r^0 into Equation (9) to calculate a residual, δ^0 .
- 5. Use δ^0 and \widehat{T}^* to calibrate the estimate of r using

$$r^{1} = r^{0} + \frac{\delta^{0}}{\widehat{T}^{\star} - \frac{\delta^{0}}{r^{0}}} \tag{35}$$

6. Repeat step (4) to to derive a new δ^i , then step (5) to refine r^i , until converging on a stable r after some 20 or so iterations, depending on the degree of precision desired (\widehat{T}^* is not updated in this process).

The above procedure is more computationally efficient than minimizing the absolute residual of Equation (9) using a generic optimizer. Alternatively, one could use the method of cumulants to estimate r, but this would entail less precision than the above (unless one goes beyond, say, five cumulants), and this has not yet been worked out at this time.

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