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SUPPLEMENTARY MATERIALS
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Supplementary Text
Figs. S1 to S10
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References (34-42)

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## DEMOGRAPHY

# Inequality in mortality decreased among the young while increasing for older adults, 1990-2010 

J. Currie ${ }^{1,2,3 *}$ and H. Schwandt ${ }^{3,4,5}$


#### Abstract

Many recent studies point to increasing inequality in mortality in the United States over the past 20 years. These studies often use mortality rates in middle and old age. We used poverty level rankings of groups of U.S. counties as a basis for analyzing inequality in mortality for all age groups in 1990, 2000, and 2010. Consistent with previous studies, we found increasing inequality in mortality at older ages. For children and young adults below age 20, however, we found strong mortality improvements that were most pronounced in poorer counties, implying a strong decrease in mortality inequality. These younger cohorts will form the future adult U.S. population, so this research suggests that inequality in old-age mortality is likely to decline.


Poorer people tend to have shorter lives and are more likely to die than richer people at all ages. Understanding the evolution of these inequalities in mortality is a central concern of economists, policy-makers, and the public. Not surprisingly, a great deal of highly publicized research has investigated changes in inequality in life expectancy and mortality in the United States over the past 20 years. A preponderance of the existing evidence points to alarming increases in inequality in mortality over this time period (1-16). Some studies investigating mortality trends across educational groups and geographic areas argue not only that inequality in life expectancy is widening, but that overall life expectancy is actually falling among the most disadvantaged groups (11-13).
However, much of the recent literature focuses on adults, and in particular on life expectancy at age 40 or 50 , exploiting rich data sets that link individuals' career earnings to deaths at older ages (1-8). By construction, these analyses omit children, teens, and young adults. A second strand of research analyzes demographic subgroups defined by education, location, and/or race (9-16).

[^0]These studies typically focus on overall life expectancy at birth.
Life expectancy at birth is a summary measure that collapses all of the age-specific mortality rates observed in a given year (and in a certain demographic subgroup) into a single number. It provides information about how long a cohort of newborns can expect to live, under the assumption that the age-specific mortality rates observed in that given year remain constant into the future. This assumption is unlikely to hold in the United States, given that mortality rates at all ages have been continuously changing (mostly improving) over the past century (17).

Changes in infant and childhood mortality have been shown to be important predictors of a cohort's health and mortality at later ages, and such data may therefore be more informative about the development of future death rates for the current young. Moreover, mortality at young ages is considered a sensitive indicator of social conditions because it responds relatively quickly to changes affecting the entire population, whereas old-age mortality is partly determined by conditions in the past. The infant mortality rate has been shown to be an important indicator of health for whole populations and one that is highly correlated with more complex measures such as disability-adjusted life expectancy (18).

Therefore, to study how inequality in mortality changes over time, it is important to understand age-specific mortality trends and in particular those at younger ages. Life expectancy at birth
masks potential differences in age-specific trends, and the measure is also dominated by changes in old-age mortality because that is when most deaths occur. A recent study by Case and Deaton (19) highlights the relevance of examining age-specific mortality rates: They document increases in middle-age mortality for non-Hispanic whites, a striking development that would not be detectable in overall life expectancy at birth.

We followed an empirical approach, based on placing counties into groups, that allows us to analyze trends in age-specific mortality while taking into account population shifts across groups. We ranked all counties in 1990, 2000, and 2010 by their poverty level and then divided them into 20 groups, each representing roughly $5 \%$ of the overall U.S. population (fig. S1 and table S 1 ). This enables us to compare, for example, the $5 \%$ of the population living in the poorest counties in 1990 with the $5 \%$ of the population living in the richest counties in 1990, and analyze how the mortality differences between these groups change over time. We refer to the county groups with the highest fractions of their populations in poverty as the poorest counties, and those with the lowest fractions of their populations in poverty as the richest counties.

Our approach reassigns county groups in 1990, 2000, and 2010 to adjust for changes in county ranking and population size. That is, we compare the poorest counties representing $5 \%$ of the population in 1990 with the poorest counties representing $5 \%$ of the population in 2010, even if they are not exactly the same counties. The advantages of this procedure and a comparison with other approaches are discussed below. Our county grouping approach is similar to that of Singh and Siahpush (9), who investigated life expectancy trends ranking U.S. counties by a deprivation index (comprising a set of county characteristics) up to 2001. Our approach differs from theirs in that they did not analyze age-specific mortality, analyzed data only up to 2001, and did not reorder county groups over time.
Mortality rates were constructed at the levels of county group, gender, and age by dividing death counts from the U.S. Vital Statistics by population counts from the decennial Census. We focused on 3 -year mortality rates for Census years 1990, 2000, and 2010, based on a total of 21,175,011 deaths. Life expectancy was calculated by constructing a life table based on 19 age groups (see the supplement for additional details regarding the construction of mortality rates and life expectancy). Socioeconomic county characteristics, including poverty rate, median and per capita income, and percentage of high school
dropouts, were taken from the Census in 1990 and 2000. For 2010 we used the 2008-2012 American Community Survey (ACS), which replaced the long form of the Census for 2010. Table S 1 reports socioeconomic characteristics for the 20 county groups. The county group with the lowest fraction living in poverty had an average poverty rate of $3.75 \%$ and a median income (averaged across counties) of $\$ 62,445$ in 1990. The comparable 2010 figures are 5.58\% in poverty and a $\$ 62,752$ median income. The county group with the highest fraction living in poverty had a $30.47 \%$ poverty rate and a median income of \$23,595 in 1990. Comparable 2010 figures were $28.30 \%$ in poverty and a median income of $\$ 25,404$.

We start with the analysis of overall life expectancy at birth, so as to make a better comparison with the strand of previous literature that has focused on this measure. Figure 1A plots male and female life expectancies at birth for the 20 county groups in 1990, 2000, and 2010 (see table S2 for numerical values and standard deviations). Standard deviations are within $0.1 \%$ of the estimates, and $95 \%$ confidence intervals (CIs) would be fully covered by the estimate markers if plotted in Fig. 1A. In addition to the plotted life expectancy values, we have drawn a linear regression line through the 20 dots representing each year (the line for 2000 omits symbols to reduce clutter). A steeper slope of the regression line indicates greater inequality in life expectancy. If there was no difference in life expectancy between richer and
poorer county groups, then the line would be entirely flat.

Figure 1A shows that for men, there is a strong gradient in 1990, with those living in the richest counties enjoying 6.10 additional years of life expectancy relative to those living in the poorest counties ( 74.79 versus 68.70 ). For women, who have greater life expectancy overall, this gap is smaller at 3.01 years ( 80.20 versus 77.19). Between 1990 and 2010, life expectancy at birth increased across the entire poverty spectrum, both for men and for women. For men, the fitted lines in 1990 and 2010 are almost parallel, suggesting that life expectancy increased by similar amounts in rich and poor counties. In fact, residents of the poorest counties gained slightly more with 4.63 additional years, whereas those in the richest county group gained 4.35 years. For women, improvements were stronger for those in the richest county group (3.01 versus 2.06 years), and most of these improvements occurred between 2000 and 2010.

Figure 1B plots the changes in life expectancy between 1990 and 2010. For women, the fitted regression line is downward-sloping ( $P=0.043$ ), indicating increasing inequality in life expectancy over this period. For men, the slope of the regression line is positive but not significantly different from zero ( $P=0.103$ ), consistent with Fig. 1A's suggestion that decreases in mortality were equally distributed across men in rich and poor counties.

Turning to our key innovation-the analysis of age-specific mortality for all ages-Figs. 2 and 3


Fig. 1. Life expectancy at birth by poverty percentile and gender. (A) Average male and female life expectancy at birth by poverty percentile. Each bin represents a group of counties with about 5\% of the overall population. The solid lines provide the fitted regression lines. Higher percentiles refer to higher poverty levels. A steeper slope implies greater inequality in life expectancy at birth. Magnitudes are reported in table S2. (B) Changes in average male and female life expectancy at birth by poverty percentile. The fitted regression line has a slope of $0.0062(P=0.103)$ for men and a slope of -0.0075 ( $P=0.043$ ) for women.
show that the evolution of overall life expectancy at birth masks considerable heterogeneity in trends in mortality rates at different ages. Similar to Fig. 1, each symbol in the figure represents the age-specific 3 -year mortality rate in a bin representing $5 \%$ of the U.S. population, and the bins are ordered by county poverty rates. Mortality rates are plotted for 1990 and 2010 together with a linear regression line; only the line is shown for 2000. Regression lines are upward-sloping because mortality is higher in poorer counties, but, as in Fig. 1, a flattening of the line over time indicates a decrease in inequality. Tables S3 and S4 report standard errors for the mortality rates and tests for a change in the slope of the fitted regression lines.

The first panel in Fig. 2 shows the evolution of 3-year mortality rates for male newborns, which decreased by 4.2 per 1000 in the group of richest counties between 1990 and 2010, from 9.77 (95\% CI, 9.10 to 10.44 ) to 5.53 ( $95 \% \mathrm{CI}, 5.06$ to 6.00 ). However, infant mortality in the group of poorest counties decreased by 8.49 deaths per 1000, which is more than twice as much over the same time period, from 18.28 ( $95 \%$ CI, 17.38 to 19.17) to 9.79 ( $95 \%$ CI, 9.22 to 10.37 ). These strong reductions in mortality in the poorer county groups are reflected in a considerable flattening of the regression line in 2010 relative to 1990 . The slope of the regression line through the group values decreases by more than $50 \%$, and this change is highly significant ( $P<0.001$, table S 3 ). This flattening indicates a marked reduction in inequality in infant mortality.

A similar decline in mortality inequality can be observed up to age 20, although improvements for young children were greatest between 1990 and 2000. For older children, there were also large declines between 2000 and 2010. Looking at older ages, inequality decreased (i.e., the slope of the fitted regression line decreased significantly; see table S 3 for $P$ values) among males up to age 50 . Between ages 50 and 75 there was no significant change in inequality in mortality, but after age 75 , mortality inequality increased significantly among males. It is also striking that for adult men between 20 and 34, there was virtually no improvement in mortality rates between 2000 and 2010.

Figure 3 shows that the patterns are somewhat different for females. As it did for males, female mortality decreased strongly for age groups up to age 19, and these improvements were significantly stronger in the poorest counties, implying that inequality in mortality decreased sharply (see table S 4 for $P$ values of the differences in the slopes). However, although inequality decreased significantly for males until age 50 , this trend is observed for females only up to age 30 . For ages 30 to 45 , there is no significant change in mortality inequality, whereas for all age groups over 45 , inequality in mortality increases.

Turning to the mortality rates themselves, it is noteworthy that there was practically no improvement in mortality among women aged 30 to 45 between 1990 and 2010. This is a striking

development in light of the progress made in other age categories. A further remarkable fact is that mortality rates actually increased in some of the richest counties among females aged 20 to 29. After age 45 , there are mortality declines, but they are larger in the richer county groups, driving the increase in mortality inequality noted above.

The results discussed so far are all based on ranking counties by poverty rates, which is arguably the most relevant measure if one is focusing on differences between the rich and poor. However, several additional measures of socioeconomic status are available at the county level. The age-specific trends in mortality are very similar when ranking counties by these alternative measures, including the share of high school dropouts, median income, and average life expectancy (figs. S2 to S4).
In contrast to many recent analyses of inequality in mortality that focus on life expectancy at middle age, we find overall improvements in life expectancy at birth both in counties with high poverty rates and counties with low poverty rates. However, we argue that life expectancy measures are not (despite their name) intended to be predictive of the number of future years of
life that any particular cohort can expect to attain, and that it is more informative to examine age-specific mortality rates. Our analysis of these rates indicates that inequality in mortality between rich and poor counties has strongly declined among infants, children, and young adults up to age 30 of either gender, as well as among adult males up to age 50 . Among older adults, mortality has continued to decline, although declines are generally greatest in the richest counties, indicating increasing inequality in mortality, which is in line with the literature that has focused on inequality trends at older ages ( $1-8$ ).

Our focus on using county groups to examine inequality has advantages and disadvantages. Unlike subgroups defined by race and education or by individual counties, county groups are large enough to provide precise mortality estimates in age ranges with low mortality. Moreover, the county of residence is consistently reported both in the Vital Statistics and the Census data, which makes mortality rates by county group subject to less measurement error than using other demographic groups that can be constructed with these data sets. For example, education is often missing from death certificates, and education measures
were switched from years of schooling to degrees for some states in the mortality files but not in the Census. Even race is not always consistently reported. For example, the Census introduced multiple race categories in 2000 while the Vital Statistics reports permit only single-race identification. These changes in the reporting of race and education introduce a fundamental bias because of their different impacts on the numerator and denominator of a given subgroup's mortality rate. And because these biases change over time, they confound the estimation of trends in inequality.
Changes in the composition of the analyzed demographic subgroups present another serious source of bias (20-23). For example, Olshansky et al. (13) documented decreasing life expectancy among non-Hispanic white women without a high school degree between 1990 and 2008. But the share of the population of white nonHispanic females in this education category fell by about two-thirds between 1990 and 2010, which suggests that the average female high school dropout today is much more disadvantaged relative to her peers than the average female high school dropout in 1990. Bound et al. (20) argued that there is in fact no decrease in life expectancy




















| $\Delta 1990$ | $=2000$ | O 2010 |
| :--- | :--- | :--- |

Fig. 3. Female 3-year mortality rates by poverty percentile across age groups. Data are displayed as in Fig. 2. Table S4 reports key magnitudes and standard errors.
for the least educated once these compositional changes are accounted for. Similarly, single counties that experienced declining life expectancy $(11,12)$ tend to be poor places that have lost population over the past 20 years. If the healthiest people leave, then the ones who remain will be less healthy on average, biasing the estimated changes in mortality inequality. Our approach accounts for potential compositional changes by reordering county groups so that they represent constant shares of the population over time. In the context of county groups, however, such compositional changes do not seem to play a crucial role, as our results look very similar when we keep the county groups assigned in 1990 fixed and follow them up to 2010 (fig. S6).

One limitation of our approach is that it necessarily focuses on differences between groups of counties, whereas much of the increase in (for example) individual income inequality may be occurring within counties. However, by its nature, mortality must be calculated relative to some reference group. Using county groups as the reference allows one to cleanly answer questions about inequality between these groups in a way that may not be possible with respect to other reference groups, such as education.

What are potential causes for the different age-specific trends that we observe? Aizer and

Currie (24) highlighted many possible reasons for large reductions in infant mortality among the poor, which have reduced inequality in mortality among infants. We are not aware of any research that has looked at the causes of reductions in mortality inequality among older children and young adult males. Some possibilities include expansions of public health insurance (25-29), other social safety net programs such as Head Start $(30,31)$, and reductions in pollution, which tend to have disproportionate effects on the poor (32).

Among older adults, it is likely that at least some of the increasing disparities in mortality reflect differential patterns of both taking up and quitting smoking over their life cycles. For example, better-educated people stopped smoking much more quickly after the U.S. Surgeon General's 1964 report on the dangers of smoking (33, 34). Improvements in medical care for conditions such as heart disease also tend to benefit the rich before they reach the poor. The outbreak of the opioid epidemic is another factor that may be driving increased mortality inequality and actual increases in mortality rates in middle age (19). As Case and Deaton (19) showed, it may be possible to get some insight into these questions by studying the causes of death in the Vital Statistics mortality data, although changes in mea-
surement, measurement error, and missing data about causes mean that these data have to be interpreted cautiously.

Our results point to decreasing inequality in mortality, particularly among the younger cohorts who will form the future adult and elderly population of the United States. It is possible that survivors who would otherwise have died will be in poor health as they age and thus reduce the average level of health in the population. However, another possibility is that the declines in mortality at younger ages reflect improvements in the entire underlying distribution of health (35). In at least one important example-the case of expansions of public health insurance for poor infants and young children in the late 1980s and early 1990s-the reduced early death rates in these cohorts are associated with better health (27-29) and higher earnings (25) as these cohorts reach young adulthood. Thus, there appears good reason to hope that today's young will also be healthier when they reach old age, and that inequality in mortality will decrease among these elderly in the future.

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## SUPPLEMENTARY MATERIALS

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## NEURODEVELOPMENT

# Complement and microglia mediate early synapse loss in Alzheimer mouse models 


#### Abstract

Soyon Hong, ${ }^{1}$ Victoria F. Beja-Glasser, ${ }^{1 *}$ Bianca M. Nfonoyim, ${ }^{1 *}$ Arnaud Frouin, ${ }^{1}$ Shaomin Li, ${ }^{2}$ Saranya Ramakrishnan, ${ }^{1}$ Katherine M. Merry, ${ }^{1}$ Qiaoqiao Shi, ${ }^{2}$ Arnon Rosenthal, ${ }^{3,4,5}$ Ben A. Barres, ${ }^{6}$ Cynthia A. Lemere, ${ }^{2}$ Dennis J. Selkoe, ${ }^{2,7}$ Beth Stevens ${ }^{1,8}{ }^{\boldsymbol{8}}{ }^{\boldsymbol{6}}$ Synapse loss in Alzheimer's disease (AD) correlates with cognitive decline. Involvement of microglia and complement in AD has been attributed to neuroinflammation, prominent late in disease. Here we show in mouse models that complement and microglia mediate synaptic loss early in AD. C1q, the initiating protein of the classical complement cascade, is increased and associated with synapses before overt plaque deposition. Inhibition of C1q, C3, or the microglial complement receptor CR3 reduces the number of phagocytic microglia, as well as the extent of early synapse loss. C1q is necessary for the toxic effects of soluble $\beta$-amyloid (A $\beta$ ) oligomers on synapses and hippocampal long-term potentiation. Finally, microglia in adult brains engulf synaptic material in a CR3-dependent process when exposed to soluble A $\beta$ oligomers. Together, these findings suggest that the complement-dependent pathway and microglia that prune excess synapses in development are inappropriately activated and mediate synapse loss in AD.


enome-wide association studies implicate microglia and complement-related pathways in Alzheimer's disease (AD) (1). Previous research has demonstrated both beneficial and detrimental roles of complement and microglia in plaque-related neuropathology ( 2,3 ); however, their roles in synapse loss, a major pathological correlate of cognitive decline in AD (4), remain to be identified. Emerging research implicates microglia and immunerelated mechanisms in brain wiring in the healthy

[^1]brain (1). During development, Clq and C3 localize to synapses and mediate synapse elimination by phagocytic microglia (5-7). We hypothesized that this normal developmental synaptic pruning pathway is activated early in the AD brain and mediates synapse loss.
The degree of region-specific synapse loss is a stronger correlate of cognitive decline in AD than counts of plaques, tangles, and neuronal loss $(8,9)$. To determine how early synapse loss occurs, we used superresolution structured illumination microscopy (SIM) (10) to quantify synapse density in hippocampal CA1 stratum radiatum of familial AD-mutant human amyloid precursor protein (hAPP) ("J20") transgenic mice (11). Quantification of colocalized pre- and postsynaptic puncta [synaptophysin and postsynaptic density 95 (PSD95) (Fig. 1A); synaptotagmin and homer (fig. S1, A to D)] revealed a significant loss of synapses in J20 hippocampus at 3 to 4 months old (mo), an age that precedes plaque deposition (11, 12). Synapse loss in preplaque J20 CA1 was confirmed by electron microscopy (fig. S1G). Confocal imaging also showed synapse loss in CA1, CA3, and dentate gyrus of 3 mo J20 hippocampus but not in striatum (fig. S1E). Synapse

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J. Currie and H. Schwandt (April 21, 2016)

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Editor's Summary

## Narrowing of the life expectancy gap

In the United States, the rich can expect to enjoy better health and a longer life than the poor. Despite policies directed at improving the health of both the young and the poor, there is little evidence that this relationship has changed. Currie and Schwandt looked specifically at the life expectancy of present-day children and young adults, finding that mortality inequality has in fact declined over the past 25 years (see the Perspective by Bailey and Timpe).

Science, this issue p. 708; see also p. 661

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[^0]:    ${ }^{1}$ Center for Health and Wellbeing, Princeton University, Princeton, NJ 08542, USA. ${ }^{2}$ National Bureau of Economic Research, Cambridge, MA 02138, USA. ${ }^{3}$ Institute for the Study of Labor (IZA), 53072 Bonn, Germany. ${ }^{4}$ Department of Economics, University of Zurich, 8031 Zurich, Switzerland. ${ }^{5}$ Center for Economic Performance, London School of Economics, London WC2A 2AE, UK.
    *Corresponding author. Email: jcurrie@princeton.edu

[^1]:    ${ }^{1}$ F.M. Kirby Neurobiology Center, Boston Children's Hospital (BCH) and Harvard Medical School (HMS), Boston, MA 02115, USA. ${ }^{2}$ Ann Romney Center for Neurologic Diseases, Department of Neurology, Brigham and Women's Hospital (BWH) and HMS, Boston, MA 02115, USA. ${ }^{3}$ Alector Inc., 953 Indiana Street, San Francisco, CA 94107, USA. ${ }^{4}$ Annexon Biosciences, 280 Utah Avenue Suite 110, South San Francisco, CA 94080, USA. ${ }^{5}$ Department of Anatomy, University of California San Francisco (UCSF), San Francisco, CA 94143, USA. ${ }^{6}$ Department of Neurobiology, Stanford University School of Medicine, Palo Alto, CA 94305, USA. ${ }^{7}$ Prothena Biosciences, Dublin, Ireland. ${ }^{8}$ Stanley Center for Psychiatric Research, Broad Institute of MIT and Harvard, Cambridge, MA 02142, USA.
    *These authors contributed equally to this work. †Corresponding author. Email: beth.stevens@childrens.harvard.edu

