Decomposition- Class 3

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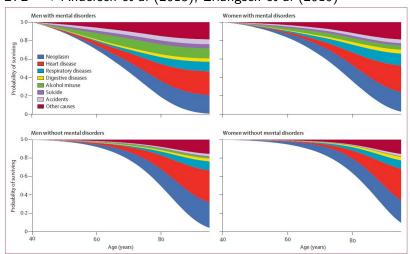


Kitagawa (1955)

$$\Delta CDR = \underbrace{\sum_{x} \left(\frac{M_{x}(t_{2}) + M_{x}(t_{1})}{2} \right) \left(\frac{N_{x}(t_{2})}{N(t_{2})} - \frac{N_{x}(t_{1})}{N(t_{1})} \right)}_{\text{Changes in x-composition}} + \underbrace{\sum_{x} \left(\frac{N_{x}(t_{2})}{N(t_{2})} + \frac{N_{x}(t_{1})}{N(t_{1})} \right) \left(M_{x}(t_{2}) - M_{x}(t_{1}) \right)}_{\text{Changes in rates}}$$
(1)

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LYL → Andersen et al (2013), Erlangsen et al (2016)



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Demography matters

Life expectancy at age x is defined as

$$e(x) = \frac{\int_{x}^{\infty} \ell(a) da}{\ell(x)}$$

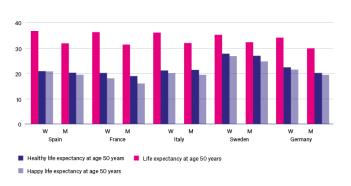
Then we can define disability-free life expectancy as

$$e^{DF}(x) = \frac{\int_x^{\infty} [1 - \pi(a)] \ell(a) da}{\ell(x)}$$

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Happy life expectancy, Solé-Auró et al (2018)

Graph 3. Life expectancy, healthy life expectancy and happy life expectancy.



Note 1: data on life expectancy and healthy life expectancy were obtained from Eurostat. Data on happy life expectancy were obtained from the SHARE database.

Note 2: W. women: M: men.

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Recap

Arriaga (1984) Effects of mortality change by age groups on life expectancies ($\sum_{n} \Delta_{x} = \text{Total change}$):

$${}_{n}\Delta_{x} = \underbrace{\frac{\ell_{x}^{1}}{\ell_{0}^{1}} \left(\frac{{}_{n}L_{x}^{2}}{\ell_{x}^{2}} - \frac{{}_{n}L_{x}^{1}}{\ell_{x}^{1}} \right)}_{\text{Direct effect}} + \underbrace{\frac{T_{x+n}^{2}}{\ell_{0}^{1}} \left(\frac{\ell_{x}^{1}}{\ell_{x}^{2}} - \frac{\ell_{x+n}^{1}}{\ell_{x+n}^{2}} \right)}_{\text{Indirect and interaction effects}}$$

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Funding

Acknowledgements

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Supplementary data

The impact of violence on Venezuelan life expectancy and lifespan inequality 3

Jenny García, José Manuel Aburto

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Abstract

Background

Venezuela is one of the most violent countries in the world. According to the United Nations, homicide rates in the country increased from 32.9 to 61.9 per 100 000 people between 2000 and 2014. This upsurge coincided with a slowdown in life expectancy improvements. We estimate mortality trends and quantify the impact of violence-related deaths and other causes of death on life expectancy and lifespan inequality in Venezuela.

Outline

- ► Cause-deleted life tables
- ► Measures of variation in ages at death
- ► Linear integral decomposition

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Cause-deleted life tables (based on Chapter 4, Preston et al 2001)

Often we are interested in knowing what a life table would look if only one cause of death were operating to diminish a cohort.

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'What would happen if...'

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► The rate of decrement from $\mu^i(x)$ if i were the only decrement $(*m^i(x))$ differs from what it would be if i were working in the presence of other decrements $(m^i(x))$.

$$\mu^{i}(a) = R_{i} \cdot \mu(a)$$
 for $x \leq a \leq x + n$

$$\mu^{i}(a) = R_{i} \cdot \mu(a) \quad \text{for } x \leq a \leq x + n$$
 (2)

► Since, by assupmtion,

$$_{n}^{*}p_{x}^{i}=e^{-\int_{x}^{x+n}\mu^{i}(a)da}=e^{-\int_{x}^{x+n}R_{i}\cdot\mu(a)da}$$

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$${}_{n}^{*}p_{x}^{i} = e^{-R_{i}\int_{x}^{x+n}\mu(a)da} = [e^{-\int_{x}^{x+n}\mu(a)da}]^{R_{i}}$$

$$\mu^{i}(a) = R_{i} \cdot \mu(a) \quad \text{for } x \leq a \leq x + n$$
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$$_{n}^{\ast}p_{x}^{i}=[_{n}p_{x}]^{R_{i}}$$

$$_{n}^{\ast }p_{x}^{i}=[_{n}p_{x}]^{R_{i}}$$

$$= [{}_{n}p_{x}]^{\frac{nD_{x}^{i}}{nD_{x}}}$$

Now with this simple relation we can create hypothetical scenarios.

https://life-expectancy.org

Lifespan variation indicators

Encopair variation malcators								
Indicators	•	Conventional life table notation						
Life disparity	e^{\dagger}	$\sum_{y=0}^{\omega} d_y e(y)$						
Gini coefficient	G	$1-rac{1}{e_0}\sum_{y=0}^{\omega}\ell_{y+1}^2$						
Theil's index	Т	$\sum_{y=0}^{\omega} d_y \left[rac{ar{x}_y}{e_0} \ln rac{ar{x}_y}{e_0} ight]$						
Mean logarithmic deviation	MLD	$\sum_{y=0}^{\omega} d_y (\ln(e_0/ar{x}_y))$						
Variance	V	$\sum_{y=0}^{\omega} d_y (\bar{x}_y - e_0)^2$						
Standard deviation	SD	\sqrt{V}						
Interquartile range	IQR	$\hat{x}_3 - \hat{x}_1$						

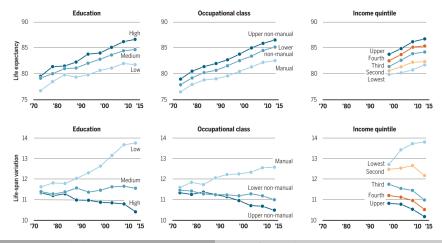
Table 1 Pearson correlation coefficients between pairs of indices, calculated from birth (ages 0–110+) in the top panel and calculated conditional on survival to age 10 (ages 10–110+) in the bottom panel, for all female and male life tables in the Human Mortality Database (7,516 in total)

	• • • • • • • • • • • • • • • • • • • •								
	e^{\dagger}	G	T	MLD	S	V	IQR		
e^{\dagger}	1.000								
G	.978	1.000							
T	.947	.991	1.000						
MLD	.965	.991	.992	1.000					
S	.981	.933	.893	.930	1.000				
V	.987	.945	.911	.944	.996	1.000			
IQR	.967	.966	.948	.956	.920	.944	1.000		
	e_{10}^{\dagger}	G_{10}	T_{10}	MLD_{10}	S_{10}	V_{10}	IQR_{10}		
e_{10}^{\dagger}	1.000								
G_{10}	.986	1.000							
T_{10}	.978	.995	1.000						
MLD_{10}	.979	.990	.995	1.000					
S_{10}	.986	.962	.961	.973	1.000				
V_{10}	.984	.964	.971	.980	.998	1.000			
IQR_{10}	.981	.978	.977	.976	.958	.966	1.000		

Source: van Raalte & Caswell (2013)

Trends in life expectancy and life-span variation for Finnish females, 1971–1975 to 2011–2014

Life expectancy is the average age at death, and life-span variation is the standard deviation, conditional upon survival to age 30, with age-specific death rates frozen at those observed in the given year. See supplementary materials for data and methods, including trends for males (which are qualitatively similar), and robustness checks using alternative measures of life-span variation.



Cause-deleted life tables
Lifespan variation measures
Linear integral decomposition
Replacement method

How can we decompose by age and cause of death these (any) function?

Linear integral decomposition (Horiuchi et al 2008)

▶ Relies on the assumption that values of the covariates change **continuously**, or gradually, along an actual or hypothetical dimension

Linear integral decomposition (Horiuchi et al 2008)

- ▶ Relies on the assumption that values of the covariates change **continuously**, or gradually, along an actual or hypothetical dimension
- ➤ Applied when difference in a dependent variable is expressed as a sum of the effects of differences in its covariates. (TFR,mean completed parity)

Let y be a demographic function (e.g. e^{\dagger}), which is differentiable, of n covariates (e.g. rates) denoted by $\mathbf{x} = [x_1, x_2, \dots, x_n]$.

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- Assume they both depend on an underlying dimension t (e.g. time).

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- ► Assume they both depend on an underlying dimension *t* (e.g. time).
- ▶ We have observations at two time points and x is a differentiable vector function of t between t_1 and t_2 .

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- Assume they both depend on an underlying dimension t (e.g. time).
- ▶ We have observations at two time points and x is a differentiable vector function of t between t_1 and t_2 .

$$y(t) = f(x(t)) = f(x_1(t), x_2(t), \dots, x_n(t))$$
 (3)

By the fundamental theorem of calculus

$$y(t_2) - y(t_1) = \int_{t_1}^{t_2} \frac{d}{dt} y(t) dt$$

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$$y(t_2) - y(t_1) = \int_{t_1}^{t_2} \frac{d}{dt} y(t) dt$$
 (4)

Applying the chain rule for partial derivatives of a composite function

$$y(t_2) - y(t_1) = \int_{t_1}^{t_2} \left[\sum_{i=1}^n \frac{\partial}{\partial x_i(t)} y(t) \cdot \frac{d}{dt} x_i(t) \right] dt$$
(5)

Exchange of integration and summation, and applying the substitution rule for definite integrals

$$y(t_2) - y(t_1) = \sum_{i=1}^{n} \int_{x_i(t_1)}^{x_i(t_2)} \frac{\partial}{\partial x_i(t)} y(t) dx_i(t)$$

Exchange of integration and summation, and applying the substitution rule for definite integrals

$$y(t_2) - y(t_1) = \sum_{i=1}^{n} \int_{x_i(t_1)}^{x_i(t_2)} \frac{\partial}{\partial x_i(t)} y(t) dx_i(t)$$
 (6)

(Simplifying notation)

$$y_2 - y_1 = \sum_{i=1}^n \int_{x_{i1}}^{x_{i2}} \frac{\partial y}{\partial x_{i1}} dx_i = \sum_{i=1}^n c_i$$
 (7)

$$y_2 - y_1 = \sum_{i=1}^n c_i \tag{8}$$

 c_i is the total change in y produced by changes in the i-th covariate, x_i .

Important: Theoretical foundation for decomp analysis: implies that even if a dependent variable is not an additive function of its covariates, a change in the dependent variable can be expressed as a sum of effects of the covariates.

Stepwise decomposition (Andreev et al 2002)

- ➤ To decompose a function for which there is no analytic solution (or for which the analytic solution is complicated).
- ➤ Step-wise decomposition alters the rates one element at a time (the element usually age), and recalculates the index function (i.e. life expectancy or any other summary measure).

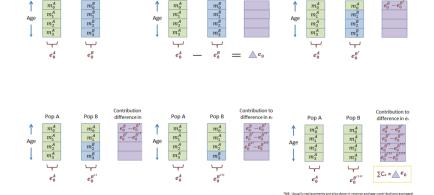
Pop A

Stepwise decomposition

Step-wise decomposition method

Pop B

Pop A



Pop B

Contribution to

difference in e

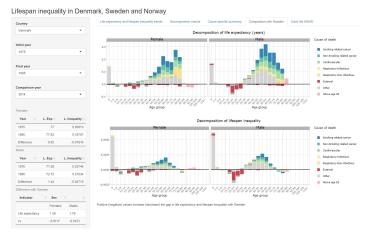
Pop A

Pop B

Contribution to

difference in e

Extending to causes of death



https://jmaburto.shinyapps.io/DK_App/

Outline Cause-deleted life tables Lifespan variation measures Linear integral decomposition Replacement method

Challenge 4

Use the linear integral model to decompose the change in the standard deviation of the age-at-death distribution and life expectancy by age and cause of death for 3 countries you might be interested in (over time or between them). Interpret the results of life expectancy alongside standard deviation. Make it interesting. You can use data from HCoD, HMD, WHO, GBD.

Exercises 1 and 2