Adverse early life conditions may have lasting effects on old-age health and mortality. Some consider reductions in early life disease exposure to be a primary driver of historical mortality declines. Although the precise mechanisms linking early disease exposure to poor adult health remain unclear, numerous pathways have been postulated including those relating to fetal undernutrition and dysregulation of immune function.

In animal models, experimental evidence suggests a negative causal effect of early disease exposure on later health. For humans, historical epidemics have been used to study the effects of early disease exposure on later health. These studies often find that those born around the time of an epidemic exhibit worse adult health and mortality than do neighboring cohorts. However, the causes of death contributing to the excess mortality are not known. Moreover, research on early exposure to the deadliest epidemic of the 20th century—the 1918 influenza pandemic—is mixed, showing increased cardiovascular disease prevalence and lower socioeconomic attainment, but no long-term mortality effects.

We investigated whether US cohorts with early exposure to the 1918 pandemic experience differential mortality at old age compared with neighboring cohorts. The 1918 pandemic, caused by the influenza A virus (subtype H1N1), arrived in the United States in 3 waves. During the first wave, which began in March 1918 and was completed by July 1918, incidence rates were high, but mortality was only slightly elevated. The second and the deadliest wave began in September 1918 and lasted until the end of the year. The third wave, with a mortality impact between those of the first 2 waves, occurred from January 1919 to March 1919. Approximately 30% of the US population was infected and about 0.5% of the population died because of the pandemic, mostly from pneumonia. Excess mortality had an unusual pattern as those aged 20 to 40 years were affected particularly strongly.

The advantages of focusing on the 1918 pandemic are threefold. First, the pandemic arrived unexpectedly and lasted for only a short period, allowing treatment of the pandemic as a “natural experiment” wherein cohorts born months apart experienced different exposures but were otherwise compositionally similar in terms of other childhood characteristics and environmental conditions. Moreover, the exposed and nonexposed cohorts were born in a narrow enough time interval that timing of birth is not systematically linked to subsequent differences in the adult environment. Second, in contrast to older epidemics, existing data permit cause-of-death analyses. Third, although food shortages and disease tended to co-occur in historical populations, the 1918 pandemic allows focusing on disease because there were no generalized food shortages in the United States during the pandemic.

Objectives. We sought to analyze how early exposure to the 1918 influenza pandemic is associated with old-age mortality by cause of death.

Methods. We analyzed the National Health Interview Survey (n = 81,571; follow-up 1989–2006; 43,808 deaths) and used year and quarter of birth to assess timing of pandemic exposure. We used Cox proportional and Fine-Gray competing hazard models for all-cause and cause-specific mortality, respectively.

Results. Cohorts born during pandemic peaks had excess all-cause mortality attributed to increased noncancer mortality. We found evidence for a trade-off between noncancer and cancer causes: cohorts with high noncancer mortality had low cancer mortality, and vice versa.

Conclusions. Early disease exposure increases old-age mortality through noncancer causes, which include respiratory and cardiovascular diseases, and may trigger a trade-off in the risk of cancer and noncancer causes. Potential mechanisms include inflammation or apoptosis. The findings contribute to our understanding of the causes of death behind the early disease exposure–later mortality association. The cancer–noncancer trade-off is potentially important for understanding the mechanisms behind these associations.
METHODS

We used data from the National Health Interview Survey (NHIS), an annual cross-sectional survey of the US noninstitutionalized population. We used the 1989–2004 surveys because we focused on US-born people and country of birth is not known before 1989, and death linkages are currently not available for surveys conducted after 2004. The 1989–2004 surveys are linked with the National Death Index through December 31, 2006, in the NHIS–Linked Mortality Files. These data allow for mortality analysis by year and quarter of birth. The mortality period assessed (1989–2006) falls under both the International Classification of Diseases, Ninth Revision (ICD-9; 1979–1998) and International Classification of Diseases, Tenth Revision (ICD-10; 1999–2006) guidelines for US cause-of-death coding. We used a consistent set of 113 underlying cause-of-death recodes provided in the NHIS–Linked Mortality Files, with deaths occurring before 1999 recoded into comparable ICD-10 groupings by the National Center for Health Statistics. We analyzed all-cause mortality and mortality by 3 major cause-of-death categories: (1) cardiovascular diseases including heart disease, cerebrovascular diseases, and diseases of the circulatory system (ICD-10: I00–I78; hereafter: CVD); (2) malignant neoplasms excluding those of the trachea, bronchus, and lung (ICD-10: C00–C97 excluding C33–C34), and (3) all other causes, among which respiratory diseases is the largest category.

Birth Cohorts and Exposure Timing

We included the 1913–1924 birth cohorts. These included those born during the 2 years spanning the pandemic (1918–1919) and 5 cohorts born before (1913–1917) and after (1920–1924) the pandemic. The sample size was 81,571 persons with 43,808 deaths.

We grouped the cohorts into 5 categories according to exposure timing:

1. the 1913q1–1917q2 cohorts (“q” refers to quarter of year) were exposed after their first birthday;
2. the 1917q3–1918q1 cohorts were exposed during first year of life but not at birth or in gestation;
3. the 1918q2, 1918q4, and 1919q1 cohorts were exposed in the third trimester and at birth, as each of the cohorts was born during one of the pandemic waves;
4. the 1918q3 and 1919q2–q4 cohorts were each exposed early in gestation (first or second trimester) but not at birth; and
5. the 1920q1–1924q4 cohorts were not directly exposed.

These 5 categories are not an exhaustive description of the exposure experience. For example, in group 3, the 1919q1 cohort was exposed not only late in gestation and at birth (to the third wave), but also in the second trimester (to the second wave). Likewise, in group 4, the 1918q3 cohort had second-trimester exposure to the first wave and postbirth exposure to waves 2 and 3. However, this categorization provides a useful map describing which cohorts were exposed (1) after first year of life, (2) during the first year of life, (3) late in gestation and at birth, (4) early in gestation but not at birth, or (5) were not directly exposed. A supplementary table (available as a supplement to this article at http://www.ajph.org) illustrates the exposure timing by birth cohort.

The NHIS samples from the noninstitutionalized population, but death linkages capture deaths to those institutionalized after being surveyed. Excluding the institutionalized population at baseline is unlikely to bias our results because the fraction of institutionalized population in the relevant age groups and periods is small; for example, in 1990, the fraction of institutionalized population aged 65 to 74 years was less than 2%. We combined the nonexposed 1920q1–1924q4 cohorts and treated them as the omitted reference category. We also combined the cohorts exposed after their first birthday (1913q1–1917q2) into a single category. Among those exposed during gestation or before first birthday (1917q3–1919q4), we used quarter and year of birth indicators to capture the exposure timing.

We controlled for age and age squared at baseline to capture nonlinearities in the association between log-mortality and age. We controlled for a cohort trend in mortality by including a continuous birth year variable. We adjusted all models for gender. We excluded lung cancer deaths from the cancer analyses because lung cancer risk is largely determined by lifetime cigarette smoking and preliminary analyses reveal that smoking behaviors did not vary across the birth cohorts examined. Removing a major cause of death that is mostly determined by adult behavior (rather than early life exposures) allows for a more accurate analysis of the remaining causes of death. We describe results from sensitivity analyses that include deaths from lung cancer, control for season of birth, and other robustness checks.

Additional methodological details are available as a supplement to this article at http://www.ajph.org. We conducted all analyses by using Stata/SE version 11.2 (StataCorp LP, College Station, TX).

RESULTS

Table 1 describes the data. The sample size was 81,571, average age at baseline was 74.5 years, and average follow-up was 9.0 years. The majority of the sample were women (47,583 vs 33,988 persons). During the follow-up, 43,808 (53.7%) persons died, 34,411 of noncancer and 9397 of cancer causes. Among noncancer causes, cardiovascular disease was the most common (19,382 cases). For the key cohorts 1917q3–1919q4 exposed in utero, at birth, or during the first year of life, the
The fraction of dead was largest (70%: 16.796 of 23.880) for the 1913q1–1917q2 cohort that was exposed after the first birthday, and lowest (43%) for the nonexposed 1920q1–1924q4 cohort. Among those that were exposed during the first year of life, in the third trimester and at birth, or in both or either the first or second trimester, the fractions of dead were 59%, 58%, and 52%, respectively (calculations not shown). These differences show that cohorts that were born earlier in time had higher mortality than those that were born later, and tentatively suggest that the cohorts that were exposed in the third trimester or at birth may have higher mortality than those that were exposed early in gestation. However, these differences also reflect the age differences of the respective cohorts when entering our study, which we controlled for in the regressions.

Table 2 shows HRs for all-cause mortality and for causes of death by birth cohort and exposure timing. The 1918q2 and 1919q1 cohorts, which were exposed in third trimester and at birth have excess all-cause mortality, the HRs being 1.08 and 1.09, respectively (both \( P < 0.05 \)). The excess mortality is fully attributable to noncancer causes, among which cardiovascular and respiratory diseases are the major causes of death. For the 1919q3 cohort, which was exposed to the second and third waves early in gestation, we observed decreased noncancer, in particular CVD, mortality (HR = 0.87; \( P < 0.05 \)) and increased cancer mortality (HR = 1.26; \( P < 0.01 \)).

Thus, of the 3 cohorts that were exposed late in gestation and at birth, 2, the 1918q2, 1919q1 experienced excess old-age mortality. The exception is the 1918q4 cohort that was born during the second wave and was exposed to the third wave soon after birth.

Our results additionally suggest that mortality HRs for cancer and noncancer causes are negatively correlated. For the cohorts that had increased noncancer mortality (1918q2 and 1919q1), the point estimates for cancer mortality were below 1.00. For other cohorts, cancer HR was above 1.00. In addition, for the only cohort with significant excess cancer mortality (1919q3), CVD mortality was significantly decreased.

Table 3 shows the HR correlations between cancer and other causes of death for the 1917q3–1919q4 cohorts. For comparison, we also show cohorts exposed after first birthday (1913q1–1917q2) and nonexposed cohorts (1920q1–1924q4). For the 1917q3–1919q4 cohorts the correlations were calculated from the HRs of Table 2. For other cohorts we estimated additional mortality regressions by birth quarter and year and calculated the correlations (see material available as a supplement to this article at http://www.ajph.org). For the 1917q3–1919q4 cohorts the cancer–noncancer HR correlation ranged from −0.70 to −0.87 \( (P < 0.05) \). For other cohorts the correlations were small and nonsignificant. Thus, we observed the cancer–noncancer trade-off only for the cohorts that were exposed early in life.

Figure 1 illustrates the mortality trade-offs for the 1917q3–1919q4 cohorts by comparing the HRs for cancer with all noncancer mortality. The cancer HRs were smaller and nonsignificant for the only cohort with significant excess cancer mortality (1918q2 and 1919q1), the point estimates for cancer mortality were below 1.00. For other cohorts, cancer HR was above 1.00. In addition, for the only cohort with significant excess cancer mortality (1919q3), CVD mortality was significantly decreased.

Sensitivity checks are available as supplementary data at http://www.ajph.org. First, we included season of birth controls; estimated the cancer results with lung cancer; included controls for race/ethnicity and education; expanded or narrowed the cohort window from 1913–1924 to 1912–1925 or 1915–1922; and used the Cox model for cause-specific mortality. The key results did not change. Second, we estimated the results by gender. The statistical power decreased but the gender-specific results were qualitatively similar to our main results.
We studied the association between early disease exposure and old-age mortality by using the 1918 influenza pandemic as an exogenous shock. Previous research on early exposure to the 1918 pandemic has found no long-term association with mortality but has relied on annual cohorts, which combines long-term association with mortality but has exposure to the 1918 pandemic has found no exogenous shock. Previous research on early disease exposure and old-age mortality by using the 1918 influenza pandemic as an exogenous shock.

Three cohorts were exposed to the pandemic late in gestation and at birth (1918q2, 1918q4, 1919q1). Of these, the 1918q2 and 1919q1 cohorts had 8% to 9% excess old-age all-cause mortality, corresponding to 0.6 years of decreased life expectancy at age 70 years in a population with life expectancy at birth of 75 years. This estimate should be a lower bound. Only one third of the US population was infected, and we are unable to ascertain the infection status of respondents. The excess mortality among the infected (or whose mothers were infected) is likely higher. In addition, selective mortality before the observation period may further bias the estimates downward as approximately a quarter of those born in the 1910s died before the entry age 63 years.

The fact that we did not observe excess mortality for the 1918q4 cohort could also be attributable to earlier life selection. The 1918q4 cohort was exposed to the first wave early in gestation, a period during which spontaneous abortions were most common, and which is thought to be associated with increased mortality at young and middle age, born during the deadliest second wave, and continually exposed to the second and third waves up to 3 to 6 months of age. The 1918q2 and 1919q1 cohorts had less intense early gestation and postbirth exposures. Thus, stronger pre- and postbirth selection may bias the estimates for the 1918q4 cohort downward more so than that of the 1918q2 and 1919q1 cohorts, resulting in a null finding that is attributable to selection. Other estimates, including the observed increase in cancer mortality for the 1919q1 cohort, may also be conservative because of earlier-life selective mortality.

We found decreased noncancer mortality, in particular CVD mortality, for the 1919q3 cohort that was exposed to the deadly second wave of the pandemic. The excess mortality among the infected (or whose mothers were infected) is likely higher. In addition, selective mortality before the observation period may further bias the estimates downward as approximately a quarter of those born in the 1910s died before the entry age 63 years.

The table below shows the hazard ratios for all-cause mortality and cause-specific subdistribution hazard ratios at ages 63 to 95 years by birth cohort and timing of exposure to the 1918 influenza pandemic and cause of death: National Health Interview Survey, 1913–1924 cohorts.

<table>
<thead>
<tr>
<th>Birth cohort, exposure timing</th>
<th>All Causes, HR (95% CI)</th>
<th>CVD, HR (95% CI)</th>
<th>Other, HR (95% CI)</th>
<th>Cancer, HR (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1917q3, 1st yr of life</td>
<td>0.99 (0.92, 1.07)</td>
<td>0.90 (0.81, 1.01)</td>
<td>1.11 (0.98, 1.25)</td>
<td>1.07 (0.88, 1.29)</td>
</tr>
<tr>
<td>1917q4, 1st yr of life</td>
<td>1.06 (0.99, 1.15)</td>
<td>1.00 (0.89, 1.12)</td>
<td>1.07 (0.94, 1.21)</td>
<td>1.07 (0.88, 1.31)</td>
</tr>
<tr>
<td>1918q1, 1st yr of life</td>
<td>1.03 (0.96, 1.11)</td>
<td>1.05 (0.94, 1.17)</td>
<td>1.10 (0.98, 1.25)</td>
<td>1.04 (0.86, 1.26)</td>
</tr>
<tr>
<td>1918q2, at birth and 3rd trimester</td>
<td>1.08* (1.00, 1.16)</td>
<td>1.15*** (1.06, 1.25)</td>
<td>1.02 (0.91, 1.14)</td>
<td>1.28*** (1.14, 1.45)</td>
</tr>
<tr>
<td>1918q3, 1st and 2nd trimester</td>
<td>0.97 (0.90, 1.05)</td>
<td>0.95 (0.87, 1.03)</td>
<td>0.99 (0.89, 1.10)</td>
<td>0.93 (0.81, 1.06)</td>
</tr>
<tr>
<td>1918q4, at birth and 3rd trimester</td>
<td>1.02 (0.95, 1.10)</td>
<td>1.00 (0.92, 1.09)</td>
<td>0.97 (0.87, 1.09)</td>
<td>1.04 (0.91, 1.18)</td>
</tr>
<tr>
<td>1919q1, at birth and 3rd trimester</td>
<td>1.09* (1.01, 1.17)</td>
<td>1.13** (1.05, 1.23)</td>
<td>1.07 (0.96, 1.19)</td>
<td>1.20** (1.07, 1.34)</td>
</tr>
<tr>
<td>1919q2, 1st and 2nd trimester</td>
<td>1.05 (0.97, 1.13)</td>
<td>1.05 (0.97, 1.14)</td>
<td>1.06 (0.95, 1.19)</td>
<td>1.05 (0.93, 1.19)</td>
</tr>
<tr>
<td>1919q3, 1st and 2nd trimester</td>
<td>0.97 (0.90, 1.04)</td>
<td>0.92* (0.84, 1.00)</td>
<td>0.87* (0.78, 0.98)</td>
<td>1.02 (0.90, 1.15)</td>
</tr>
<tr>
<td>1919q4, 1st and 2nd trimester</td>
<td>1.00 (0.93, 1.07)</td>
<td>0.99 (0.92, 1.07)</td>
<td>0.93 (0.84, 1.03)</td>
<td>1.08 (0.96, 1.21)</td>
</tr>
<tr>
<td>1920q1-1924q4, not exposed (Ref)</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
</tr>
</tbody>
</table>

Note. CI = confidence interval; CVD = cardiovascular disease; HR = hazard ratio. We estimated results for all-cause mortality with the Cox proportional hazards model, and results for cause-specific mortality with the Fine-Gray competing risks model. All models controlled for age and age squared at baseline, gender, and linear trend in birth year.

*Excludes lung cancer.

**P < .05; ***P < .01; ****P < .001.
and third waves in early gestation. Selection may play a role here as well. Early gestation shocks are particularly likely to result in miscarriage, and one study suggests that the risk of miscarriage or fetal death was elevated among those that were exposed to the 1918 pandemic.1 Several pathways could link late gestation or postnatal disease exposure with later noncancer mortality. Developing organisms adapt to environmental signals.28 If early life environment is different from that experienced later in life, the adaptations may be harmful.9,31 For example, disease exposure may cause nutritional deprivation and permanent changes in glucose–insulin metabolism. Such adaptation might be helpful in a nutritionally deprived environment during later life but increase CVD and diabetes risk in an affluent environment.32,33 Early disease exposure may also prime the immune system to be constantly alert, leading to chronic inflammation,34 which increases CVD risk.9,34 As the third trimester is critical for lung maturation,35 late gestation exposure may increase respiratory disease mortality. Prenatal exposure may also result in preterm birth,36 increasing the risk of several health conditions.37

Our study gains leverage from jointly analyzing all 3 waves of the pandemic. Other studies have focused on the most virulent second wave, possibly because the mortality impact of the third wave was milder, and the impact, or even existence, of the first wave is not always recognized. Epidemiological studies, however, confirm that in the spring of 1918 an influenza wave with a signature W-shaped excess mortality pattern hit New York City,28 US Army camps,39 and Mexico.40 These studies provide strong evidence for a “herald” spring wave in North America.

Previous research on adverse early life exposures suggests various critical periods of exposure. Studies on the 1918 pandemic have found that late gestation or at-birth exposures are most important for later life outcomes.1,4 Analyses using historical data and population-level mortality rates as proxies for disease exposure have found that exposures at birth and during the first year of life are most important.2,29 However, some famine studies suggest that first or second trimester exposures are most important.6,27,41 Our finding that late gestation and at-birth exposures are important for later mortality is consistent with existing literature on early disease exposure and adult health and mortality. With respect to nutrition, it is possible that the mechanisms and critical periods are different.

**Correlation Between Causes of Death**

Collectively, our findings suggest that subtle differences in exposure timing may have important but complex implications for later mortality. Early disease exposure may trigger processes that increase late noncancer mortality but are protective against cancer, and vice versa. However, a negative correlation in HRs could also occur if the risk is increased for both cancer and noncancer causes: if the risks for both CVD and cancer are elevated, but people tend to systematically die from one cause before the other occurs, a negative correlation in HRs may arise. Unfortunately, without strong assumptions, it is not possible to test whether the negative correlation in HRs is driven by negative or positive correlations in the individual-level risks.42 We can nevertheless speculate on the likelihood that the negative HR correlation is caused by a positive versus negative correlation in the individual-level risks.42 We can nevertheless speculate on the likelihood that the negative HR correlation is caused by a positive versus negative correlation in the individual-level risks. Our simulations suggest that moderate positive correlation in the individual-level risks may lead to a moderately positive or negative HR correlation. Moderate negative correlation in the individual-level risk may lead to a moderate to large negative HR correlation.43 We observed a correlation of –0.87 (P<.01) between cancer and noncancer HRs. We consider this correlation to be strong, suggesting a negative rather than positive correlation in the individual-level risks.

Similar trade-offs have been documented in other settings. For example, heart disease,
diabetes, and Alzheimer’s disease have been associated with decreased risk for 1 or more cancers. Low birth weight is positively associated with cardiovascular disease but negatively associated with several cancers. Androgen deprivation therapy treats prostate cancer but may increase cardiovascular disease and diabetes mortality. Our study is the first to demonstrate that early disease exposure may trigger a similar trade-off.

Inflammation and apoptosis may help in understanding the trade-off. Although it is beyond the scope of this study to test these explanations, they provide a plausible mechanism through which early disease exposure may have differential effects by cause of death. Chronic inflammation may be triggered by early disease exposure. Inflammation, in turn, is linked with apoptosis and cellular senescence, which are protective against cancer but may predispose to other aging-related diseases such as ischemic heart disease and neurodegenerative diseases. In particular, apoptosis and cellular senescence are regulated by the protein p53. In unstressed cells, p53 levels are low. DNA damage activates p53. Activated p53 may initiate cell cycle arrest, which prevents damaged DNA from replicating or allows DNA repair. When DNA is damaged beyond repair, p53 may initiate apoptosis. These processes control carcinogenesis. Indeed, increasing p53 levels may decrease cancer risk but also accelerate other aging-related diseases. Previous research documents that inflammation may influence the functioning of p53, but has not considered the role of early life exposures. Our findings suggest that early disease exposure may permanently imprint physiological processes that may result in a trade-off between cancer and other causes of death, potentially through altered functioning of p53.

Conclusions

Using the 1918 pandemic as an exogenous shock we show that early disease exposure increases old-age mortality through noncancer causes and may trigger a trade-off in the risk for cancer and noncancer causes. The findings enhance our understanding of the causes of death that contribute to the association between disease exposure early in life and adult mortality. Our study also provides suggestive evidence that early disease exposure may trigger a trade-off between cancer and other causes of death.

**Note.** Other noncancer causes include all causes other than cardiovascular disease. The sample size was n = 81,571.

**FIGURE 1—Mortality hazard ratios at ages 63–95 years by birth quarter among cohorts exposed to the 1918 influenza pandemic in utero or during first year of life for cancer and (a) all noncancer causes, (b) cardiovascular disease, and (c) other noncancer causes: National Health Interview Survey.**
evidence on why earlier research, which has only considered annual birth cohorts, has not found a mortality association for exposure to the 1918 pandemic. Early disease exposure has complex effects on later-life health, so that the magnitude and even the sign of the effect may critically depend on the timing of exposure and on the cause of death analyzed. Identification of these patterns is not possible in analyses using annual birth cohorts, wherein cohorts exposed at different stages of gestation are combined. The finding of an early disease exposure that triggers a trade-off between cancer and other causes may help further elucidate our understanding of the origins of certain cancers.

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Contributors
M. Myrskylä and N. K. Mehta conceptualized and designed the study with input from V. W. Chang. M. Myrskylä and N. K. Mehta conducted the statistical analysis. All authors contributed to the interpretation of the findings and to the writing of the article.

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Human Participant Protection
No protocol approval was required because data were obtained from secondary sources.

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