

University of Groningen

Impact of different mortality forecasting methods and explicit assumptions on projected future life expectancy

Stoeldraijer, L.; van Duin, C.; van Wissen, Leo J G ; Janssen, Fanny

Published in:
Demographic Research

DOI:
[10.4054/DemRes.2013.29.13](https://doi.org/10.4054/DemRes.2013.29.13)

IMPORTANT NOTE: You are advised to consult the publisher's version (publisher's PDF) if you wish to cite from it. Please check the document version below.

Document Version
Publisher's PDF, also known as Version of record

Publication date:
2013

[Link to publication in University of Groningen/UMCG research database](#)

Citation for published version (APA):

Stoeldraijer, L., van Duin, C., van Wissen, L. J. G., & Janssen, F. (2013). Impact of different mortality forecasting methods and explicit assumptions on projected future life expectancy: The case of the Netherlands. *Demographic Research*, 29, 323-354. [13]. DOI: 10.4054/DemRes.2013.29.13

Copyright

Other than for strictly personal use, it is not permitted to download or to forward/distribute the text or part of it without the consent of the author(s) and/or copyright holder(s), unless the work is under an open content license (like Creative Commons).

Take-down policy

If you believe that this document breaches copyright please contact us providing details, and we will remove access to the work immediately and investigate your claim.

Downloaded from the University of Groningen/UMCG research database (Pure): <http://www.rug.nl/research/portal>. For technical reasons the number of authors shown on this cover page is limited to 10 maximum.



DEMOGRAPHIC RESEARCH

A peer-reviewed, open-access journal of population sciences

DEMOGRAPHIC RESEARCH

VOLUME 29, ARTICLE 13, PAGES 323-354

PUBLISHED 27 AUGUST 2013

<http://www.demographic-research.org/Volumes/Vol29/13/>

DOI: 10.4054/DemRes.2013.29.13

Research Article

Impact of different mortality forecasting methods and explicit assumptions on projected future life expectancy: The case of the Netherlands

Lenny Stoeldraijer

Coen van Duin

Leo van Wissen

Fanny Janssen

© 2013 Stoeldraijer, van Duin, van Wissen & Janssen.

*This open-access work is published under the terms of the Creative Commons Attribution NonCommercial License 2.0 Germany, which permits use, reproduction & distribution in any medium for non-commercial purposes, provided the original author(s) and source are given credit.
See <http://creativecommons.org/licenses/by-nc/2.0/de/>*

Table of Contents

1	Introduction	324
2	Data and methods	326
2.1	Methodology	326
2.2	Data	328
2.3	The models in more detail	328
3	Forecasting methods in Europe	330
4	Forecasts for the Netherlands	332
5	Results of different methods for Dutch mortality	335
6	Discussion	339
6.1	Summary of the results	339
6.2	Reflection on the differences in the outcomes	339
6.3	Reflection on the explicit assumptions used	340
6.4	Explanation of the observed projection outcomes	342
6.5	Forecasting considerations	344
7	Overall conclusion	346
	References	347

Impact of different mortality forecasting methods and explicit assumptions on projected future life expectancy: The case of the Netherlands

Lenny Stoeldraijer^{1,3}

Coen van Duin¹

Leo van Wissen^{2,3}

Fanny Janssen^{3,4}

Abstract

BACKGROUND

With the rapid aging of the population, mortality forecasting becomes increasingly important, especially for the insurance and pension industries. However, a wide variety of projection methods are in use, both between and within countries, that produce different outcomes.

OBJECTIVE

We review the different mortality forecasting methods and their assumptions in Europe, and assess their impact on projections of future life expectancy for the Netherlands.

METHODS

For the Netherlands, we assess the projections of life expectancy at birth (e0) and at age 65 (e65) up to 2050 resulting from different methods using similar explicit assumptions regarding the historical period and the jump-off rates. We compare direct linear extrapolation, the Lee-Carter model, the Li-Lee model, a cohort model, separate projections of smoking- and non-smoking-related mortality, and the official forecast.

RESULTS

In predicting mortality, statistical offices in Europe mostly use simple linear extrapolation methods. Countries with less linear trends employ other approaches or different assumptions. The approaches used in the Netherlands include explanatory

¹ Statistics Netherlands, The Hague, the Netherlands.

² Netherlands Interdisciplinary Demographic Institute (NIDI), The Hague, the Netherlands.

³ Population Research Centre, Faculty of Spatial Sciences, University of Groningen, the Netherlands.

⁴ Unit of PharmacoEpidemiology & PharmacoEconomics (PE2), Department of Pharmacy, University of Groningen, the Netherlands.

models, the separate projection of smoking- and non-smoking-related mortality, and the projection of the age profile of mortality. There are clear differences in the explicit assumptions used, including assumptions regarding the historical period. The resulting e_0 in 2050 varies by approximately six years. Using the same historical period (1970-2009) and the observed jump-off rates, the findings generated by different methods result in a range of 2.1 years for women and of 1.8 years for men. For e_{65} , the range is 1.4 and 1.9 years, respectively.

CONCLUSIONS

As the choice of the explicit assumptions proved to be more important than the choice of the forecasting method, the assumptions should be carefully considered when forecasting mortality.

1. Introduction

With the rapid aging of the population, mortality forecasts have become more important. Recent reforms in the pension systems in Europe—which were necessary to ensure that pensions remain sustainable—have made the link between pensions and changes in life expectancy more apparent than ever. In general, monthly pension payments are based on remaining life expectancy when people retire. But whereas in some countries benefit levels are linked to life expectancy (Germany, Finland, and Portugal), in others the pension age is set to rise with increasing life expectancy (Denmark, the Netherlands), or the contribution period for pensions is set to be extended as people live longer (France) (OECD 2007). The accurate modelling and projection of mortality rates and life expectancy are therefore of growing interest to researchers.

As mortality forecasts have become increasingly important, numerous models for mortality modelling and forecasting have been developed (for reviews see Pollard 1987; Tabeau 2001; Wong-Fillipp and Haberman 2004; Booth and Tickle 2008). The various methods for mortality forecasting can be divided into three approaches: extrapolation, explanation, and expectation (Booth and Tickle 2008). Extrapolative methods make use of the regularity typically found in both age patterns and trends in time. The explanation approach makes use of structural or epidemiological models of mortality from certain causes of death for which the key exogenous variables are known and can be measured. The expectation approach is based on the subjective opinions of experts involving varying degrees of formality. It should be noted that some mortality forecasting methods include aspects of one or more approaches.

In the past, most methods were relatively simple and were largely based on subjectivity (Pollard 1987). Over time, however, more sophisticated methods that make increasing use of standard statistical methods have been developed and applied (Booth and Tickle 2008). The majority of these methods can be classified as extrapolative approaches, of which the Lee-Carter method has become dominant. This method summarises mortality by age and period for a single population as an overall time trend, an age component, and the extent of change over time by age (Lee and Carter 1992).

One of the strengths of the Lee-Carter method and of extrapolation methods in general is their robustness in situations in which age-specific log mortality rates have linear trends (Booth et al. 2006). However, some countries have less linear trends (e.g., Booth, Maindonald, and Smith 2002 for Australia; Renshaw and Haberman 2006 for England and Wales; Janssen, Kunst, and Mackenbach 2007 for the Netherlands). It is therefore important to debate whether merely “objective” linear extrapolation methods should be employed, despite the non-linearity in the trends, or whether adding information—e.g., by including a cohort effect or trends in other countries, or by using more explanatory models—is preferable, despite the subjectivity this would involve.

One example of a method which includes additional information is coherent forecasting (Li and Lee 2005). This extension of the Lee-Carter model seeks to ensure that the forecasts for related populations maintain certain structural relationships based on commonalities in their historical trends; for example, that forecasts for similar countries are not radically different. The Lee-Carter method has also recently been extended to include a cohort dimension (Renshaw and Haberman 2006), and other stochastic models have been introduced to integrate the cohort dimension in mortality forecasting (see Cairns et al. 2011). Other examples are forecasting methods using valuable medical knowledge and information on behavioural and environmental changes, such as smoking and/or obesity (e.g. Pampel 2005; Olshansky et al. 2005; Bongaarts 2006; Janssen and Kunst 2007; Stewart, Cutler, and Rosen 2009; Wang and Preston 2009; King and Soneji 2011; Janssen, van Wissen, and Kunst 2013). Although these new types of methods have many advantages, the more explanatory methods involve a large element of subjective judgment (see Section 6.5 for a further discussion). Furthermore, as all of the above-mentioned methods are fairly new, the number of times they have been applied is still relatively small.

The advent of new methods has led to a variety of types of methods being used to produce projections within a single country (e.g., Wong-Fillipow and Haberman 2004), which have produced different forecast outcomes. Most existing studies that have compared the outcomes of different methods have focused predominantly on variants within one model, such as the Lee-Carter model and its variants, extensions, or generalisations. These include Booth et al. (2002, 2005, 2006); Li and Lee (2005); Renshaw and Haberman (2006); Hyndman and Ullah (2007); Wang and Liu (2010);

and Shang, Booth, and Hyndman (2011). Other studies (CMI 2005, 2006, 2007; Cairns et al. 2011) have compared the Lee-Carter model (and its cohort extension) with other extrapolative statistical models, such as P-splines models and the statistical model CBD (see Cairns, Blake and Dowd, 2006). These studies showed that using different assumptions leads to different outcomes, and that comparing different variants and extensions does not automatically result in the identification of a single best method. The historical period used is the main determinant of large differences in outcomes (see also Janssen and Kunst 2007), especially when there is considerable non-linearity in the trends.

Comparison studies of different types of methods, including both extrapolation and explanatory approaches, are not often undertaken. The comparison of outcomes from different studies is hampered by differences in the explicit assumptions; i.e., in the specific choices that must be explicitly stated in a method, such as the choice of the length of the historical period and of the jump-off rates. It would be helpful to examine, however, whether differences in projection outcomes within a country are caused by different methods, or by the use of the explicit assumptions.

The purpose of this study is to review the different mortality forecasting methods and their assumptions in Europe, and to assess their impact on projections of future life expectancy for the Netherlands.

More specifically, (i) we will review the current methods used in official mortality forecasts in Europe; (ii) compare the outcomes and the assumptions of different projection methods within the Netherlands; and (iii) compare the outcomes of different types of methods for the Netherlands using similar explicit assumptions, including the same historical period.

2. Data and methods

2.1 Methodology

After first reviewing the current methods for forecasting mortality used by statistical offices in Europe and the different national and international forecasts/projections that exist for the Netherlands, we will show to what extent different methods applied to Dutch data for the period 1970-2009 lead to different future values of life expectancy up to 2050. For the latter, we look at two outcome measures: life expectancy at birth and, in light of pension reforms, life expectancy at age 65. Moreover, we limit our own calculations to methods based on extrapolating the trends in age-specific death rates.

We found mortality forecasting methods used by statistical offices in the Netherlands and Europe in publications, including online publications, up to 2011 by

using the following search words: “mortality forecasting,” “forecasting,” “mortality projection,” “population projection,” and “projection.” Information on the methods is given for Austria, Belgium, Denmark, France, Italy, Ireland, Luxemburg, the Netherlands, Norway, Poland, Portugal, Spain, Sweden, and the UK. Mortality forecasts for the Netherlands are published by Statistics Netherlands, the Actuarial Society, RIVM (National Institute for Public Health and the Environment), Eurostat (EUROPOP2010), and by four research projects: TOPALS (De Beer 2012); UPE (Alders et al. 2007); Janssen, van Wissen, and Kunst (2013); and the European Demographic Datasheet (VID/IASA/PRB 2012).

The six different methods we applied to the Dutch data for 1970-2009, separately for both sexes, are:

- direct linear extrapolation;
- the Lee-Carter model (Lee and Carter 1992);
- an extension of the Lee-Carter model that includes the mortality experiences of other countries (Li and Lee 2005);
- an extension of the Lee-Carter model that includes a cohort dimension (Renshaw and Haberman 2006);
- a model in which smoking-related and non-smoking-related mortality is projected separately (Janssen and Kunst 2010; Janssen, van Wissen, and Kunst 2013); and
- the method used between 2004 and 2010 in the official forecast by Statistics Netherlands.

Our review showed that these methods—which represent fundamentally different approaches—were among the mortality forecasting methods used most frequently by the statistical offices in Europe, including in the Netherlands. We have chosen to avoid explicitly applying an expectation approach method because of the high degree of dependence on expert opinion in setting the target; e.g., every outcome can be set. We have also decided to ignore other methods that do not specifically extrapolate trends in age-specific death rates. See Section 2.3 for a more detailed description of the applied methods.

We chose data for 1970-2009 because of the data requirements of the method used by Statistics Netherlands. In addition to using a fixed historical period, we will use the observed values for 2009 as the jump-off rates for all the methods. Whenever possible, the further specifications and assumptions within each framework are also kept the same.

2.2 Data

Data on all-cause mortality and population numbers by sex, age (0, 1-4, 5-9, ..., 90-94, 95+), and year (1970-2009) were obtained from Statistics Netherlands. For the Li-Lee model, the same data were also obtained for Denmark, England and Wales, Finland, France, Italy, Norway, Spain, Sweden, Switzerland, and West Germany from the Human Mortality Database. Lung cancer mortality data and cause-specific mortality data were obtained from Statistics Netherlands for the separate projection of smoking- and non-smoking-related mortality and for the official forecast, respectively.

2.3 The models in more detail

The direct linear extrapolation model is given by

$$\ln(m_{x,t}) = a_x + b_x t + \varepsilon_{x,t} \quad (1)$$

where $m_{x,t}$ denotes the central mortality rate at age x and year t , a_x indicates the constant age pattern, b_x denotes the set of age-specific constants that describe relative rate of change at any age, and $\varepsilon_{x,t}$ denotes the residual error.

The Lee-Carter model (Lee and Carter 1992) is given by

$$\ln(m_{x,t}) = a_x + b_x k_t + \varepsilon_{x,t} \quad (2)$$

where k_t denotes the underlying time development. a_x is set equal to the average over time of $\ln(m_{x,t})$. b_x and k_t are found using Singular Value Decomposition under the assumptions $\sum_x b_x = 1$ and $\sum_t k_t = 0$. After estimation, we adjusted k_t to fit the observed life expectancy (Lee and Miller 2001) and extrapolated k_t using a random walk with drift.

The Li-Lee method is an extension of the Lee-Carter method, which takes into account the mortality experiences of other populations, such as countries (Li and Lee 2005). Short-term differences in mortality are preserved, but in the long term, the age-specific death rates within the group of countries are limited to a constant ratio to one another. In essence, the Lee-Carter method is applied twice: first to all countries combined ($\ln(M_{x,t}) = A_x + B_k K_t + E_{x,t}$), and then to the residuals ($\ln(\hat{M}_{x,t}) - \ln(m_{x,t,i}) = b_{x,i}^{res} k_{t,i}^{res} + \varepsilon_{x,t,i}^{res}$). $M_{x,t}$ denotes the central mortality rate (at age x and year t) of all countries combined, and $m_{x,t,i}$ indicates the central mortality rate of country i . The estimates are combined into one model for the country concerned:

$$\ln(m_{x,t,i}) = a_{x,i} + B_x K_t + b_{x,i}^{res} k_{t,i}^{res} + \varepsilon_{x,t,i} \quad (3)$$

where $a_{x,i}$ equals the average over time of $\ln(m_{x,t,i})$. The time parameter of the residual ($k_{t,i}^{res}$) is extrapolated using an autoregressive model (AR(1)). The other specifications are the same as in the Lee-Carter method described above. The mortality experiences of the 10 low-mortality countries surrounding the Netherlands are taken as the experiences of the other populations; i.e., Denmark, England and Wales, Finland, France, Norway, Spain, Sweden, Switzerland, and West Germany.

The cohort model represents an extended version of the Lee-Carter model with an extra parameter γ_{t-x} which denotes the underlying cohort effect that is a function of the year of birth $t - x$ (Renshaw and Haberman 2006). The model is given by

$$\ln(m_{x,t}) = a_x + b_x k_t + c_x \gamma_{t-x} + \varepsilon_{x,t} \quad (4)$$

Thus, this model includes the age effect, the period effect, and the cohort effect. Because of the inclusion of cohorts, age groups zero and 1-4 are combined so that the age groups are of equal length. Furthermore, cohorts with fewer than four observation years are not taken into account. The parameters are estimated by an iterative process (by maximum likelihood) using the constraints $\sum_x b_x = 1$, $\sum_x c_x = 1$, $\sum_t k_t = 0$ and $\sum_x \gamma_{t-x} = 0$. After estimation, k_t is adjusted to fit the observed life expectancy and is extrapolated using a random walk with drift. To avoid unrealistic future mortality patterns, gamma (γ_{t-x}) is set to zero for the cohorts outside the data. We have chosen zero because no clear trend for the cohort parameter was shown in the Dutch data, and its average over the whole period was close to zero. Because of this constraint, caution is warranted when interpreting the outcomes of this cohort model.

The model which projects non-smoking-related mortality and smoking-related mortality separately is referred to as the “smoking+non-smoking” model. First, non-smoking-related mortality—i.e., mortality after the exclusion of deaths caused by smoking—is calculated using etiological fractions. The etiological fractions are the age- and sex-specific proportions of total mortality determined by smoking. They are estimated by an adaptation of the indirect Peto-Lopez method (see Janssen and Kunst (2010) and Janssen, van Wissen, and Kunst (2013) for more details). Non-smoking-related mortality is projected using the Lee-Carter method. The assumed future etiological fractions and the projected non-smoking-related mortality combined give the projected total mortality. The future etiological fractions are taken from Janssen, van Wissen, and Kunst (2013).

Statistics Netherlands publishes a mortality forecast as part of its official population forecast for the Netherlands (Van Duin et al. 2011). Overall mortality is forecast using decomposition by cause of death (Van Duin et al. 2011; De Jong and Van

Der Meulen 2005). The probability of dying from a specific cause-of-death group in a specific age interval is projected for selected sample years. The techniques used are extrapolation combined with expert opinion, but epidemiological information (smoking) is also used. The all-cause probability of surviving is calculated by multiplying the probability of surviving for each cause of death. Intermediate years are calculated by interpolation. The Brass-logit method (Brass 1971) is used to calculate age-specific probabilities from the probabilities per age interval. The smoothed age-specific probabilities of dying of the last two observed years are used as a model curve.

3. Forecasting methods in Europe

The approaches currently used by statistical offices in Europe to project future mortality vary considerably (see Table 1). Extrapolation methods are used most frequently. These methods are either a direct linear extrapolation of the logarithm of the age-specific mortality rates (Austria, Belgium, France, and Spain), or a variant of the Lee-Carter model (Denmark, Italy, Norway, Portugal, and Sweden). Ireland, Luxembourg, Poland, and the UK use a more subjective target approach. For Poland and Luxembourg, information on trends in other countries is included directly in the projection. Statistics Netherlands projects cause-specific mortality using extrapolation techniques combined with expert opinion (Van Duin et al. 2011; De Jong and Van Der Meulen 2005). This is the only national statistical office in our selection that includes epidemiological information directly in its projection. Portugal, France, Ireland, and the UK also include expert opinion in their mortality forecasts. Indirectly, through the knowledge of the experts, this could include trends in other countries and epidemiological information.

In addition to the differences in the methods used, there are also differences in the variants and the extensions employed. Denmark, Italy, Portugal, and Sweden use different variants of the original Lee-Carter method. Norway and Denmark extend the original method. Belgium and Spain extend the direct extrapolation method with a re-estimation after smoothing the age-specific parameter, but use a different period for the re-estimation. Belgium and France both make some adjustments for old-age mortality. Ireland and the UK make a similar assumption about the target value; namely, a constant improvement rate after some year in the future. The UK includes a cohort approach for the convergence because of the apparent cohort effects in that country.

Moreover, the historical period used differs considerably by country. Ireland, Norway, and the UK use long periods (82, 109, and 109 years, respectively), whereas Denmark, France, Spain, and Sweden use relatively short periods (20, 15, 17, and 13 years, respectively). There is also variation in the length of the forecasted period,

ranging from 28 (Poland) to 91 years (Denmark), although this does not seem to correlate with the length of the historical period.

It appears that the observed past trends determine which method and historical period is used. Life expectancy at birth in western Europe has increased by six to 10 years since 1970 (WHO Health Database, Human Mortality Database). All of the countries in western Europe have experienced a rise in life expectancy, albeit at different rates and with periods of stagnation. Countries with a more linear trend (e.g., France and Switzerland) use extrapolation methods with an average historical period, while countries with more non-linear trends (e.g., Denmark, the Netherlands, and Norway) use different approaches in order to take non-linearity into account. Denmark, which has a history of having a less linear trend among women in particular, uses an extrapolation method, but with a short historical period. The Netherlands, which has non-linear trends among both men and women, uses epidemiological information in the forecast. Norway, with a period of stagnation in the 1980s among men, uses a very long period, but includes a quadratic age effect to account for the non-linearity.

Table 1: Methods and assumptions behind mortality forecasting methods of statistical offices in Europe

Country	Reference	Type of method	Assumptions	Historical period	Forecasted period
Austria	Hanika (2010)	Direct extrapolation	Extension: using more recent data for the short-term trend and convergence to a plausible function of age and sex for the long term (Ediev 2008)	1970-2008	2010-2050
Belgium	Bureau fédéral du Plan (2009)	Direct extrapolation	Extension: old-age adjustment; Extension: 10-year period for re-estimation after smoothing the age-specific parameter	1970-2007	1990-2060
Denmark	Hansen and Stephensen (2010)	Lee-Carter	Variant: $k(t)$ is adjusted to fit the observed life expectancy (Lee and Miller 2001); Extension: short-term correction to account for the error between estimated jump-off rate and observation (Bell 1997); Extension: smoothing mechanism (De Jong and Tickle 2006),	1990-2009	2010-2100
France	Blanpain and Chardon (2006)	Direct extrapolation, Expert opinion	Ages 3-13 no improvement after 2040; Extension: old-age adjustment	1988-2002	2007-2060
Italy	Salvini, Santini, and Vignoli (2006)	Lee-Carter	Variant: an autoregressive time-series model with a deterministic time trend	Unknown	2001-2051
Ireland	Central Statistics Office (2008)	Target value, Expert opinion	Long-term rate of 1.5% per annum from 2031 onwards; Extension: linear interpolation between mortality declines in 2005 and 2031	1926-2005	2011-2041

Table 1: (Continued)

Country	Reference	Type of method	Assumptions	Historical period	Forecasted period
Luxembourg	STATEC (2005)	Target value	Assumptions of Eurostat (convergence in 2100 for all EU countries; BMS method (Booth et al. 2002) for total group)	1962-2005	2005-2055
Netherlands	Van Duin et al. (2011)	Cause of death, Direct extrapolation, Lee-Carter, Expert opinion	Different assumptions per cause of death, depending on historical trend	1970-2009	2010-2060
Norway	Keilman and Pham (2005), Brunborg and Texmon (2010)	Lee-Carter	Extension: a quadratic age effect	1900-2008	2010-2060
Poland	Główny Urząd Statystyczny (2009)	Target value	"Catch-up" with developed countries 21-22 years later	1950-2005	2008-2035
Portugal	Instituto Nacional de Estatística (2009)	Lee-Carter, Expert opinion	Variant: Poisson log-bilinear model (Brouhns et al. 2002, Bravo 2007)	1980-2007	2008-2060
Spain	Instituto Nacional de Estadística (2009)	Direct extrapolation	5-year moving average; Extension: 3-year period for re-estimation after smoothing the age specific parameter	1991-2007	2009-2049
Sweden	Statistiska centralbyrån (2005)	Lee-Carter	Variant: no correction for time component	1990-2002	2003-2050
United Kingdom	Office for National Statistics (2009)	Target value, Expert opinion	Annual rate of improvement converges to 1.0% in 2033 and remains constant thereafter; Variant: partly cohort approach for convergence	1900-2008	2008-2083

4. Forecasts for the Netherlands

Like in other European countries, in the Netherlands a large number of national and international projections have been undertaken in recent years. The Netherlands is an exception to the broadly parallel upward trend in life expectancy at birth for men and women in western Europe (WHO Health Database, Human Mortality Database). Dutch life expectancy at birth rose from 76.5 years in 1970 to 82.7 years in 2009 for women, and from 70.8 to 78.5 years for men. The yearly increase in male life expectancy was smaller until 2000, and was larger than in other countries from 2000 onwards. Female life expectancy experienced an extended period of relative stagnation between 1980 and 2000. Overall, mortality improvements in other countries in western Europe in the period 1970-2009 were slightly larger and more linear.

Because of this non-linearity, it is not surprising that various agencies and researchers in the Netherlands have paid particular attention to mortality projection methodology. They used different methods and different assumptions, resulting in a number of different mortality projections for the Netherlands (see Table 2). International projects, which contain results for the Netherlands (EUROPOP2010, TOPALS, UPE, and European Demographic Datasheet), use a more general method of extrapolation and targeting. TOPALS (De Beer 2012) makes use of a linear spline, which produces a smoother age curve than, for example, the Lee-Carter method. The European Demographic Datasheet (VID/IASA/PRB 2012) takes into account that mortality dynamics are characterised by considerable inertia, which generates a more optimistic outlook for future mortality. All four take into account trends in other countries.

Projection methods developed specifically for the Netherlands (Actuarial Society, Statistics Netherlands, RIVM and Janssen, van Wissen, and Kunst (2013)), rely less on direct extrapolation. They use different approaches to account for the non-linearity observed in the trends. The method of the Actuarial Society combines a short-term trend with a long-term trend. The short-term trend (eight years of observations) determines the development in the near future, while the eventual level of the forecast is determined by the long-term trend (20 years of observations) using direct linear extrapolation. The RIVM, Janssen, van Wissen, and Kunst (2013) and Statistics Netherlands use epidemiological information in their models. The two former models also use information on trends in other countries. The differences between these two approaches lie in the different extrapolations of $k_{t,i}^{res}$ in equation (3) and in the use of modelled instead of observed jump-off rates. Statistics Netherlands uses a cause-of-death decomposition.

Table 2: Methods, assumptions, and outcomes (e0 in 2050) of different national and international mortality forecasts/projections for the Netherlands

Forecast / projection by	Reference	Type of method	Assumptions	Historical period	e0 2050 NL		
					F	M	Diff.
European Demographic Datasheet 2012	VID/IIASA/PRB (2012), Ediev (2011)	Direct extrapolation, Coherent forecasting, Cohort approach	Extrapolation of exposure-adjusted life table assuming non-divergence and constant mortality conditions. Based on the concept of mortality inertia (Ediev 2011)	1980-2010	91.8	87.7	4.1
EUROPOP2010	Eurostat (2012), Eurostat (2007)	Lee-Carter, Target approach	Convergence mortality rates in 2100 for all EU countries; Variant: Booth et al. (2002) for all countries combined	1960-2009	88.0	84.0	4.0
TOPALS 1	De Beer (2012)	Target approach, direct extrapolation	Extrapolation of the past trends in the risk ratio (ratio between age-specific probabilities of death and a smooth, standard age schedule; i.e., projected age-specific probability of death of Japanese women) for each country separately	1976-2006	86.6*	82.9*	3.7
TOPALS 2	De Beer (2012)	Target approach, Coherent forecasting	Extrapolation of the past trends in the risk ratio of 15 countries in Europe	1976-2006	88.4*	84.7*	3.7
TOPALS 3	De Beer (2012)	Target approach, Coherent forecasting	Extrapolation of the past trends in the risk ratio of 15 countries in Europe and the half time will be half of TOPALS 5	1976-2006	92.1*	89.1*	3.0
UPE	Alders et al. (2007)	Target approach	The same rate of decline for all countries in 2030 (the eventual rate of decline was empirically estimated using 11 countries in a 30-year period). Extension: the change to the eventual rate is linear	1967/1971-1997/2001	86.4	82.5	3.9
Actuarial Society	Actuariel Genootschap & Actuariel Instituut (2010)	Direct extrapolation	Two-year moving average; Extension: old-age adjustment; Extension: correction females	1988-2008	87.3	85.5	1.8
Statistics Netherlands	Statistics Netherlands (2012), Van Duin et al. (2011)	Direct extrapolation, Lee-Carter, Expert opinion, Cause of death	Different assumptions per cause of death, depending on the historical trend	1970-2009	86.6	83.7	2.9
RIVM	Janssen and Kunst (2010)	Explanatory approach, Coherent forecasting	Separate projection of smoking- and non-smoking-related mortality; Including mortality experiences of 10 other European countries	1970-2006	88.1	83.8	4.3
Janssen, van Wissen, and Kunst (2013)	Janssen, van Wissen, and Kunst (2013)	Explanatory approach, Coherent forecasting	Separate projection of smoking- and non-smoking-related mortality; Including mortality experiences of 10 other European countries	1970-2006	87.4	83.6	3.8

Note: * Results for 2050 obtained from the author.

The different national and international mortality forecasts and projections for the Netherlands produce outcomes for life expectancy at birth in 2050 (see Table 2) that vary by 5.7 years for women and by 6.6 years for men. This large range may be caused by the different methods and the different explicit assumptions, including the different historical periods, used. The highest life expectancy at birth in 2050 is given by TOPALS 3; namely, 92.1 years for women and 89.1 years for men. The lowest values are given by UPE for both women and men; namely, 86.4 and 82.5 years, respectively. The higher projected outcomes for both the European Demographic Datasheet and TOPALS 3 result from projection methodologies that are different from the projection of trends in age-specific mortality rates. Instead, the European Demographic Datasheet uses the notion of mortality inertia; i.e., that younger cohorts are healthier than their older peers, and their future mortality rates may therefore be lower than those of the currently old cohorts (Ediev 2011). By contrast, TOPALS 3 assumes an acceleration in the decrease in future mortality. Furthermore, TOPALS uses a best-practice level of mortality in which direction the death probabilities move. The speed is determined by a partial adjustment model.

There is no clear difference in the outcomes of the methods used in the international projects and of the methods specifically developed for the Netherlands. The European Demographic Datasheet and TOPALS 3 predict life expectancies at the high end of the range for both men and women, whereas TOPALS 1 and UPE predict life expectancies at the lower end of the range. In addition, the inclusion of trends in other countries generates mixed results. The European Demographic Datasheet and TOPALS 3 present a high life expectancy in 2050 for both men and women, but TOPALS 1 and UPE give low values. However, the inclusion of trends in other countries, either by coherent forecasting or by the choice of the target value, produces a greater difference in life expectancy between the sexes in 2050 than the other extrapolation methods.

The historical period used ranges from 20 years (Actuarial Society) to 43 years (EUROPOP2010). Statistics Netherlands includes the most recent data.

5. Results of different methods for Dutch mortality

If different methods are applied to the same historical period, a different range of outcomes can be expected. Thus, we apply methods, similar to the ones in Table 1 and 2, which are used in Europe and the Netherlands to Dutch mortality data for the period 1970-2009 and compare the outcomes.

To recap, the methods applied to the Dutch data range from “simple” extrapolation models (direct linear extrapolation and Lee-Carter) to extrapolation models which

account for non-linearity in the data, either by including cohort effects or trends in other populations in the Lee-Carter model, or by using more explanatory approaches; i.e., the separate projection of smoking and non-smoking mortality and the projection by cause of death, as is done in the official Dutch forecasts. These methods are all based on the extrapolation of age-specific death rates. See Section 2 for more details.

Direct linear extrapolation results in a life expectancy at birth in 2050 of 86.5 years for women and 83.3 years for men (Figure 1, Table 3); i.e., an increase of 3.8 years for women and of 4.7 years for men over the next 40 years. The Lee-Carter method gives higher life expectancy values; i.e., 87.4 years for women and 83.8 years for men. The Li-Lee model generates values of 87.7 years for women and 85.0 years for men, which is the highest of the values for men. The cohort model gives a life expectancy at birth of 87.8 years for women and 83.5 years for men. The smoking+non-smoking model, in which smoking-related mortality and non-smoking-related mortality are projected separately, leads to the highest predicted values; i.e., 88.6 years for women and 84.2 years for men. Statistics Netherlands, which uses a cause-of-death decomposition, projects a life expectancy at birth of 86.6 years for women and 83.7 years for men in 2050.

The difference between the models in life expectancy at birth in 2050 is thus 2.1 years for women and 1.8 years for men. The average increase in life expectancy at birth between 2009 and 2050 is 4.8 years for women and 5.4 years for men. The direct extrapolation model results in a lower life expectancy for both men and women than the other models find. The methods which account for the non-linearity generally generate higher outcomes than the simple extrapolation models do.

The increase is almost a straight line for the extrapolation methods, while the cohort model, the smoking+non-smoking model, and the method of Statistics Netherlands are less linear (Figure 1). The straight line of the extrapolation methods is a result of linear, but slightly declining, increases in life expectancy at birth in the period 2009-2050. The yearly increases of the cohort model, the smoking+non-smoking model, and the method of Statistics Netherlands differ from year to year. They differ not only from the extrapolation methods, but also from each other (see Table 3 and compare the observation in 2009 and the outcomes in 2030 and 2050). For instance, compared to all other methods, the cohort model predicts a small increase in life expectancy at birth for men in the period 2009-2030 and a relatively large increase in the period 2030-2050. For women, the smoking+non-smoking model predicts larger increases in the first half and smaller increases in the second half of the period, which results in the same increase in the period 2009-2030 as in the period 2030-2050. The method of Statistics Netherlands predicts greater yearly increases in the first half of the period than the other methods, and constant increases in the second half for both men and women.

Three of the six methods—i.e., the Lee-Carter model, the cohort model, and the smoking+non-smoking model—forecast a larger sex difference in life expectancy at birth in 2050 than was observed in 2009.

When comparing the forecasted values of remaining life expectancy at age 65 in 2050 according to the different forecasting methods, we find that the differences amount to 1.4 years for women, and 1.9 years for men (Figure 2, Table 3). Just as for life expectancy at birth, the smoking+non-smoking model results in the highest remaining life expectancy at 65 for women (25.0 years), and the Li-Lee model results in the highest value for men (22.0 years). The direct extrapolation model results in the lowest value for men (20.2 years) and the second-lowest value for women (23.7 years).

Table 3: Life expectancy at birth and at age 65: observed (2009) and projected values (2030, 2050) for different mortality forecasting methods, the Netherlands, by sex

	Women		Men		Sex difference	
	e0	e65	e0	e65	e0	e65
Observed values 2009	82.65	20.77	78.53	17.41	4.12	3.36
Projected values 2030						
Direct Extrapolation	84.73	22.32	81.23	18.92	3.50	3.40
Lee-Carter	85.25	22.73	81.59	19.16	3.67	3.57
Li-Lee	85.39	22.74	82.05	19.81	3.34	2.92
Cohort model	85.47	22.99	80.78	18.44	4.69	4.55
Smoking+non-smoking model	85.61	22.62	81.77	19.69	3.84	2.92
Statistics Netherlands	84.90	22.40	81.73	19.72	3.17	2.68
Projected values 2050						
Direct Extrapolation	86.49	23.68	83.26	20.16	3.23	3.52
Lee-Carter	87.39	24.41	83.82	20.55	4.57	3.86
Li-Lee	87.72	24.50	85.02	22.03	2.70	2.47
Cohort model	87.80	24.87	83.45	20.27	4.35	4.60
Smoking+non-smoking model	88.59	24.96	84.18	21.48	5.28	4.41
Statistics Netherlands	86.57	23.59	83.65	21.07	2.92	2.52

The largest difference between the outcomes at birth and at age 65 is observed for the smoking+non-smoking model. For the short-term, its effect—in terms of a smaller increase in life expectancy—is much more evident for e65 than for e0. Among men, the outcomes at age 65 drift apart more than the outcomes at birth.

Figure 1: Life expectancy at birth; observed (1970-2009) and projected (2010-2050) values for different mortality forecasting methods, the Netherlands, by sex

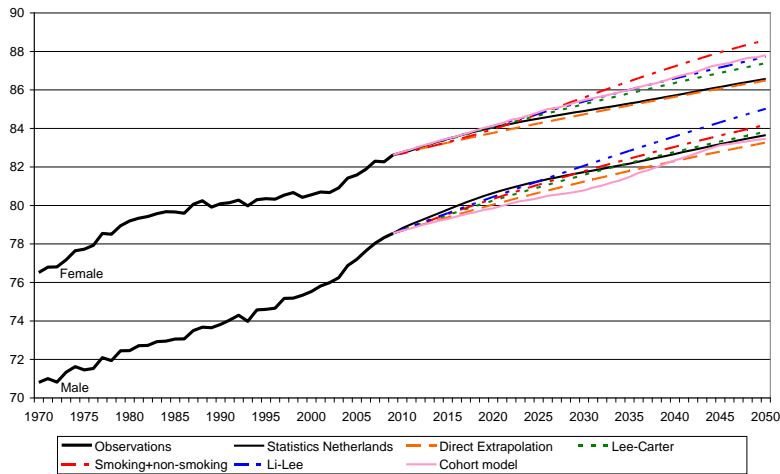
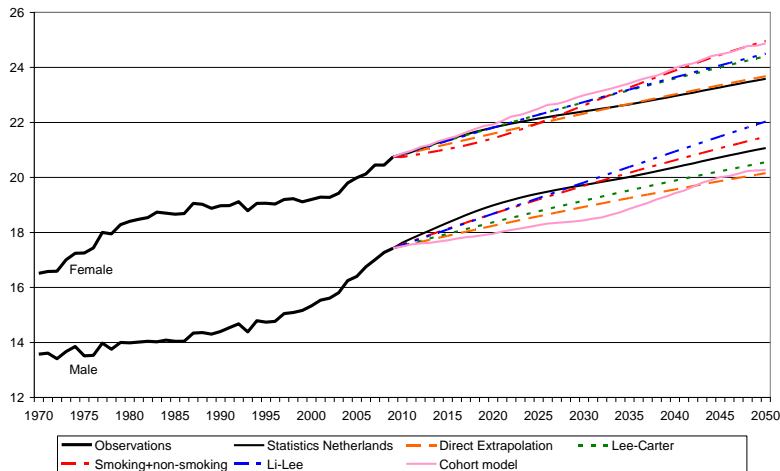


Figure 2: Life expectancy at age 65; observed (1970-2009) and projected (2010-2050) values for different mortality forecasting methods, the Netherlands, by sex



6. Discussion

6.1 Summary of the results

Mortality projection methods and assumptions clearly differ both between and within countries. In the context of new mortality projection methodologies with a focus on extrapolation, statistical offices in Europe currently predict mortality mainly using extrapolation methods (either direct or Lee-Carter), but they also make use of target values, expert opinion, and cause-specific mortality projections. The method and the historical period used seem to reflect past mortality trends in the country, with simple linear extrapolation being used by countries with gradual increases in life expectancy, and other approaches or different assumptions being applied by countries with non-linear trends.

The approaches used in national and international projections for the Netherlands also include simple linear extrapolation methods, but these are mainly methods that take into account the non-linearity in the observations by including trends in other countries, projecting smoking and non-smoking-related mortality separately, or making projections based on causes of death. The 10 different projections for the Netherlands resulted in a wide range for life expectancy at birth in 2050 of 5.7 for women and 6.6 for men, which may be caused by both the different methods and the different explicit assumptions, including the length of the historical period, that are used.

When we compare different methods using the same historical period, including simple linear extrapolation and models that account for non-linearity, we can see that the differences in outcomes become smaller. Life expectancy at birth in 2050 then ranges from 86.5 to 88.6 for women and from 83.3 to 85.0 for men. For life expectancy at age 65 in 2050, the values range from 23.6 to 25.0 for women and from 20.2 to 22.0 for men. The models that account for non-linearity in past trends predict a less linear trend for the future as well.

6.2 Reflection on the differences in the outcomes

The resulting differences in the outcomes of the different methods using the same explicit assumptions—i.e., 2.1 years for women and 1.8 years for men—are small compared to the differences in the outcomes of the 10 different projections for the Netherlands, which were 5.7 years for women and 6.6 years for men. Although the methods used for the Netherlands and the methods we applied to the Dutch data are not exactly the same, it is clear from these results that using the same historical period and other explicit assumptions result in a smaller range of outcomes. Janssen and Kunst

(2007) also found that using different periods may lead to larger differences between the outcomes of the models than the type of models themselves. Moreover, the resulting differences between the outcomes are small compared to the expected average increase in life expectancy at birth between 2009 and 2050, which is 4.8 years for women and 5.4 years for men.

In addition, the range of outcomes, using the same explicit assumptions, is small compared to the range of outcomes for other types of uncertainty. The differences between point forecasts calculated here only describe the uncertainty regarding the type of method. However, there are other types of uncertainty, such as parameter uncertainty (i.e., the uncertainty in the values of the parameters in a given model) and structural uncertainty (i.e., uncertainty because of the stochastic nature of a given model) (Cairns 2000). Parameter and structural uncertainty can be presented by prediction intervals, where parameter uncertainty is very important for long-run forecasts. Statistics Netherlands takes 10 years as a good indicator for the uncertainty of life expectancy at birth in 2050 (Carolina and van Duin 2010). Keilman and Pham (2004) found a 95% prediction interval of life expectancy at birth in 2050 of 13.1 years for women and 7.7 years for men in the Netherlands. This indicates that all projections for the Netherlands presented in this paper can be rated as acceptable, given the estimated uncertainties for earlier forecasts for the Netherlands. It should be noted, however, that the uncertainty of the projections depends to a large extent on the choice of the explicit assumptions, which are not captured by prediction intervals. Furthermore, the variability in rates, which is used to estimate uncertainty, is, on average, underestimated by most methods (Shang et al. 2011).

6.3 Reflection on the explicit assumptions used

In our comparison of the different methods, we used the same explicit assumptions to the greatest extent possible, including the same historical period and the same modelling of k_t . In addition, the most recently observed mortality rates are used as the jump-off rates in all of the projections. These assumptions warrant some attention, however, as they can affect the outcome of the projection, and thus the range of the outcomes.

We used the period 1970-2009 to fit the models. This is the period Statistics Netherlands used in its 2010 official forecast. As we noted above, the choice of the historical period may produce different outcomes. This is expected to influence the outcome of the simple extrapolation methods more than of the more complicated methods, because the latter take into account the possible non-linearity of the data. For instance, the smoking+non-smoking model excludes the non-linear smoking-related

mortality trends from all-cause mortality trends. This allows us to obtain a more robust long-term trend that can be used as the projection basis (Janssen, van Wissen, and Kunst 2013). Furthermore, using a shorter historical period as the projection base may reduce jump-off error, but this advantage may be lost after a number of years.

We used the most recently observed mortality rates as jump-off rates to ensure that the first year of the forecast matches smoothly and closely, as well as to account for a possible jump-off error (Lee and Miller 2001). However, it could also be argued that the actual purpose of a forecast should be robustness; i.e., that when the observations are updated in the future, these small changes in the observations result in only modest changes in the forecasts (Cairns et al. 2011). The advantage of using the original Lee-Carter method (Lee and Carter 1992)—and indeed of using many time series approaches—is that it makes it possible to avoid the peculiarities of mortality rates in a particular year by taking the modelled mortality rates as jump-off rates. The two above-mentioned approaches to forecasting (matching the first year of the forecast versus the robustness of the results) conflict with each other, and may lead to substantial differences in outcomes (Janssen, van Wissen, and Kunst 2013). In addition, the range of outcomes might be different if the modelled death rates are used as jump-off rates. Additional analysis reveals that the outcomes of the simple extrapolation methods and the outcomes for men (because of the large increase in life expectancy since 2002) are influenced the most by the choice of the jump-off rates. For instance, life expectancy at birth in 2050 for men using the direct extrapolation method and modelled jump-off rates is one year lower than with observed jump-off rates. For women, the difference is 0.5 years. The Li-Lee method results in a difference of 0.5 for men and 0.2 for women. Because the effect of using modelled instead of observed mortality rates as jump-off rates is different for the various methods, the range will also be affected.

The modelling of k_t in the Lee-Carter model and its extensions determines part of the future development. We used a random walk with drift, which assumes a linear relationship, and that each forecasted mortality rate changes at a constant exponential rate. Moreover, within the direct linear extrapolation and Lee-Carter model, the rates of decrease at different ages maintain the same ratio to one another over time, because of the assumption of a certain pattern of change in the age distribution of mortality. In practice, the relative speed of decrease at different ages may vary.

The patterns of mortality improvement show a progressive shift of high rates from lower to higher ages in some countries (Andreev and Vaupel 2005). In the past decades, most of the improvement occurred within the younger age groups, and only recently has some improvement been visible at older ages. None of the methods applied here really take into account the potential gains at older ages, as the inertia of past trends dominates the prediction. Estimated mortality improvement at higher ages for the future could therefore be underestimated (see, for instance, the high outcomes of the European

Demographic Datasheet in Table 2, which uses a method that takes into account mortality inertia).

If more information is included in a model, more assumptions need to be made. For instance, the future share of smoking-attributable mortality within the smoking+non-smoking model is also based on assumptions. See Janssen, van Wissen, and Kunst (2013) for a discussion of this issue. In the cohort model we assumed that the cohort effect γ_{t-x} in equation (4) is zero. Hence, there can be a discontinuity of the last observed and the first fully projected cohort. Alternatives are, for instance, γ_{t-x} equal to the last observation or the average of the last few observations.

The longer the projection period, the larger the effect of the assumed cohort effect and of future smoking-related shares in the period life expectancy. In our projections we used a projection horizon equal to the historical period of 40 years. In fact, in more general terms, the effect of the (explicit) assumptions will depend on the chosen projection horizon.

6.4 Explanation of the observed projection outcomes

While the differences in the outcomes of the six methods applied in analysing the Dutch data are small, they are also clear. For instance, the trend in future life expectancy clearly differs. The simple linear extrapolation methods project an almost straight line, whereas the cohort model, the smoking+non-smoking model, and the method of Statistics Netherlands show a less linear future trend. In addition, the simple linear extrapolation methods generally produce lower outcomes than the methods which account for the non-linearity in the observations. Both can be linked to the non-linearity observed in the past trends. The simple linear extrapolation methods extrapolate the average increase of all-cause mortality over the whole period to the future, and thus result in a straight projection to the future. The cohort model, the smoking+non-smoking model, and the method of Statistics Netherlands include the past non-linear trend, and also extrapolate that trend into the future. In this particular case, including the effects causing non-linearity in a model results in a more optimistic increase over the observed period than extrapolating the average increase in all-cause mortality, and thus in a higher life expectancy in the future.

Because the assumptions about the source of the non-linearity vary between these models, the future non-linearity also differs.

The outcome of the cohort model, which differs from the outcomes of the smoking+non-smoking model and the method of Statistics Netherlands, might result from our assumption regarding the cohort effect for new cohorts (see before). For instance, setting the cohort effects to zero for the cohorts outside the data may also

freeze some essential developments in the data. That might also explain why the cohort model predicted different patterns for men and women than the other models.

The smoking+non-smoking model incorporates smoking history, which was not the same for men and women. Among men, smoking had already been decreasing for much of the observation period. Thus, the effect of reduced smoking among men is already reflected in the parameters of the model without the smoking variable. Among women, the turning point from an increase to a decrease in smoking has not yet been reached (Janssen, van Wissen, and Kunst 2013). Thus, the use of the smoking+non-smoking model influences future mortality improvements among women more than among men.

The trend in future life expectancy predicted by Statistics Netherlands is flattened because the cause of death with the worst future trend will dominate future total mortality. Mortality improvements are expected to be large in the near future, but are expected to be relatively small over the longer term.

Three of the six models forecast an increased sex differential in life expectancy at birth between 2009 and 2050: the Lee-Carter model, the cohort model, and the smoking+non-smoking model. Because of the separate projections of life expectancy for men and women, the past trend causes the increasing gap. The gap between male and female life expectancy in 2050 projected by the smoking+non-smoking model, in particular, is large. As men took up smoking before women, the sex difference increased. As women caught up, the gap decreased. The lag in the process for women means, in short, that the more advanced the stage of the smoking epidemic, the closer the smoking mortality rates of men and women become. Thus, in the (near) future, the gender gap may be expected to narrow because of the smaller differences in the smoking habits of men and women. However, as smoking among women and men moves towards parity, trends in non-smoking mortality become important. Because deaths other than from smoking have risen among men relative to women (Pampel 2002), the gap may increase in the future.

For life expectancy at age 65, we see roughly the same results as for life expectancy at birth: an almost straight line and generally lower outcomes for the simple linear extrapolation methods. Among women, an important difference is that, in the short run, the effect of the smoking+non-smoking model in terms of the increase in life expectancy is much more evident for e_{65} than for e_0 . This could be explained by the assumption of an increase in smoking-attributable mortality at older ages, but a decrease at younger ages. A smaller difference is that, among women, the cohort model generates higher life expectancy values at age 65 in 2050 than the other models, relative to the results for life expectancy at birth in 2050. This might again be due to our assumption regarding the cohort effect when projecting e_0 . Among men, we see greater differences between the smoking+non-smoking model on the one hand, and the linear

extrapolation model on the other. This may be related to the ongoing decrease in smoking among older Dutch men.

6.5 Forecasting considerations

Mortality forecasting involves a number of decisions. In addition to the forecasting approach, these include the measure to be forecasted, the specification of an underlying data model, and the specific forecasting method. Among the most important issues that must be taken into account when assessing a forecasting method are the amount of subjectivity, robustness, and whether the outcomes will be logical (e.g., Cairns et al. 2009, 2011). An example of a logical outcome is the plausibility of the projected age-sex profiles of the death rates.

The choice of the forecasting approach (extrapolation, explanation, or expectation) may depend on several criteria. For instance, because the extrapolative approach generally requires a lengthy series of data for long-term forecasting, data availability and the projection period are important criteria. Other examples of criteria are the aim and the accuracy of the forecast.

The decision of which measure is to be extrapolated is also important in forecasting mortality. All of the extrapolation methods used in this paper and by the statistical offices in Europe are variants of extrapolation of death rates. Life expectancy may also be used as the measure to be forecast. Oeppen and Vaupel (2002) reported stable trends in the record life expectancy at birth over 160 years, and White (2002) reported a near-perfect fit of average life expectancy of 21 high-income countries to a straight line from 1955 to 1996; an extrapolation of the life expectancy itself would therefore be an alternative to the methods used in this paper. This different approach will generally be more optimistic than an extrapolation based on death rates. Other examples are methods that take into account mortality inertia (see the European Demographic Datasheet above) or an acceleration of the decrease in mortality (see TOPALS above).

Even when the focus is on the extrapolation of trends in age-specific mortality rates, different specific forecasting methods exist. Especially crucial is the choice of either simple linear extrapolation methods or methods that include additional information. Including additional (e.g., epidemiological) information or adding an extra dimension to the data (e.g., cohort) will automatically lead to the use of a more subjective method, but it may also lead to the use of a more robust method (with respect to small changes in the explicit assumptions, such as the historical period used) that generates more logical (i.e., more epidemiologically sound) outcomes.

Models which capture age, period, and cohort effects in mortality will provide a better model fit than age-period models, given that a cohort effect exists in the mortality data. For some countries, cohort effects are clearly visible, although they are generally much smaller than period effects. However, there is no guarantee that models with a better fit will produce better forecasts. Cohort models in combination with age and period are fairly new and need to be fully tested, particularly in terms of their stability in response to changes in the age range or fitting period (Booth and Tickle 2008).

Epidemiological information can be included in the model in different ways (see, for instance, Pampel 2005; Bongaarts 2006; Janssen and Kunst 2007; Stewart, Cutler, and Rosen 2009; Wang and Preston 2009; King and Soneji 2011; Janssen, van Wissen, and Kunst 2013). Forecasters first have to identify the correlation between the determinant and mortality, and then formulate assumptions about the future. This requires them to have sufficient data on the determinants or indirect techniques to allow them to estimate mortality that can be attributed to a certain risk factor. Currently, no well-defined explanatory models are available, and the determinants are well understood (and measurable) for only a few causes of death. Comorbidities and dependencies among causes make such models even more complex. However, researchers like King and Soneji (2011) have emphasised the potential utility of including risk factors in forecasts, arguing that doing so will improve the quality, accuracy, and transparency of mortality forecasts. A classic example in which the determinant is known and can be measured is the dependence of lung cancer on smoking.

The challenge that arises when using methods that include additional information, such as a cohort effect or epidemiological information, lies in the prediction of the additional information itself. The advantage of using additional information in the forecasting method diminishes if the additional information cannot be forecast more accurately than mortality itself.

The inclusion of trends in other countries in the models is based on the observation that mortality evolution in most developed countries is similar because of similarities in socio-economic factors, life style, medical treatment, etc. Mortality levels will probably continue to develop in parallel. Again, several decisions have to be made, such as about how the information should be included: i.e., by a target approach or by coherent forecasting. Another crucial question is which countries determine the central tendency, or the basic mortality trend, that will be applied to the other countries. In addition, many different coherent forecasting methods exist, such as those of Li and Lee (2005); of Hyndman, Booth, and Yasmeeen (2013); and of Li (2012). The methods for coherent forecasting are relatively new, and more research on this topic is needed.

All in all, the determination of which extrapolation method is optimal will depend on the amount of linearity in the past. If past trends in mortality have been largely

linear, it is better to use the simple extrapolation methods, especially because the outcomes of different extrapolation methods using the same explicit assumptions do not differ greatly. If the past trends have been non-linear, including additional information is likely to result in a more robust forecast if the main effects of the non-linearity are successfully captured. Even though this involves more assumptions and more subjectivity, the right balance between added information and added subjectivity should be achieved.

7. Overall conclusion

In this paper, we have examined the various projection methods currently used in official mortality forecasts in Europe and mortality projections and forecasts in the Netherlands. The methods and the historical period used seem to reflect past mortality trends in the country. The Netherlands, along with other countries, use methods that take into account the non-linearity observed in the past trends. The different projections for a country lead to different results, which may have large implications for the insurance and pension industries.

For the Netherlands, the differences in the outcomes proved to be smaller if the same explicit assumptions were used, such as the same historical period and observed jump-off rates. The remaining sensitivity was shown to be small compared to other forms of uncertainty, and small compared to the increase in life expectancy over the long time horizon. The remaining differences in the outcomes mainly reflect differences between the methods which include the observed non-linearity, and the simple linear extrapolation methods which do not. For countries with more linear trends, smaller differences are likely to result.

Because the choice of explicit assumptions contributes more to the differences in outcome than the choice of the forecasting approach, the choice of the projection method should be based not only on different approaches, but, more importantly, on the explicit assumptions. The method which depends the least on the choice of the explicit assumptions might be the best option. Moreover, in order to improve the comparability of mortality projections from different institutions, the use of sensitivity analyses in which the range of different underlying explicit assumptions are applied would be an important step forward. Finally, it is important to realise that prediction intervals do not capture the uncertainty of the projections due to the choice of the explicit assumptions. Caution is therefore warranted when judging the uncertainty of projections based on prediction intervals only.

References

- Actuariel Genootschap & Actuariel Instituut (2010). AG Prognosetafel 2010-2060 [electronic resource]. Utrecht: Actuariel Genootschap & Actuariel Instituut. [<http://www.ag-ai.nl/download/9166-HR-binnenwerk+Prognosetafel.pdf>].
- Alders, M., Keilman, N., and Cruijsen, H. (2007). Assumptions for long-term stochastic population forecasts in 18 European countries. *European Journal of Population* 23(1): 33–69. doi:10.1007/s10680-006-9104-4.
- Andreev, K. and Vaupel, J. (2005). *Patterns of mortality improvement over age and time in developed countries: Estimation, presentation and implications for mortality forecasting*. Paper presented at the Association of America 2005 Annual Meeting program, Philadelphia, Pennsylvania, March 31 – April 2 2005.
- Bell, W.R. (1997). Comparing and assessing time series methods for forecasting age-specific fertility and mortality rates. *Journal of Official Statistics* 13(3): 279-303.
- Blanpain, N. and Chardon, O. (2006). Projections de population 2007-2060 pour la France métropolitaine: méthode et principaux résultats [electronic resource]. France: Institut national de la statistique et des études économiques. [http://www.insee.fr/fr/publications-et-services/docs_doc_travail/docf1008.pdf].
- Bongaarts, J. (2006). How long will we live? *Population and Development Review* 34(4): 605-628. doi:10.1111/j.1728-4457.2006.00144.x.
- Booth, H., Hyndman, R.J., Tickle, L., and De Jong, P. (2006). Lee-Carter mortality forecasting: A multi-country comparison of variants and extensions. *Demographic Research* 15(9): 289–310. doi:10.4054/DemRes.2006.15.9.
- Booth, H., Maindonald, J., and Smith, L. (2002). Applying Lee-Carter under conditions of variable mortality decline. *Population Studies* 56(3): 325–336. doi:10.1080/00324720215935.
- Booth, H. and Tickle, L. (2008). Mortality modelling and forecasting: A review of methods. *Annals of Actuarial Science* 3(1-2): 3-43. doi:10.1017/S1748499500000440.
- Booth, H., Tickle, L., and Smith, L. (2005). Evaluation of the variants of the Lee-Carter method of forecasting mortality: A multi-country comparison. *New Zealand Population Review* 31(1): 13–34.
- Brass, W. (1971). On the scale of mortality. In: Brass, W. (ed.). *Biological aspects of demography*. London: Taylor and Francis: 69-110.

- Bravo, J.M. (2007). Tábuas de mortalidade contemporâneas e prospectivas: Modelos estocásticos, aplicações actuariais e cobertura do risco de longevidade. [PhD thesis]. Portugal: University of Évora, Economics.
- Brouhns, N., Denuit, M., and Vermunt, J. (2002). A Poisson log-bilinear regression approach to the construction of projected lifetables. *Insurance: Mathematics and Economics* 31(3): 373-393. doi:10.1016/S0167-6687(02)00185-3.
- Brunborg, H. and Texmon, I. (2010). Befolkningsframskrivninger 2010-2060. *Økonomiske analyser* 4/2010: 28-39.
- Bureau fédéral du Plan (2009). Quotients de mortalité prospectifs par sexe et unisexes [electronic resource]. Brussel: Bureau fédéral du Plan. [http://www.plan.be/admin/uploaded/201002040818570.wp200918.pdf].
- Cairns, A.J.G. (2000). A discussion of parameter and model uncertainty in insurance. *Insurance: Mathematics and Economics* 27(3): 313-330. doi:10.1016/S0167-6687(00)00055-X.
- Cairns, A.J.G., Blake, D., and Dowd, K. (2006). A two-factor model for stochastic mortality with parameter uncertainty: Theory and calibration. *Journal of Risk and Insurance* 73(4): 687-718. doi:10.1111/j.1539-6975.2006.00195.x.
- Cairns, A.J.G., Blake, D., Dowd, K., Coughlan, G.D., Epstein, D., and Khalaf-Allah, M. (2011). Mortality density forecasts: An analysis of six stochastic mortality models. *Insurance: Mathematics and Economics* 48(3): 355-367. doi:10.1016/j.insmatheco.2010.12.005.
- Cairns, A.J.G., Blake, D., Dowd, K., Coughlan, G.D., Epstein, D., Ong, A., and Balevich, I. (2009). A quantitative comparison of stochastic mortality models using data from England & Wales and the United States. *North American Actuarial Journal* 13(1): 1-35. doi:10.1080/10920277.2009.10597538.
- Carolina, N. and van Duin, C. (2010). Onzekeihidsmarges voor de sterfteprognose van het CBS. *Bevolkingstrends* 58(2): 32-37.
- Central Statistics Office (2008). Population and labour force projections 2011-2040 [electronic resource]. Ireland: Central Statistics Office. [http://www.cso.ie/en/media/csoie/releasespublications/documents/population/2008/poplabfor_2011-2041.pdf].
- Continuous Mortality Investigation (CMI) (2005). Projecting future mortality: Towards a proposal for a stochastic methodology. (working paper 15).

- Continuous Mortality Investigation (CMI) (2006). Stochastic projection methodologies: Further progress and P-Spline model features, example results and implications. (working paper 20).
- Continuous Mortality Investigation (CMI) (2007). Stochastic projection methodologies: Lee-Carter model features, example results and implications. (working paper 25).
- De Beer, J. (2012). Smoothing and projecting age-specific probabilities of deaths by TOPALS. *Demographic Research* 27(20): 543-592. doi:10.4054/DemRes.2012.27.20.
- De Jong, A. and Van Der Meulen, A. (2005). Prognose van sterfte naar doodsoorzaken: Model en veronderstellingen. *Bevolkingstrends* 53(2): 50-62.
- De Jong, P. and Tickle, L. (2006). Extending Lee-Carter mortality forecasting. *Mathematical Population Studies* 13(1): 1-18. doi:10.1080/08898480500452109.
- Ediev, D.M. (2008). Extrapolative projections of mortality: Towards a more consistent method. Part I: The central scenario. Vienna Institute of Demography of Austrian Academy of Sciences (working paper; WP 03).
- Ediev, D.M. (2011). Life expectancy in developed countries is higher than conventionally estimated. Implications from improved measurement of human longevity. *Journal of Population Ageing* 4(1-2): 5-32. doi:10.1007/s12062-011-9040-x.
- Eurostat (2007). *EUROPOP2007 Convergence Scenario: Summary Note*. Paper presented at the Joint Eurostat-AWG Working Group on Population Projections, Luxembourg, 29-30 November 2007.
- Eurostat (2012). Population statistics [electronic resource]. [<http://epp.eurostat.ec.europa.eu/portal/page/portal/population/data/database>].
- Główny Urząd Statystyczny (2009). Prognoza ludności na lata 2008–2035 [electronic resource]. Warsaw: Główny Urząd Statystyczny. [http://www.stat.gov.pl/cps/rde/xbcr/gus/P_Population_projection_for_Poland_2008_2035.pdf].
- Hanika, M.A. (2010). Kleinräumige Bevölkerungsprognose für Österreich 2010-2030 mit Ausblick bis 2050 („ÖROK-Prognosen“) [electronic resource]. Vienna: Statistics Austria. [http://www.statistik.at/web_de/static/kleinraeumige_bevoelkerungsprognose_fuer_oesterreich_2010-2030_mit_ausblic_051902.pdf].

- Hansen, M.F. and Stephensen, P. (2010). Danmarks fremtidige befolkning - Befolkningsfremskrivning 2010 [electronic resource]. Copenhagen: DREAM. [<http://www.dreammodel.dk/pdf/Befolkning2010.pdf>].
- HMD. Human Mortality Database [electronic resource]. [<http://www.mortality.org>].
- Hyndman, R.J., Booth, H., and Yasmeen, F. (2013). Coherent mortality forecasting: The product-ratio method with functional time series models. *Demography* 50(1): 261-283. doi:10.1007/s13524-012-0145-5.
- Hyndman, R.J. and Ullah, M.S. (2007). Robust forecasting of mortality and fertility rates: A functional data approach. *Computational Statistics and Data Analysis* 51(10): 4942-4956. doi:10.1016/j.csda.2006.07.028.
- Instituto Nacional de Estadística (2009). Proyección de la Población de España a Largo Plazo (2009-2049) [electronic resource]. Madrid: Instituto Nacional de Estadística. [<http://www.ine.es/metodologia/t20/t2030251.pdf>].
- Instituto Nacional de Estatística (2009). Projecções de população residente em Portugal [electronic resource]. Lisboa: Instituto Nacional de Estatística. [http://www.ine.pt/ngt_server/attachfileu.jsp?look_parentBoui=65945988&att_display=n&att_download=y].
- Janssen, F. and Kunst, A. (2007). The choice among past trends as a basis for the prediction of future trends in old-age mortality. *Population Studies* 61(3): 315-326. doi:10.1080/00324720701571632.
- Janssen, F. and Kunst, A. (2010). De toekomstige levensverwachting. In: Luijben, A.H.P. and Kommer, G.J. (eds.). *Tijd en toekomst; deelrapport van de VTV 2010 Van gezond naar beter*. RIVM-rapport 270061008, Houten: Bohn Stafleu Van Loghum: 13-20.
- Janssen, F., Kunst, A., and Mackenbach, J. (2007). Variations in the pace of old-age mortality decline in seven European countries, 1950–1999: The role of smoking and other factors earlier in life. *European Journal of Population* 23(2): 171-188. doi:10.1007/s10680-007-9119-5.
- Janssen, F., Van Wissen, L.J.G., and Kunst, A.E. (2013). Including the smoking epidemic in internationally coherent mortality projections. *Demography* 50(4): 1341-1362. doi:10.1007/s13524-012-0185-x.
- Keilman, N. and Pham, D.Q. (2004). Empirical errors and predicted errors in fertility, mortality and migration forecasts in the European Economic Area. Statistics Norway, Social and Demographic Research (Discussion Papers No. 386).

- Keilman, N. and Pham, D.Q. (2005). Hvor lenge kommer vi til å leve? Levealder og aldersmønstre for dødeligheten i Norge, 1900–2060. *Økonomiske analyser* 6/2005: 43-49.
- King, G. and Soneji, S. (2011). The future of death in America. *Demographic Research* 25(1): 1-38. doi:10.4054/DemRes.2011.25.1.
- Lee, R.D. and Carter, L.R. (1992). Modelling and forecasting US mortality. *Journal of the American Statistical Association* 87(419): 659-671.
- Lee, R.D. and Miller, T. (2001). Evaluating the performance of the Lee-Carter method for forecasting mortality. *Demography* 38(4): 537-549. doi:10.1353/dem.2001.0036.
- Li, J. (2012). A Poisson common factor model for projecting mortality and life expectancy jointly for females and males. *Population Studies* 67(1): 111-126. doi:10.1080/00324728.2012.689316.
- Li, N.R. and Lee, R. (2005). Coherent mortality forecasts for a group of populations: An extension of the Lee-Carter method. *Demography* 42(3): 575-594. doi:10.1353/dem.2005.0021.
- OECD (2007). Pension reform: The unfinished agenda. Policy brief, September 2007. doi:10.1787/growth-2007-en.
- Oeppen, J. and Vaupel, J.W. (2002). Broken limits to life expectancy. *Science* 296(5570): 1029-1031. doi:10.1126/science.1069675.
- Office for National Statistics (2009). 2008-based National Population Projections [electronic resource]. Newport: Office for National Statistics. [http://www.ons.gov.uk/ons/rel/npp/national-population-projections/2008-based-projections/national-population-projections.pdf]
- Olshansky, S.J., Passaro, D., Hershow, R., Layden, J., Carnes, B., Brody, J., Hayflick, L., Butler, R., Allison, D., and Ludwig, D. (2005). A potential decline in life expectancy in the United States in the 21st century. *The New England Journal of Medicine* 352(11): 1138–1145. doi:10.1056/NEJMs043743.
- Pampel, F.C. (2002). Cigarette use and the narrowing sex differential in mortality. *Population and Development Review* 28(1): 77-104. doi:10.1111/j.1728-4457.2002.00077.x.
- Pampel, F.C. (2005). Forecasting sex differences in mortality in high income nations: The contribution of smoking. *Demographic Research* 13(18): 455-484. doi:10.4054/DemRes.2005.13.18.

- Pollard, J.H. (1987). Projection of age-specific mortality rates. *Population Bulletin of the United Nations* 21-22: 55-69.
- Renshaw, A.E. and Haberman, S. (2006). A cohort-based extension to the Lee–Carter model for mortality reduction factors. *Insurance: Mathematics and Economics* 38(3): 556-570. doi:10.1016/j.insmatheco.2005.12.001.
- Salvini, S., Santini, A., and Vignoli, D. (2006). Le Previsiono della Popolazione. Teoria ed applicazioni [electronic resource]. Firenze: Università degli Studi di Firenze. [http://www.ds.unifi.it/ricerca/pubblicazioni/altre/didattica/didattica2006_1.pdf].
- Shang, H.L., Booth, H., and Hyndman, R.J. (2011). Point and interval forecasts of mortality rates and life expectancy: A comparison of ten principal component methods. *Demographic Research* 25(5): 173-214. doi:10.4054/DemRes.2011.25.5.
- STATEC (2005). Potentiel de croissance économique et Démographie Projections 2005 –2055 [electronic resource]. Luxembourg: Conférence de presse du 7 octobre 2005. [<http://www.statistiques.public.lu/fr/actualites/population/population/2005/10/20051007/PDF-Present-ProjectDemograph-2005.pdf>].
- Statistics Netherlands (2012). Statline database [electronic resource]. Den Haag/Heerlen, Centraal Bureau voor de Statistiek. [<http://statline.cbs.nl/StatWeb/default.aspx>].
- Statistiska centralbyrån (2005). SCB:s modell för befolkningsprognoser. En dokumentation [electronic resource]. Stockholm: Statistics Sweden. [http://www.scb.se/statistik/_publikationer/BE0401_2005A01_BR_BE52OP0501.pdf].
- Stewart, S.T., Cutler, D.M., and Rosen, A.B. (2009). Forecasting the effects of obesity and smoking on U.S. life expectancy. *The New England Journal of Medicine* 361(23): 2252-2260. doi:10.1056/NEJMsa0900459.
- Tabeau, E. (2001). A review of demographic forecasting models for mortality. In: Tabeau, E., Van Den Berg Jeths, A., and Heathcote, C. (eds.). *Forecasting mortality in developed countries: Insights from a statistical, demographic and epidemiological perspective*. Dordrecht: Kluwer Academic Publishers: 1-32.
- Van Duin, C., De Jong, G., Stoeldraijer, L., and Garssen, J. (2011). Bevolkingsprognose 2010-2060: Model en veronderstellingen betreffende de sterfte. *Bevolkingstrends* 59(2): 28-41.

- VID/IIASA/PRB (2012). [electronic resource] [http://www.oeaw.ac.at/vid/datasheet/download_2012.shtml].
- Wang, C.-W. and Liu, Y.-L. (2010). Comparisons of mortality and forecasting – empirical evidence from Taiwan. *International Research Journal of Finance and Economics* 37: 46-55.
- Wang H. and Preston, S.H. (2009). Forecasting United States mortality using cohort smoking histories. *PNAS* 106(2): 393-398. doi:10.1073/pnas.0811809106.
- White, K.M. (2002). Longevity advances in high-income countries, 1955-96. *Population and Development Review* 28(1): 59-76. doi:10.1111/j.1728-4457.2002.00059.x.
- WHO. Global Health Observatory Data Repository [electronic resource]. [<http://apps.who.int/ghodata/>].
- Wong-Fupuy, C. and Haberman, S. (2004). Projecting mortality trends: Recent developments in the United Kingdom and the United States. *North American Actuarial Journal* 8(2): 56–83. doi:10.1080/10920277.2004.10596137.

