Seasonal Mortality in Denmark:
The role of sex and age

Roland Rau (rau@demogr.mpg.de)
Gabriele Doblhammer (doblhammer@demogr.mpg.de)
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Roland Rau, Gabriele Doblhammer*

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1 Introduction

“Death strikes from every side, but not at random” [37, p. 505]. While an individual death may not be predicted, it is possible to make a probabilis- tic forecast about the timing of death. In Western countries, one is most likely to die during the first few months of the year. During summer, on the contrary, mortality is lowest [3, 16, 20, 28, 32, 33, 34]. These fluctuations “are one of the ‘deep structures’ that identify the main environmental and cultural factors that form a given population” [53, p. 100]. Although climate shapes the basic seasonal pattern of mortality, one cannot necessarily equate colder climate with larger monthly oscillations. Indeed, Canada, Russia and the Scandinavian countries show a lower percentage of excess deaths during winter than, for example, the UK and many Mediterranean countries [19, 41]. It has been estimated for the late 1970s that about half a million deaths per year in North America, the USSR, and Europe were cold-related [19]. It is argued that these deaths are mainly an outcome of exposure to outdoor and indoor cold; people living in rather severe climatic zones protect themselves better against both kinds of hazards [11, 12, 13, 14, 15, 25, 26].

Previous studies often focussed on the dampening of the seasonal fluctuations in mortality over time [28, 36, 38, 40]. However, this trend towards deseasonalization is not generally applicable as shown, for instance, for France and the UK until the 1970s [3, 51]. As a consequence, we first examined whether seasonality is still present in Danish mortality. Denmark may not

*Both authors: Max Planck Institute for Demographic Research, Konrad Zuse Str. 1, 18057 Rostock, Germany. E-Mail-Adresses: rau@demogr.mpg.de; doblhammer@demogr.mpg.de; The authors would like to thank Francesco C. Billari for his valuable technical comments.
inevitably follow the pattern of other countries in seasonal mortality as it sets itself apart from its neighbors in mortality trends. Life expectancy, for example, rose slower than anywhere else in western Europe in the 1980s [7, 10]. Instead of examining the trend over time, we analyzed the factors sex and age. They usually received little attention in previous studies, despite their paramount influences on mortality [49, p. 7–8]. A usual assumption among demographers asserts that mortality measures the current conditions of the ecological and social environment [47]. According to that assessment, women and men vary in their susceptibility towards environmental hazards as reflected by the lower female age-specific mortality-rates throughout the life-course. Consequently, we were puzzled by the results of several studies on seasonality that included the factor sex. They typically found no significant differences between women and men in seasonality in all-cause mortality [14, 17, 43, 62].

The factor age has been analyzed in more detail than sex. However, the basis of the data implied some problems in previous studies. Sometimes no age distinction was made at all [3, 5, 50, 54]. In other studies the highest included age or the beginning of the last, open-ended, age category was chosen at an age after which most deaths in a population occur [6, 9, 11, 13, 14, 15, 22, 23, 26, 42, 55]. Thus, results from these studies may simplify or blur the relationship between age and seasonal fluctuations in mortality. Some studies analyzed heat-related mortality up into very advanced ages [35, 43], but only a few investigated overall seasonal mortality in these age-groups [16, 40, 44]. The general trend in those studies asserts an increase of seasonality with age. This fits our framework of decreasing resistance towards environmental hazards with age.

2 Data and Methods

Our data consisted of all Danes who were 50 years or older on 1 April 1968. These 1,374,536 individuals were followed for 30 years until March 1998. 1,994 people were lost (censored) during the observation period (0.1 percent). 1,171,535 individuals (85.2 percent) died, leaving 201,007 survivors (14.6 percent) in the end of March 1998 behind. For each individual, birth and death (or censoring) have been recorded by month and year.

We calculated winter excess mortality for the whole population following Grut [19] (which is similar to McKee [41]) to have a descriptive and comparable measurement to other countries in respect to the amount of possibly

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1 An exception is represented by the article of Robine and Vaupel [48] which analyzes exclusively people above ages 100 and 110, respectively.
preventable deaths:

\[ EW_D = D - \left( 12 \times \frac{D_{JUL} + D_{AUG} + D_{SEP}}{3} \right) \]  

\(^{(1)}\)

\(EW_D\) represents the number of excess winter deaths, \(D\) is the number of all deaths during the follow-up and \(D_{JUL}, D_{AUG}, D_{SEP}\) represent the number of deaths during July, August, and September, respectively. The proportion of winter excess deaths is obtained by computing \(EW_D/D\).

Our main objective was to analyze the changing risks of dying in different months of the year. We created a data-set that contains a person as many times as he/she has lived in months between April 1968 and March 1998. When using the whole Danish population at ages 50+ this procedure would lead to a huge data-set. We therefore opted to draw a random-sample of each cohort and sex. The final data-set included 46,293 individuals (≈ 3.4 percent sample).

We further divided the population by using four ten-year birth cohorts: the oldest cohort (Cohort I) was born between April 1878 and March 1888. In April 1968 they were aged between 80 and 89 years and 11 months. We followed them until they reached a maximum age of 99 years and 11 months (March 1978 to March 1988). The second cohort (birth dates between April 1888 and March 1898) is aged 70 to 79 years and 11 months in 1968 and is followed to age 99 years and 11 months (time period March 1988 - March 1998). The third cohort (born April 1898 — March 1908) is aged 60 to 69 years and 11 months at baseline. At the end of the follow-up (March 1998) they have reached ages 90 to 99 and 11 months. The fourth cohort (born April 1908 — March 1918) is aged 50 to 59 and 11 months at baseline and reaches maximum ages of 80 to 89 and 11 months in March 1998. Please see Figure 1 for a graphical description of this classification. The exact numbers of individuals in each cohort by birth date, sex, and survival status are given in Table 1. We controlled for different length of month by standardizing all records to a weight of 30 days.

The age specific analysis is based on a slightly different definition of cohorts. The goal was that every member of a specific cohort should theoretically be able to reach each analyzed age. For example, people in Cohort III who were 60 years old in the beginning of the follow-up could attain a maximum of 90 years. Analogously, people in the same cohort aged 69 years in 1968 have no possibility to be exposed to the risk of dying at age 65 anymore. Therefore, we transformed the cohorts into parallelograms as follows (see Figure 2):

In each of the cohorts we constructed two age groups. The oldest cohort was followed from age 90 (time period April 1968 to April 1978) until they
reached the age 94 years and 11 months (Fig. 2, $D_1$). Within the same birth cohort we compared this age group with the 5-year age group that reached age 95 between April 1973 and April 1978 and attained a maximum age of 99 years and 11 months between March 1978 and March 1988 ($D_2$). In the second cohort the first age group consists of all those who were aged 80 in the time period April 1968 to April 1973 and reached a maximum age of 89 years and 11 months between March 1978 and March 1988 ($C_1$). The second age group in Cohort II comprises ages 90 (time period April 1978 to April 1988) to 99 years and 11 months (time period March 1988 to March 1998) ($C_2$). The first age group in the third cohort consists of ages 70 to 79 years and 11 months ($B_1$), the second of ages 80 to 89 years and 11 months ($B_2$). In the youngest cohort, the first age group comprises ages 60 to 69 years and 11 months ($A_1$); the second age group, ages 70 to 79 years, 11 months ($A_2$).
The time periods for the respective age groups in the two youngest cohorts are the same as those specified in Cohort II.

We analyzed the mortality of our subjects by choosing a logistic regression model. Following Allison [2], the effects of the covariates are modeled by using Equation 2:

Although death can happen at any point in time, our data-set specifies only month of death. Thus, from a theoretical point of view, a continuous-time approach cannot be justified because we are interested in analyzing the effect of a time-varying covariate (current month) which changes its values at exactly the same intervals of time as the transition variable death. In practice, the differences between our logistic regression model and a Cox proportional hazards model turned out to be rather negligible. See Table 1 in the Appendix.
Table 1: Sample Population of follow-up from 1968 until 1998 of elderly Danish people by cohort classification, sex, and survival status

<table>
<thead>
<tr>
<th>Cohort</th>
<th>Birth Date</th>
<th>Sex</th>
<th>Alive in April 1968</th>
<th>Person-months lived</th>
<th>Surviving March 1998*</th>
</tr>
</thead>
<tbody>
<tr>
<td>I</td>
<td>April 1878—March 1888 Female</td>
<td>2,072</td>
<td>144,257.0</td>
<td>44</td>
<td></td>
</tr>
<tr>
<td>I</td>
<td>April 1878—March 1888 Male</td>
<td>1,551</td>
<td>100,158.0</td>
<td>25</td>
<td></td>
</tr>
<tr>
<td>II</td>
<td>April 1888—March 1898 Female</td>
<td>5,792</td>
<td>760,124.5</td>
<td>93</td>
<td></td>
</tr>
<tr>
<td>II</td>
<td>April 1888—March 1898 Male</td>
<td>4,715</td>
<td>509,106.5</td>
<td>18</td>
<td></td>
</tr>
<tr>
<td>III</td>
<td>April 1898—March 1908 Female</td>
<td>8,464</td>
<td>1,646,669.5</td>
<td>95</td>
<td></td>
</tr>
<tr>
<td>III</td>
<td>April 1898—March 1908 Male</td>
<td>8,117</td>
<td>1,341,481.5</td>
<td>80</td>
<td></td>
</tr>
<tr>
<td>IV</td>
<td>April 1908—March 1918 Female</td>
<td>6,966</td>
<td>1,578,623.5</td>
<td>177</td>
<td></td>
</tr>
<tr>
<td>IV</td>
<td>April 1908—March 1918 Male</td>
<td>8,616</td>
<td>1,790,307.0</td>
<td>152</td>
<td></td>
</tr>
<tr>
<td>I-IV</td>
<td>April 1878—March 1918 Female</td>
<td>23,294</td>
<td>4,129,674.5</td>
<td>409</td>
<td></td>
</tr>
<tr>
<td>I-IV</td>
<td>April 1878—March 1918 Male</td>
<td>22,999</td>
<td>3,741,053.0</td>
<td>275</td>
<td></td>
</tr>
<tr>
<td>I-IV</td>
<td>April 1878—March 1918 Females and Males</td>
<td>46,293</td>
<td>15,741,455.0</td>
<td>684</td>
<td></td>
</tr>
</tbody>
</table>

*Cohorts I and II: Surviving to Age 99 Years and 11 Months*
\[
\log \left( \frac{P_{it}}{1 - P_{it}} \right) = \alpha_t + \sum_{m=1}^{11} \beta_m x_{i,m} + \delta_{\text{period}} v_{i,t} + \gamma_{\text{Age}} w_{i,t}
\] (2)

The conditional probability \( P_{it} \) that the event happens to individual \( i \) at time \( t \) — given it has not happened before \((1 - P_{it})\) — is related to an intercept \((\alpha_t)\) and a set of further covariates. The \( \beta \)-parameters estimate the effect of 11 dummy variables representing current month with one month serving as a reference group. The parameters \( \delta \) and \( \gamma \) control for period effects and for age (time-varying), respectively. The odds-ratios (the exponentiated \( \beta \)-parameters) can be used to assess approximately the relative risks if the number of occurrences is rather small compared to the risk-set [61]. In our analysis, we controlled for left-truncation because of the different ages of the subjects in the beginning of the follow-up. Whenever groups were contrasted (women and men; different age-groups), we opted to calculate separate models instead of including interaction effects.

Hewitt’s test [21] was employed to investigate whether the relative risks of dying follow a seasonal pattern. This test gives ranks to each month. The value “12” is assigned to the month with the highest relative risk, and “1” to the month with the lowest relative risk. Keeping the original order of the months (January, February, ..., December), the test statistic is the maximum rank-sum of six consecutive months. Thus, this method assumes that the year is split into two 6-months-periods with a relatively high risk to experience the event in one half and a relatively low risk during the other half. Simulated significance levels [21] were applied for Hewitt’s test if exact values were not available [59].

3 Results

Monthly standardized death counts:
⇒ Denmark displays the typical Western pattern.

Figure 3 shows the monthly distribution of the 1,171,535 deaths of the whole data-set in percent after standardizing each month to thirty days. The bars indicate a seasonally changing pattern peaking in January (106,037 standardized deaths) and reaching a minimum in August (88,395 standardized deaths). According to our definition 78,375 deaths during our observation period of 30 years can be attributed to winter excess mortality or more than 2,600 deaths each year. This equals a proportion of 6.69 percent of all deaths.
Figure 3: Distribution of Monthly Mortality in Percent after Standardizing Length of Month

Number of All (standardized) Deaths: 1,155,068
Number of All Excess Winter Deaths: 78,375
Annual Number of Excess Winter Deaths: 2,612.5
Proportion of Excess Winter Deaths: 6.69 Percent
(see Grut 1987 for Definition of Excess Winter Deaths)
Monthly odds ratios for the four cohorts:

⇒ Seasonal fluctuations are larger in the older cohorts.

Table 2 shows the results of a logistic regression model based on our sample estimating the odds-ratios (OR) for each month of the year on the risk of dying. The risk for the youngest cohort (Cohort IV: ages 50/59 to ages 80/89) is 17% higher in January than in the reference month August. The older the cohorts, the higher the differences between the minimum and the maximum. The height of the amplitudes increases to 34 percent in the oldest cohort (Cohort I: ages 80/89 to ages 90/99). The patterns differ in the four cohorts. The p-values for Hewitt’s test for seasonality reach an acceptable level of significance only for Cohorts I - III (p ≤ 0.0253). The youngest cohort (Cohort IV), on the contrary, does not seem to follow a typical seasonal pattern (p = 0.2908).

Sex-specific odds ratios by season for the four cohorts:

⇒ Men experience larger seasonal fluctuations in mortality than women.

Figure 4 gives the result of a similar estimation as Table 2 with two exceptions: First, we conducted separate analyzes for women and men. Second, months have been summarized to seasons (Winter: January, February, March; Spring: April, May, June; Summer: July, August September; Fall: October, November, December) to clarify trends.

The youngest cohort (Cohort IV: ages 50/59 to ages 80/89, time period 1968 to 1998) displays virtually no difference between the sexes. With odds-ratios of 1.08 for women and 1.07 for men, the mortality risk is highest for both sexes in winter. Cohort III (ages 60/69 to ages 90/99, time period 1968 to 1998), which is on average 10 years older than Cohort IV, shows a similar pattern with a peak in winter and a trough in summer for both sexes. However, the relative mortality risk is remarkably higher for men than for women (OR winter: 1.14 vs 1.08; OR spring: 1.07 vs. 1.03; OR fall: 1.10 vs 1.06). Since both cohorts cover the same time period the increase in the seasonality must be due to differences in age.

In Cohort II (ages 70/79 to age 99, time period 1968 to 1988/1989), men face excess mortality of 20 percent in winter and 21 percent in spring, women’s excess mortality is 1.14 and 1.11, respectively. The oldest people (Cohort I: ages 80/89 to age 99, time period 1968 to 1978/1988) are reversing the trend. Men’s seasonal fluctuations are smaller than women’s and their trough changed from summer to spring (OR spring: 0.97). Women in that cohort faced higher relative risks during each season than in any other cohort, reaching a maximum in winter with a relative risk of 31 percent in
Table 2: Odds-Ratios (OR) and Levels of Significance for Month of Death from Logistic Regression (Reference Group: August)

<table>
<thead>
<tr>
<th>Month of Death</th>
<th>Cohort I (Born April 1878 - March 1888)</th>
<th>Cohort II (Born April 1888 - March 1898)</th>
<th>Cohort III (Born April 1898 - March 1908)</th>
<th>Cohort IV (Born April 1908 - March 1918)</th>
</tr>
</thead>
<tbody>
<tr>
<td>January</td>
<td>1.29 ***</td>
<td>1.21 ***</td>
<td>1.11 ***</td>
<td>1.17 ***</td>
</tr>
<tr>
<td>February</td>
<td>1.23 **</td>
<td>1.18 ***</td>
<td>1.04</td>
<td>1.05</td>
</tr>
<tr>
<td>March</td>
<td>1.33 ***</td>
<td>1.22 ***</td>
<td>1.16 ***</td>
<td>1.08 *</td>
</tr>
<tr>
<td>April</td>
<td>1.11</td>
<td>1.22 ***</td>
<td>1.08 **</td>
<td>1.05</td>
</tr>
<tr>
<td>May</td>
<td>1.19 *</td>
<td>1.16 ***</td>
<td>1.03</td>
<td>1.07</td>
</tr>
<tr>
<td>June</td>
<td>1.06</td>
<td>1.17 ***</td>
<td>1.01</td>
<td>1.06</td>
</tr>
<tr>
<td>July</td>
<td>1.03</td>
<td>1.06</td>
<td>1.01</td>
<td>1.04</td>
</tr>
<tr>
<td>August</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
</tr>
<tr>
<td>September</td>
<td>1.04</td>
<td>1.02</td>
<td>0.97</td>
<td>1.03</td>
</tr>
<tr>
<td>October</td>
<td>1.10</td>
<td>1.14 ***</td>
<td>1.02</td>
<td>1.06</td>
</tr>
<tr>
<td>November</td>
<td>1.24 **</td>
<td>1.11 **</td>
<td>1.06</td>
<td>1.04</td>
</tr>
<tr>
<td>December</td>
<td>1.22 **</td>
<td>1.20 ***</td>
<td>1.12 ***</td>
<td>1.16 ***</td>
</tr>
<tr>
<td>Height of Amplitude</td>
<td>33%</td>
<td>22%</td>
<td>20%</td>
<td>17%</td>
</tr>
</tbody>
</table>

* $p < 0.1$, **$p < 0.05$, ***$p < 0.01$

rank-sum (Hewitt’s Test) | 56          | 55          | 57          | 53          |

$p$-value for Hewitt’s Test | 0.0253      | 0.0483      | 0.0123      | 0.01299     |
Figure 4: Odds-Ratios and 95% Confidence Intervals for Seasonal Mortality by Sex (Reference Group: Summer)

Cohort I
(Born April 1878 - March 1888)

Cohort II
(Born April 1888 - March 1898)

Cohort III
(Born April 1898 - March 1908)

Cohort IV
(Born April 1908 - March 1918)
winter compared to summer.

Since the oldest cohort only covers the time period 1968/1973 to 1978/1983 (opposed to the other cohorts covering 1968/1978 to 1988/98), we cannot distinguish whether the particular pattern in this cohort is due to age effects or period effects.

**Sex and age-specific odds ratios by season for the four cohorts:**

⇒ *Seasonal fluctuations in mortality start increasing at later ages for women than for men.*

Figures 5 and 6 show the effect of age on the seasonality in mortality. Men, generally speaking, exhibit an increase of seasonal mortality fluctuations with age. In the youngest cohort the relative mortality risk for septagenarians compared to 60-69 year olds increased by 1 percentage point in spring, 5 percentage points in fall, and by 6 percentage points in winter. The increase in fluctuations further intensifies in the oldest cohort. In addition, we observe an intermediary peak in mortality during summer among the oldest men.

Women show a more complicated pattern than men. In Cohort IV, they have a decreasing trend in seasonality from 60-69 years to 70-79 years. Also the comparison of 70-79 year olds with 80-89 year olds in Cohort III yields the same result: either there was almost no change at all (winter) or the estimates became smaller (spring and fall). Cohorts I and II, however, show an obvious increase of seasonality in mortality with age. The increase from octagenarians to nonagenarians amounts to 5 percentage point in winter (1.13 to 1.18), 9 percentage points in spring (1.10 to 1.21), and 5 percentage points in fall (1.10 to 1.15). By comparing 90-94 year old with 95-99 year old women in Cohort I, the odds-ratios increase with age from 1.31 to 1.58 in winter, 1.09 to 1.64 in spring, and 1.01 to 1.57 in fall. Contrary to men, we do not observe a summer peak among women.

4 Discussion

Before discussing the results, it is useful to briefly point out the advantages of the study. Some of the potential shortcomings will be discussed at the end of this section. The design of our data-set makes our analysis unique. The major drawback of previous longitudinal studies was the non-availability of the exact risk-set. Either researchers used deaths counts (e.g. [17, 54]) or interpolated data to obtain the specific population at risk (e.g. [34, 39]). To our knowledge, only one previous longitudinal study used exact exposures and occurrences of events [56]. However, our analysis has the advantage that it is based on a sample of the whole population and is, therefore, not
Figure 5: Odds-Ratios for Seasonal Mortality by Sex and Age-Group Part I
(Reference Group: Summer)
Figure 6: Odds-Ratios for Seasonal Mortality by Sex and Age-Group Part II
(Reference Group: Summer)

Cohort III
(Born April 1898 - March 1908)

Cohort IV
(Born April 1908 - March 1918)
restricted to one occupation (civil servants) and one sex (men) [56].

We found that mortality in Denmark follows the well-known seasonal pattern of countries in the Northern Hemisphere. Denmark is experiencing 2,612 winter excess deaths annually, which corresponds to a proportion of 6.69 percent of all deaths. Denmark is on level terms with the US in 1978 [19] and fares better compared to some other European countries like Ireland (14.6%) and Portugal (13.7%) during the period 1976–83. However, a comparison [41] with its neighboring countries in the Baltic like West Germany (5.4%), Sweden (5.4%), Norway (3.9%), and Finland (3.8%) during 1976–84 lets us conclude that further progress can be made in Denmark in reducing the annual cold-related death toll.

It is open to discussion how large the effect of saving lives in winter on mortality in general actually is. Two extreme opinions are imaginable. The true effect will lie somewhere in between: either one assumes that the subject is in a frail condition and would have died anyway relatively soon. The other assumption would be that the individual does not differ in his/her robustness from the rest of the population. Saving a life in the latter case would be “perfect repair” in terms of reliability engineering. Further analysis on the actual causes of death could shed some light on this question. If accidents and infectious diseases dominated the seasonal fluctuations, the overall effect could rather tend towards the “perfect repair extreme”. Contrastingly, if chronic diseases mainly shaped the seasonal pattern, the effect on reducing overall mortality would either be relatively small or it would require more efforts in medicine, public health and general living improvements to obtain the same effect as in the other case where relatively inexpensive interventions like vaccinations may result in remarkable improvements.

We find that age plays an important role for seasonal mortality. As our cohorts were constructed with a 10-year-age-difference between successive cohorts, we could identify an increase with age in the amplitude of the monthly mortality risks. But not only the heights of the fluctuations were increasing. Also the seasonal pattern changed. We can make a broad distinction between the relatively older cohorts (Cohorts I and II) on the one hand and the younger cohorts (Cohorts III and IV) on the other hand. In the older cohorts, the trough is restricted to the three warmest months (July-September). Throughout the remainder of the year, excess mortality is relatively high. The younger group, conversely, shows a different pattern. The trough ranges approximately from spring until early fall. Peaks in mortality can only be found in the coldest months. This result may indicate a change in the sensitivity towards environmental hazards with age: in the two younger cohorts, only the extreme cold weather proved to be dangerous for the Danish. If we look, though, at the two older cohorts, we can see that anything else but
summer climate seems to be perilous for survival for women.

The intermediary summer peak for men suggests that the oldest men are affected by extreme climatic conditions in general: cold weather in winter as well as hot spells during summer. We suspect that this changing pattern has been missed by previous studies either because of the limited age-range they investigated or by an underrepresentation of people at very advanced ages.

The slight mortality dip in all cohorts in February suggests a similar explanation as hypothesized for historical English populations by Oeppen [45]: while people dying in January rather die of immediate causes of the cold climate, deaths in March are due to the accumulation of detrimental effects during the cold season.

Our data help explaining the surprising result of previous studies which found no sex differences in relation to seasonal mortality [14, 17, 43, 62]: we find the same result — but only for the youngest cohort (Cohort IV). In the remaining three cohorts women and men differ to a large extent. We can therefore claim that the limited age-range used in previous studies (for instance: 65–74 years in the Eurowinter Study) is too narrow to conclude that the susceptibility towards cold-related hazards does not differ between women and men.

In Cohorts III and II, which are on average about 10 and 20 years older than Cohort IV, seasonal fluctuations are considerably larger for men than for women. This suggests that men at advanced ages are more susceptible to environmental hazards than women. At first sight, the oldest cohort (Cohort I) seems to contradict this finding: women’s fluctuations outstrip men’s oscillations. To interpret this, we should keep in mind that changes in population parameters can be caused by three different forces [58]. The first explanation refers to the small sample size of our data for the logistic regression model. A direct effect (second potential explanation) assumes that men have become weak to such an extent that any unfavorable conditions may act as lethal. We find support for this explanation in the intermediary summer peak indicating heat-related mortality. This follows a previous finding for Texas [18]. Among all men, the oldest displayed the highest death rates during heat spells. However, Mackenbach and his colleagues [35] found contradictory evidence: women’s excess mortality during hot periods is higher than men’s. Additionally, age “does not appear to be consistently related to excess mortality at high outside temperatures” [35, p. 1298]. Due to these opposing views in the literature, we suppose that our finding may be the outcome of a compositional effect, too (third possible explanation): as the frail tend to die earlier, the male survivors in the oldest cohort (Cohort I) are a selected subsample of their initial birth cohort being more robust on average.
Among men, seasonal mortality fluctuations increase with age. We suggest two explanations for the stationary or even decreasing trend for women in the two younger cohorts: either women actually have increasing seasonal fluctuations in mortality but this trend is offset by beneficial period-effects. However, this explanation seems to be less likely as it implies that some secular events had positive consequences for women in those age-ranges but neither at later ages nor for men. We rather support the opposing idea that the aging process in the younger observed age-groups does not affect the relatively strong resistance of women towards changing environmental hazards. At later ages women also face an increasing susceptibility with age. Our interpretation therefore is that women as well as men show larger seasonal fluctuations in mortality as they age. The main differences are that women’s susceptibility starts at later ages and is restricted to cold temperatures. Men, on the contrary, show increases with age already at younger ages. In the oldest observed age-categories, men have become not only susceptible to cold outside temperatures but to any unfavourable climatic conditions (e.g. summer heat).

The data of our analysis can be criticized in some respects. First, we rarely found significant values for the analysis by age. Our remaining tables and figures, however, give us several indications that the general trend we observed is correct and the lack of significant values can be primarily attributed to our sample size. Secondly, causes of death were not available. Their inclusion would have helped us, for example, to verify whether the dip in mortality in February is actually due to the different mechanisms mentioned above. Further analyses on seasonal mortality should aim to incorporate the whole population, causes of death and risk factors such as housing and deprivation already mentioned in the literature [4, 8, 14, 27, 29, 30, 31, 46, 52, 60]. Using those covariates in a population-based longitudinal study with exact risk-set will improve our understanding of the mechanisms regulating seasonal mortality that are still discussed ambiguously.

5 Conclusion

This study analyzed seasonal mortality in Denmark which is relatively high compared to its neighboring countries. We have shown that the previous assessment that women and men do not differ with respect to seasonal fluctuations in mortality tends to be an oversimplification. Men seem to be more susceptible to hazardous environmental conditions than women. This is what one could expect from the generally lower female mortality rates throughout the life-course. We found evidence that seasonality increases with age.
However, the pattern varies between women and men. Women’s seasonal fluctuations start increasing at later ages than men’s. Compared to women, men seem to be susceptible not only to cold weather but also to extreme hot climatic conditions.

Tackling seasonal mortality will gain further relevance for public health researchers from two directions. First, the chances to survive into ages where seasonal fluctuations in mortality reach remarkable levels are increasing [24, 57]. Secondly, the vast amount of literature on risk factors and the smaller fluctuations in regions with colder climate let us conclude that the annual amount of excess mortality may be reducible by preventive health and safety measures [54].
References


Human Life-Table Database, *Data by country: Denmark*. Contributions from Väinö Kannisto and Danmarks Statistik, accessible online at: http://www.lifetable.de, April 2003.


Table 1: Cox-Proportional-Hazards Model and Binary Logistic Regression yield the same results in practice. An example from estimating the effect of current month on the chance of dying for Cohort I.

<table>
<thead>
<tr>
<th>Month of Death</th>
<th>Cox-Proportional-Hazards (Continuous Time)</th>
<th>Binary Logistic Regression (Discrete Time)</th>
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<tr>
<td>January</td>
<td>0.2516</td>
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