Pursuing the limits of child survival in the most and least developed countries

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Summary paragraph

The epidemiological transition from young to old deaths in high-income countries reduced
mortality at all ages, but a major role was played by a decline of infant and child mortality
from infectious diseases[1,2] that greatly increased life expectancy at birth[2,3]. Over time,
declines in infectious disease continue but chronic and degenerative causes persist[4,5], so we
might expect under-5 deaths to be concentrated in the first month of life. However, little is
known about the age-pattern of this transition in early mortality or its potential limits.
Here we first describe the limit using detailed data on Denmark, Japan, France, and the
USA— developed countries with low under-5 mortality. The limiting pattern of under-5
deaths concentrates in the first month, but is surprisingly dispersed over later ages: we call
this the early rectangularization of mortality. Then we examine the progress towards this
limit of 31 developing countries from sub-Saharan Africa (SSA)—the region with the
highest under-5 mortality[6]. In these countries, we find that early deaths have large age-
heterogeneities; and that the age patterns of death is an important marker of progress in
the mortality transition at early ages. But a negative association between national income
and under-5 mortality levels, confirmed here, does not help explain reductions in child
mortality during the transition.

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Main

The epidemiological transition from young to old deaths in high-income countries has

characterized mortality decline at all ages. But the decline of infant and child mortality from

exogenous causes (infections or parasites, accidents)[1,2] led to substantial increases in life

expectancy at birth[2,3], and an increasing modal age-at-death for adults[7]. Over time, as declines in

exogenous mortality continue, under-5 mortality (U5M) should shift towards ages close to birth,

but is likely to remain high during the first month of life due to the persistence of endogenous

causes of death (congenital malformations, injuries connected with birth)[8]. Here we identify a limit

reached by the rich countries with very low U5M, and call this limiting pattern of death the *Early*

Rectangularization of the Mortality Curve. This limiting age-distribution for U5M in rich countries

is not reflected in Sustainable Development Goal 3, which sets a target of neonatal (NMR) and

under-5 mortality rates of 12, respectively 25, deaths per 1,000 births by 2030 for countries with

currently high U5M[6]. Here, we investigate the extent of progress towards our U5M limit in 31 sub-

Saharan African (SSA) countries.

In any year, U5M levels vary dramatically with age, and change at different rates with respect to age and across countries. These factors are not accounted for by established techniques, e.g., Gini coefficient, standard deviation of age at death, and life disparity (these do work at old ages as measures of mortality compression or rectangularization[1,9,10], or as measures of lifespan inequality at adult ages[11–14]). To overcome these limitations for U5M, we use a generalized Gini ($G_{[0]}$) index that is proportional to the average mortality rates, and accounts for relative as well as absolute differences across ages (Methods) (1). We expect inequality as measured by $G_{[0]}$ to decline over time in response to declines in U5M at any age. In contrast, at adult ages inequality in lifespan can decline or increase depending on the ages at which mortality declines[15,16].

We rely on historical data from Denmark (1901-2016), France (1975-2016), Japan (1980-2016), and the USA (1968-2016), with high-quality and fine-grained U5M records measured in weeks/months (Supplementary Methods). These countries are not only highly developed economically, but also display the lowest U5M levels ever recorded after more than 100 years of continuing declines (Fig 1a). Long-run convergence is shown by U5M in Denmark during the first 80 years of the 20th Century, with $G_{[0]}$ declining from 1.6 to 0.1 x1000 (an average annual decline of 2.4%). More recent data reveal that Japan has reached the lowest concentration of under-5 mortality in 2016 ($G_{[0]} = 0.025$ x1000) and we argue that Japan is close to the actual limit of U5M at that time as it is converging more rapidly than other developed countries—$G_{[0]}$ in Japan declined by 74% between 1980 and 2016, followed by France (65%), Denmark (60%), and the U.S. (54%) during the same period. The lower level and more rapid decline of neonatal mortality in Japan during the period explains its leading role in the mortality convergence (Fig 1b). By 1980, Japan's neonatal and under-5 mortality were around one third of the levels set by the SDG-3 by 2030 (4.1

and 8.4 per 1,000 births, respectively), and continued declining to lower levels of 0.7 and 2.3

deaths per 1,000 births by 2016, respectively. Likewise, by 1975 France's NMR (7.6) and U5M

(13.9) were around two-thirds of the respective SDG-3 limits, Denmark reached those limits

between 1955-1965 (U5M was 24.7 x1000 in 1955 and NMR was 12.5 x1000 in 1965), and the

USA was about to reach them around 1968 (NMR=13.6, U5M=21.5) (Fig. 1b).

In Figure 1c we show that the limiting pattern of under-5 mortality can vary, but the rectangular

(or 'L') shape—characterized by an age pattern where mortality rates decline overall but remain

high during the first month of life— remains a distinctive characteristic even at very advance stages

of the mortality transition. Had all deaths occured within the first month of life (as illustrated by

the dashed diagonal line in Fig. 1b), the limiting pattern would be represented by a vertical line at

age 0 and flat for older ages. The recent age-pattern in Japan is much less concentrated, even

though U5M remains high in the first month after birth. The recent age-patterns of U5M from

Denmark and France are similar to that in Japan, but with higher NMRs; whereas the age pattern

from the USA also exhibits an 'L' shape, but neonatal and infant mortality (deaths within the first

year of life) rates were the highest among the 4 countries (Fig. 1c).

The speed of convergence to the limiting pattern is determined by the combined effects of changing

mortality levels and age structure. Thus, with some variations in neonatal mortality levels,

Denmark and France have experienced similar transitions in the shape and concentration of

mortality during the last 40 years (Fig. 2), but both are still behind Japan. The transition in the

USA has been slower as its age pattern in 2016 was still similar to that observed in Japan in 1980,

which explains the remaining differences in mortality concentration and age-inequality between

these countries during the last 40 years. The generalized Lorenz curve flattens rapidly in Japan as

its $G_{[0]}$ approaches the mean under-5 mortality rate ($G_{[0]}$=0.66·$\bar{q}$; i.e., closest to the theoretical limit

of G*=0.98·$\bar{q}$), with the lowest $\bar{q}$ observed among these 4 countries (at $\bar{q}$=0.038 x 1000) (Fig. 2b).

Heterogeneities in mortality shapes, levels, and rates of convergence are far more pronounced

across the least developed countries, as their epidemiologic transition has started more recently.

To appreciate this from a historical perspective, we note that Denmark in 1901 exhibited high

mortality for neonates and infants and then declining smoothly and rapidly with age. But by 2016,

U5M in Denmark exhibits a clear rectangular shape (Extended Fig. 1). We observe roughly similar

declines for Rwanda between 1994 and 2014, with a rapid transition and early rectangularization

— even so, although Rwanda is likely to reach the SDG-3 by 2030[17], the country is likely to be

well short of the U5M limit in developed countries even by 2050 (Extended Fig. 1). In general, the

SSA region is converging unevenly, as shown by values of the generalized Gini (Fig. 3a). As we

show there, Rwanda and Senegal have made outstanding progress in mortality decline and reduced

age-inequality during the last 20 years; meanwhile, countries like Nigeria have fallen behind and

still exhibit mortality levels and shapes only observed in the developed world a century ago.

Unlike Rwanda, other SSA countries will fall short in meeting the SDG-3 targets by 2030. For

instance, Kenya would make it by 2050, but Nigeria (and other countries) is expected to fall short

even by 2050 (Extended Data Fig. 1)[17]. A major cause of this delayed transition is the slow

convergence of neonatal deaths in most countries of the region (Fig. 3b).

At that current pace, what are the prospects of convergence towards Japan's current limit

distribution for countries in SSA? According to our predictions, the generalized Gini in Rwanda

100 and Kenya would still be 10 times larger in 2030 and 2050 than Japan in 2016, respectively; or 19

101 times larger in Nigeria by 2050 (Extended Data Fig. 1). Therefore, most SSA countries face

102 an important challenge just to meet the SDGs, but convergence to our early rectangularization

103 limit of U5M does not look realistic in the foreseeable future (Fig. 3b). In any case, we expect that

104 the latter convergence will become increasingly challenging in SSA countries, and will require

105 progress in both technology and socioeconomic conditions.

107 Our analysis relied on data from countries at opposite sides of the income distribution, and by

108 using 2017 mortality and per capita GDP data we found an expected inverse association[18]—a

109 concave-up shape with a declining monotonic hazard—as levels of U5M fall with economic

110 development (Fig. 4a and Extended Data Fig. 2). However, income does not explain the

111 remarkable differences in the rate of mortality decline within economic regions over time (between

112 1990-2017), particularly at the lower tail of the income distribution—between 1990 and 2017, the

113 association between income and the decline in the under-5 mortality rate changed between 0.20%

114 and 0.73% in low-income countries (LIC), while it was positive and increasing for upper middle-

115 income countries (UMC) or only varied from 0.13% to 0.42% in high-income countries (HIC)

116 (Fig. 4b). These results contribute to the mixed evidence on the relationship between income and

117 health/mortality, and the shape of that association[19,18,20–23].

118 Our findings also suggest that economic performance was not a precondition for reducing U5M in

119 SSA during its early transition, where policies and interventions have been effective in countries

120 that have met the Millennium Development Goals for child mortality reduction (e.g., Rwanda and

121 Tanzania)[24]. It is less clear, however, whether further progress would be feasible in the region

122 while maintaining low levels of economic development.

Our limit for child survival has scientifically important implications for the causes of early death.

It is important in terms of whether future medical/technological advances would allow further

reductions of neonatal deaths attributed to endogenous etiology (e.g., conditions originating in the

perinatal period or congenital causes)[25]. This limit also has important policy implications for

establishing realistic benchmarks in specific contexts; e.g., policies aiming to reach specific goals

in poor regions, or to reduce gaps among disadvantaged or minority groups in high-income areas[26].

Our limit also shows that the age distribution of mortality is important in analyzing mortality

convergence, and reveals sharp heterogeneities across and within regions. Age-specific progress

in SSA may require a combination of structural investments that were effective in high-income

countries independently of vaccination and medical technologies[27], and a significant scale-up of

programmes and interventions that target maternal and child care and services[28]. Finally, limited

resources in SSA suggest that cost-effective/efficient strategies, such as investments skewed more

to primary care than secondary/tertiary care, are likely be more beneficial at the current stage of

their mortality transition[29].

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Methods

Sources and retrospective mortality data

We collected age profiles of under-1 mortality from publicly available sources from Denmark

(1901-2016)[30], France (1975-2016)[31], Japan (1980-2016)[32], and the USA (1968-2016)[33]. Mortality

rates for ages 1, 2, 3, and 4 were not available for these 4 countries, but we obtained the

corresponding rates as reported in the Human Mortality Database (HMD) for each country[34]

(Supplementary Fig. S1). We then adjusted monthly-based under-5 mortality profiles using a two-

dimensional linear interpolation method[35].

Data from sub-Saharan Africa (SSA) were obtained from birth histories of 106 nationally

representative Demographic and Health Surveys (DHS) from 31 countries from the period between

1990 and 2017. The DHS Program collects health and demographic information for women in

reproductive age (15-49 years old) and their children, using a two-stage stratified cluster sampling

design that defines strata by region and by rural-urban within each region. A sampling frame

consisting of census enumeration units or tracks or Primary Sampling Units (PSUs) that cover the

entire country is used for the random selection of PSUs in the first stage, with probability

proportional to sampling size. Clusters correspond to selected or further split PSUs (or blocks). In

the second stage, households are selected systematically from a list of previously enumerated

households within each selected cluster or block.

Mortality profiles by age were assessed using full birth history (FBH) data available for individual

women aged 15 to 49 years in DHS surveys, where the respondent mother is asked about the date

of birth of each of her ever-born children, and the age at death if the child has already died[36]. The

complete list of countries included in this study are in the Extended Data Table 1.

'Conditional' life-table age distribution of under-5 deaths

We used life-table age distributions of death, constructed by Mejía-Guevara et al. (2019) based on

the following demographic analysis[17].

First, using death reports by households and retrospective information from full birth history

(FBH) data from DHS[36], estimates of age-specific death rates $m_{[x]}$ ([x] stands for age in months)

were obtained from survey estimates of the number of events and total time to event[36], and used

to compute period life-table probabilities of dying, $q_{[x]}$ (probability of dying between month x and

month $x+1$), assuming a uniform distribution of deaths across age.

Second, mortality rates by age were adjusted using the most recent estimates of neonatal, infant,

and under-5 mortality rates from the Inter-agency Group for Child Mortality Estimation

(IGME)[6,37] to minimize the risk of measurement errors in direct estimates of under-5 deaths based

on FBH [e.g., survivor and truncation bias][36,38]. After the previous adjustment, mortality profiles

were smoothed over ages and years by using a two-dimensional P-Spline smoothing and

generalized linear model, assuming that the number of deaths at a given rate are Poisson-

distributed[39].

Third, a variant of the Lee-Carter (LC)[40] model (LLT), suitable for mortality profiles using datasets

that contain multi-year gaps, was used to fit the age mortality profiles after smoothing for each

country. The resulting fits were first used to generate smoothed point estimates of age-specific

death rates within the 1990-2017 period.

More details about the construction of age profiles, adjustment, fit, and goodness of fit are available

elsewhere[17].

Early mortality compression, convergence in distribution, and age-inequality

1) Generalized concentration index and the potential limits of under-5 mortality decline

To measure the degree of mortality compression of U5M age profiles and age-inequality we use the Lorenz or concentration curve, denoted here as $L_{[0]}$, and the Gini coefficient. We build on Hanada (1983)[41] and Shkolnikov et al. (2003)[11] who introduced the Gini coefficient as a measure of inequality of life table data. In general, the concentration curve plots the cumulative proportion of one variable (e.g., health) against the cumulative proportion of the population ranked by another variable (e.g., income). If the ranking of units of analysis (by health) is the same as the ranking of the population, the concentration and the Lorenz curve are the same. The concentration (Lorenz) curve lies below the line of 45° when poor health is concentrated in the lowest ranked group[42]. The Concentration (Gini) index (C) is then calculated as twice the area between the concentration (Lorenz) curve and the diagonal, and we define it to measure concentration of under-5 mortality as[42]:

$$C(q|m) = \frac{2cov(q_{[x]}, R_{[x]})}{\bar{q}} = \frac{1}{n}\sum_{x=0}^{59}\frac{q_{[x]}}{\bar{q}}\left(2R_{[x]} - 1\right), \qquad (1)$$

where $q_{[a]}$ is defined as before as the probability of dying between month x and month $x+1$, $R_{[x]}$ is the relative rank of the probability of dying at month x, $\bar{q} = \sum_{a=0}^{59} q_{[a]}/n$, $(n = 60)$ is the mean level of mortality rate in the age range of 0 to 59 months.

The Gini index as defined in equation (1) is a measure of relative inequality, and it is invariant to proportionate changes in mortality levels. This represents a limitation in the context of child mortality, particularly at advance stages of the transition where levels or changes in the neonatal mortality occur disproportionately in relation to those in other age groups (see Extended Data Fig. 3). To address this limitation, we used a generalized version of the Gini index ($G_{[0]}$) that is obtained

317   by multiplying C(q | m) by the mean level of mortality rate ($\bar{q}$). With this adjustment, $G_{[0]}$ becomes

318   invariant to equal improvements in mortality across age groups. The generalized Gini can then be

319   expressed as:

$$G_{[0]}(q|m) = \frac{1}{n} \sum_{x=0}^{59} q_{[x]}\left(2R_{[x]} - 1\right). \tag{2}$$

In the context of U5M, we interpret declining values of $G_{[0]}$ as evidence of reduced inequality,

convergence or early compression of under-5 deaths, with the relative mortality concentrated at

ages close to 0, but declining with decreasing the mean probability of dying $\bar{q}$.

An important feature of this index is that it is upper bounded by $\bar{q}\left(\frac{n-1}{n}\right)$, with n=60, the number

of age groups. This upper limit is particularly relevant for this research because under-5 mortality

is more concentrated at early ages, the Lorenz curve is below the diagonal, and although the Gini

approaches 1 [more precisely $\left(\frac{1-n}{n}\right)$] as mortality concentrates around age 0, $G_{[0]}$ decreases with

the decline of the mean probability of dying. That is, the generalized Gini index approaches the

mean probability of death as the early rectangularization of mortality proceeds. In the hypothetical

event of perfect rectangularization, where all deaths are concentrated at age 0 ($q_{[x]} \to 0$, for x >

0), then the limit of $G_{[0]}$ is proportional to the neonatal mortality level [$G_{[0]} \to q_{[0]}\left(\frac{1-n}{n^2}\right)$], and

U5M approaches $q_{[0]}$ in that limit. In Figs. 1b and 3b we illustrate this hypothetical scenario—of

all under-5 deaths occurring at the neonatal level—with diagonal dashed lines, that illustrate the

extent of mortality convergence across countries as their neonatal and U5M rates decline at

different rates. Analysis of mortality concentration was conducted using the *Stata* command

'conindex'[43].

2) Other measures of lifespan inequality

To measure changes in age patterns of under-5 mortality and convergence in distribution towards

age 0, we considered other measures of inequality that have been used in previous studies to

analyze mortality of adult profiles. We first computed the minimum interval where a percentage p

of the total under-5 deaths takes place, that we denoted A_P. This indicator is equivalent to the C-

family of indicators proposed by Kannisto[10] for assessing compression of mortality, except that

the author used it for adult mortality and our indicator always includes the age of 0 (modal age for

under-5 deaths)—the age where the higher percentage of under-5 deaths occurs. For instance,

while A_{75} represents the interval where 75% of under-5 deaths take place, C_{75} is the age interval

where 75% of deaths in the whole population takes place (Extended Data Fig. 4).

Another indicator that we considered to account for differences across age groups and compression

of U5M is the 'conditional' mean-age-at-death based on survival to age 'x' months (that we denote

as [x] along this study) but death by age 5 years; we denote these by $E_{[x]}$. Thus, $E_{[0]}$, $E_{[1]}$, and $E_{[12]}$

respectively represent the 'conditional' mean age-at-death after surviving birth, survival to age 1

month, and survival to age 12 months. Finally, we assessed the between-individual inequality of

age-at-death using the 'conditional' standard deviation $S_{[x]}$ that assumes survival to age 'x' and

death by age 5 years. We selected this indicator on the basis of theoretical[11] and empirical[12]

findings on measures of lifespan inequality. Formally, we estimated $E_{[x]}$ and $S_{[x]}$ using the

following expressions:

$$E_{[x]} = \frac{\int_x^\omega a \cdot \phi(a)da}{(l_{[x]} - l_{[\omega]})}, \tag{2}$$

$$S_{[x]} = \sqrt{\frac{\int_x^\omega (a - E_{[x]})^2 \cdot \phi(a)da}{(l_{[x]} - l_{[\omega]})}}, \tag{3}$$

where ω is the oldest age in the conditional life table (59 months), $l_{[x]}$ is the survivorship or

probability of living to at least age x, and $\phi(a) = \mu(a)l(a)$ is a density representing the

probability that an individual dies at age a [$\mu(a)$ is the age-specific mortality rate].

As with the Gini coefficient, these three indicators (A_{75}, $E_{[x]}$, $S_{[x]}$) represent relative measures of

mean and variation and did not account for differences in the level of mortality across age

groups. This was particularly problematic for analyzing age profiles of developed countries,

where these indicators provided evidence of relative mortality differences that did not account

for changes in absolute mortality levels across these countries, particularly at the neonatal level.

Conditional marginal effects of income on under-5 mortality

Using information of GDP per capita—expressed in international Purchasing Power Parity (PPP)

2011 United States Dollars (USD)—and under-5 mortality for 217 countries obtained from the

World Development Indicators[44], we conducted an ecological association of mortality an national

income using a scatter plot association with values from 2017. We superimposed the association

from 1990 and 2017 using average values of U5M and income of countries stratified by 4 economic

regions: low-income (LIC), low-middle-income (LMC), upper-middle-income (UMC), and high-

income (HIC) (Fig. 3a). We further adjusted a curve that represents the prediction of under-5

mortality from an estimated fractional polynomial of income using the command 'twoway fpfit'

from Stata. We replicated the ecological analysis using values from 1990 and with neonatal

mortality rates, highlighting in color results from representative countries from each region (i.e;

Rwanda from region 1; Nigeria from region 2; Gabon from region 3; and Denmark, France, Japan,

and the U.S.A from region 4: Extended Data Fig. 2).

To estimate income elasticities of under-5 mortality, we fitted linear regression models with under-

5 mortality as dependent variable and real GPD per capita (gdp) as independent variable,

controlling for the year of observation, using the following linear specification:

$$E[u5mr|gdp] = \widehat{u5mr} = \widehat{\alpha} + \widehat{\beta} \cdot gdp + year. \tag{7}$$

379   With the predicted values of under-5 mortality $(\widehat{5mr})$ from model (7), the average elasticities are

380   estimated as follows:

$$\frac{\partial(\log u5mr)}{\partial(\log gdp)} = \frac{\partial(u5mr)}{\partial(gdp)} \cdot \frac{gdp}{u5mr}$$

$$\approx \hat{\beta} \cdot \frac{gdp}{\widehat{u5mr}} \qquad\qquad (8)$$

$$\approx \frac{\frac{\Delta \widehat{u5mr}}{\widehat{u5mr}}}{\frac{\Delta gdp}{gdp}}$$

The elasticity can then be interpreted as the percentage/proportionate change in the expected

mortality for a percentage/proportionate change in the GDP per capita. Elasticities were calculated

in a yearly-basis relative to the median GDP per capita value of the corresponding year and region

using a regression approach and the post-estimation command 'eyex' from *Stata*[45].

Strengths and limitations

Our study builds on previous research that analyzed inequality in the length of life and convergence

of adult mortality[11,12,46]. Using under-5 mortality profiles from narrow-age groups from Denmark,

France, Japan, and the USA from publicly available data, and from 31 SSA countries with data

constructed by Mejía-Guevara et al. (2019)[17] for the forecasting of mortality rates and the

assessment of the SDG-3 by 2030, this study is—to our knowledge—the first to use this kind of

mortality patterns to explore the potential limit of child survival, by assessing compression,

convergence and under-5 age-inequality. In particular, the age profiles of under-5 mortality for

high-income countries allowed us to investigate potential limits in child survival, as data for the 4

developed countries in our sample included the most recent trends of under-5 mortality patterns,

except for Denmark where additionally relied on annual information as far as 1901, allowing a

broader analysis of the mortality transition of under-5 deaths during the last 120 years. We

investigated the extent of early compression, age-inequality and variation in under-5 mortality,

and we found important differences in the patterns of change across countries over time, with a

positive association between the rates of mortality change and age-inequality. We identified Japan

as the leading country in the under-5 mortality convergence, not only due to its lowest neonatal

and under-5 mortality levels, but also because the early compression and the least inequality of

mortality patterns that it has achieved in the most recent years.

To highlight differences in the evolution of age-patterns, we examine the youngest ages at which

75% of under-5 mortality takes place in a country that we denote by A_{75} (Extended Data Fig. 4).

In 1980, 75% of deaths in Japan occurred within the first 12 months of life and that percentage

remained around that level because its sharp decline in neonatal mortality relative to other ages, in

contrast with that observed in the USA over the past 50 years, where they occurred within the first

5 months because the relative importance of neonatal mortality remained the same. With much

higher levels of mortality at all ages and despite being on track to meet the SDG-3 of child

mortality reduction(13), Rwanda and Senegal will expect to exhibit mortality profiles comparable

to those observed in Denmark and France more than 50 years ago (Fig. 2 and Extended Data Fig.

1), and with larger neonatal levels. The gap is even larger for several SSA countries, like Chad and

Nigeria (Extended Data Table 1).

We identified salient differences across SSA countries in terms of distribution of deaths by age

and time, convergence, and inequality in mortality reduction by age; particularly after stratifying

countries on the basis of their achieved ARR between 1990 and 2016. On the one hand, countries

with an ARR below 3.2% experienced a slower convergence process both in terms of mean and

distribution of age-at-death. Mortality reductions occurred at different rates across under-5 ages,

the mean and dispersion of age-at-death remained constant or have increased in most countries,

which is in line with results from previous studies[47], showing unusual high values of $_4q_1$ (the

probability of children dying between age 1y and 5y) relative to $_1q_0$ (the probability of dying

during the first year of life) in several countries from SSA, that may be attributable to true

epidemiological patterns (e.g., disease environments characterized by high prevalence of

infectious diseases—malaria, measles, and diarrhea) rather than to data quality issues as similar

patterns have been found in previous studies based on Demographic Surveillance Systems (DSS)

data[47,48]. On the other hand, our findings revealed substantive evidence of mortality convergence,

both in terms of mean and distribution, for countries that have reached higher ARR ($\geq 3.2\%$),

including countries that have met the MDG-4—e.g., Ethiopia, Malawi, Niger, Tanzania, and

Rwanda[49]. However, when compared with the age profiles of highly developed countries, even

those like Rwanda and Senegal which are on track to achieve the SDGs by 2030[17], face a

difficult path ahead to achieve further reductions of child mortality according to our estimates.

The relationship between national income and under-5 mortality is in line with previous evidence

highlighting income as a key predictor of mortality gaps across economic regions[18]. We included

that evidence as our analysis is based on mortality distributions from two groups of countries

situated at both extremes of the income distribution and stages of the mortality transition. The

analysis reveal that despite that association, it is unclear whether increases in income are enough

to overcome the mortality gaps between countries in advanced transitions, for example between

Japan and the USA, or between countries in the earliest transition, even though the important

reductions achieved in countries like Rwanda during the past 25 years may not be attributed to

income growth changes but to the implementation of key policy interventions[50]. It is also

unclear, and our data are not suitable to response whether further reductions in mortality in poor

settings may be achievable without substantial improvements in economic development, or the

extent to which the USA would be able to reach similar levels as Japan despite its more

advanced economic position.

The scope of this study is limited for several reasons. Although our data for Denmark, France,

Japan, and the USA were obtained from high quality vital registration records that represent

diverse world regions, we only had information from those 4 developed countries. For SSA, we

relied on full-life histories from survey data, which are subject to several sources of error; as vital

statistics for this group of countries are inexistent or deficient in the countries included in our

study[51]. Second, we used different measures of inequality to improve the robustness of our analysis

as they differ in their underlying properties and sensitivity to age-specific mortality change[14].

Third, because of the nature of the survey data, we cannot infer causality neither in the association

of the average decline of mortality and age-inequality nor between mortality and national income,

and further research should explore the underlying risk factors, social determinants, and true

epidemiological patterns behind under-5 mortality change by age, particularly in the SSA region.

Figures and Tables

a

b

Fig. 1 | Convergence of under-5 mortality and the limiting patterns of under-5 deaths. a,
Trends of under-5 mortality (U5M) compression by country over time. **b,** Convergence of
Neonatal and U5M in the last 65 years. **c),** Life-table distribution of under-5 deaths for Denmark
and Japan in 2016. Data are for Denmark, France, Japan, and the USA for the period between
1980 and 2016. In **a** and **b**, Mortality compression/convergence was assessed using the
generalized Gini index ($G_{[0]}$), which accounts for both changes in the age distribution and levels
of mortality across countries over time, and smaller values of $G_{[0]}$ indicate early compression
(less concentration of deaths), less inequality, or more rapid convergence of under-5 mortality.
The dashed diagonal line in **a** illustrates the hypothetical case of perfect rectangularization of the
U5M curve where all deaths are concentrated at age 0 (the mean probability of death $\bar{q} \rightarrow$
$q_{[0]}/60$, where $q_{[0]}$ represents the neonatal mortality rate). In **c**, we illustrate the early
rectangularization of the mortality distribution, represented the empirical limiting patterns of
under-5 deaths in these four countries in 2016 (probabilities of death are per 1,000 in a squared-
root scale).

Fig. 2 | Age patterns of under-5 deaths and inequality in the age at death. **a,** Detailed age-at-death distribution of under-5 deaths in four developed countries (Denmark, France, Japan, and the USA). Age is measured in months in the x-axis and life-table probabilities of dying are represented in the y-axis (x1000 in a square-root scale). **b,** Generalized Lorenz curves in the same selected countries. Generalized Lorenz curves are constructed as the product of the standard Lorenz curve (Extended Data Fig. 1) and the mean mortality rates across ages ($\bar{q}$, measured in months). It therefore illustrates the cumulative percentage of under-5 population (x-axis), ranked by the level of mortality at each age, and graphed against the cumulative mean under-5 deaths (y-axis). The Lorenz curve flattens down more rapidly in Japan as its $G_{[0]}$ approaches the mean under-5 mortality rate ($G_{[0]}=0.66 \cdot \bar{q}$; i.e., closer to the theoretical limit of $G^*=0.98 \cdot \bar{q}$), with the lowest $\bar{q}$ observed among these 4 countries (at $\bar{q}=0.038$ x 1000) (Fig. 2b) (Supplementary Methods and Extended Data Table1).

Fig. 3 | Compression and convergence of under-5 mortality. a, Trends of under-5 mortality compression by country over time. **b,** Convergence of neonatal (hollow markers) and under-5 mortality (solid markers)by country over time. Data are for Denmark (1901-016), France (1975-2016), Japan (1980-2016), the USA (1968-2016), and from 31 SSA countries (1990-2017). Compression and convergence of mortality were assessed using the generalized Gini index

($G_{[0]}$), where smaller values of $G_{[0]}$ indicate early compression (concentration of deaths towards age 0), convergence to lower mortality levels, or less inequality. Data are from publicly available mortality records for high-developed countries and from the Demographic and Health Surveys for 31 SSA countries (Supplementary Methods).

a

b

Fig. 4 | Worldwide association of under-5 mortality and economic development. a, Country-wise association of GDP per capita income and under-5 mortality rates in 2017. GDP per capita is expressed in international 2001 Purchasing Power Parity (PPP) United States Dollars (USD), under-5 mortality rates as deaths per 1,000 live births, and yearly average values are shown for the period between 1990 and 2017 by economic region (in color). The adjusted curve represents the prediction of under-5 mortality from an estimated fractional polynomial of income. **b,** Conditional marginal effects of income between 1990 and 2017 by economic region. Income elasticity of mortality represents a % change in under-5 mortality for a % change in GDP per capita. Vertical dotted lines indicate the median GDP at the initial and end of the period in each economic region. Data are from World Development Indicators, retrieved online from The World Bank (2019), which included 217 countries worldwide (8 countries with GDP above 70,000 are not displayed in panel **b**, classified in 4 economic regions: low-income (LIC—pink), low-middle-income (LMC—maroon), upper-middle-income (UMC—blue), and high-income (HIC—orange) (Supplementary Methods).

Extended Data

Extended Data Fig. 1 | Age patterns of under-5 deaths and inequality in the age at death. **a,** Detailed age-at-death distribution of under-5 deaths in Denmark (1901-2016) and four selected countries from SSA (Gabon, Kenya, Nigeria, and Senegal). Age is measured in months in the x-axis and life-table probabilities of dying are represented in the y-axis (x1000 in square-root scale). Forecasts of mortality age profiles were conducted for countries in SSA by 2030 and 2050. **b,** Generalized Lorenz curves in the same selected countries showed in **a)**. Generalized Lorenz curves are constructed as the product of the standard Lorenz curve (Extended Data Fig. 1) and the mean mortality rates across ages (measured in months). It therefore illustrates the cumulative percentage of under-5 population (x-axis), ranked by the level of mortality at each age, and graphed against the cumulative mean under-5 deaths (y-axis) (Supplementary Methods and Extended Data Table1).

a

b

Extended Data Fig. 2 | Worldwide association of neonatal and under-5 mortality and economic development. a, Country-wise association of GDP per capita income and neonatal mortality rates in 1990 and 2017. **b,** Country-wise association of GDP per capita income and under-5 mortality rates in 1990 and 2017. GDP per capita is expressed in international 2001 PPP USD, mortality rates as deaths per 1,000 live births, and yearly average values are shown for the period between 1990 and 2017 by economic region (in color). The adjusted curves represent the prediction of neonatal or under-5 mortality from an estimated fractional polynomial of income. Data are from World Development Indicators, retrieved online from The World Bank (2019), which included 217 countries worldwide (8 countries with GDP above 70,000 are not displayed in panel **b**), classified in 4 economic regions: low-income (LIC—pink), low-middle-income (LMC—maroon), upper-middle-income (UMC—blue), and high-income (HIC—orange) (Supplementary Methods).

Extended Data Fig. 3 | Under-5 inequality in the age at death. Standard Lorenz curves for the first and the latest observed years in Denmark (1901, 2016), Japan (1980, 2016), the U.S.A. (1968, 2016), and Rwanda (1994, 2014, 2030). The Lorenz curve measures age inequality in the age at death distribution of under-5 deaths, and indicate less inequality if they move away (to the right) from the 45°

line that in the context of child mortality represents mortality evenly distributed across ages. Two times the area gap between the 45° line and the Lorenz curve is known as the Gini coefficient ($G_{[0]}$), which increases as under-5 inequality declines (until certain level where the relative importance of neonatal mortality declines relative to that at other ages, as occurring in the case of Japan or the USA) (Supplementary Methods and Extended Data Table1).

Extended Data Fig. 4 | Age distribution and compression of under-5 mortality. Change in mortality rates by age and the upper quartile of the conditional distribution of age at death (A_{75}) between the first and latest observed year, and projected by 2030 (the target year for the SDG-3 of child mortality reduction) in selected countries: Denmark, France, Japan, USA, Rwanda, and Senegal. These countries were selected to represent different trajectories in developed countries and to illustrate trajectories of the most successful countries in SSA in terms of under-5 mortality reduction within the past 25 years, like Rwanda and Senegal advancing rapidly and likely to reach the SDD-3 by 2030, but with relatively high neonatal mortality as compared to Denmark and France, countries in a very advanced mortality transition and low child mortality. Vertical lines represent the age where 75% of under-5 deaths took place in the respective year. For instance, in Rwanda, 75% of under-5 deaths occurred during the first 2 years of life in 1994, before 18 months in 2014, and they would occur before 12 months in 2030; whereas in Denmark, the 12 months of life concentrated the same 75% of under-5 deaths around 1901 (130 years ago). Data are from 31 DHS surveys and publicly available mortality records from Denmark, France, Japan, and the USA (Supplementary Methods).

Extended Data Table 1 | Income, mortality, and age-at-death indicators of compression and inequality.

Country	Year (1)	Year (2)	T	GDP (2011 PPP$) (1)	(2)	AGR	Neonatal Mortality (1)	(2)	ARR	Under-5 Mortality (1)	(2)	ARR	A_{75} (1)	(2)	$E_{[0]}$ (1)	(2)	$S_{[0]}$ (1)	(2)	$G_{[0]}$ (1)	(2)
Low-income	1990	2017																		
Benin	1990	2012	24	1463	1856	1.1	45	35	1.1	177	108	2.2	24	21	14.3	13.2	16.4	16.3	1.5	1.0
Burkina Faso	1990	2010	22	844	1423	2.6	44	31	1.8	199	116	2.7	27	24	15.9	14.6	16.3	16.0	1.5	1.0
Burundi	1990	2017	11	1032	668	-1.6	40	22	2.2	179	62	3.9	23	16	14.7	10.6	16.7	14.6	1.5	0.7
Chad	1990	2015	22	1110	2067	2.5	50	36	1.3	215	131	2.0	26	25	15.7	14.5	16.5	16.7	1.7	1.1
DR of Congo	1991	2014	22	1224	785	-1.9	42	30	1.4	193	102	2.8	20	13	13.1	10.1	15.1	13.9	1.8	1.1
Ethiopia	1990	2016	28	652	1608	3.5	60	30	2.7	204	64	4.4	26	16	14.4	10.7	17.4	15.8	1.8	0.7
Gambia	1993	2013	6	1504	1585	0.3	48	30	2.3	154	73	3.7	33	22	17.8	12.7	18.8	16.6	1.0	0.7
Guinea	1990	2012	24	1412	1684	0.8	67	28	3.9	240	103	3.8	22	20	13.4	12.9	15.8	15.0	2.2	1.0
Liberia*	1991	2011	12	1292	1139	-1.2	56	29	3.2	257	96	4.8	16	15	11.6	10.4	14.2	13.5	2.7	1.0
Malawi	1990	2016	23	744	1084	1.5	52	23	3.1	256	71	4.8	21	21	13.8	12.7	15.1	15.7	2.4	0.7
Mali	1990	2012	24	1272	1805	1.6	67	39	2.4	252	126	3.1	27	27	15.6	15.1	16.4	17.3	2.0	1.0
Mozambique	1990	2011	23	379	955	4.5	60	31	3.1	261	100	4.5	18	16	12.2	11.4	15.1	15.1	2.7	1.0
Niger	1990	2012	24	895	863	-0.2	55	30	2.8	345	114	4.9	30	27	18.0	15.5	15.7	16.3	2.4	0.9
Rwanda	1994	2014	22	497	1633	6.1	43	18	4.2	208	54	6.5	25	18	15.3	12.0	16.6	15.6	1.7	0.5
Senegal	1990	2017	28	2314	3143	1.1	38	21	2.2	126	45	3.7	28	13	15.9	9.4	17.1	14.5	1.0	0.5
Sierra Leone	1993	2013	6	1052	1655	2.3	52	38	1.6	252	138	3.0	21	14	13.5	9.7	15.2	11.7	2.4	1.6
Togo	1990	2014	13	1294	1406	0.3	43	27	1.9	150	80	2.6	22	20	13.6	12.6	16.1	16.1	1.4	0.8
Uganda	1990	2016	24	769	1689	3.1	35	21	1.9	186	53	4.7	21	14	14.1	9.7	15.4	13.7	1.6	0.6
Tanzania	1990	2016	19	1473	2584	2.2	37	22	2.1	170	57	4.2	21	16	13.9	10.5	15.7	14.5	1.5	0.6
Zimbabwe	1990	2015	16	2889	2197	-1.1	29	24	0.8	103	75	1.2	24	21	14.1	13.0	16.5	16.2	0.9	0.7
Lower-middle-income																				
Angola	1991	2016	7	4538	6265	1.3	54	30	2.4	232	85	4.0	25	20	14.7	12.4	16.7	16.3	2.0	0.8
Cameroon	1990	2011	23	3020	2969	-0.1	34	29	0.9	143	106	1.4	21	20	13.5	12.9	15.4	15.1	1.3	1.0
Congo	1993	2012	13	4928	5271	0.4	34	22	2.4	126	71	2.9	20	19	13.1	12.0	15.8	15.0	1.2	0.7
Cote d'Ivoire	1990	2012	15	3194	2712	-0.7	52	37	1.6	164	112	1.7	16	16	11.2	11.2	14.8	14.9	1.7	1.2
Ghana	1991	2014	13	1966	3828	2.9	40	27	1.7	117	59	2.9	21	15	12.9	9.7	16.1	14.2	1.1	0.7
Kenya	1990	2014	15	2380	2753	0.6	34	22	1.7	136	57	3.5	21	17	13.4	11.6	15.9	16.1	1.2	0.6
Lesotho	1995	2010	11	1628	2366	2.5	42	42	0.1	111	116	-0.3	15	15	10.4	10.6	15.0	15.0	1.2	1.2
Nigeria	1990	2013	25	3359	5492	2.2	54	36	1.8	229	118	2.8	21	19	13.1	11.8	14.3	14.1	2.2	1.2
Upper-middle-income																				
Zambia	1990	2014	23	2342	3633	1.8	37	24	1.9	180	68	4.0	22	17	13.8	11.2	14.9	14.9	1.6	0.7
Gabon	1992	2012	10	19006	16159	-0.8	32	24	1.4	95	63	2.0	18	20	11.9	12.5	15.7	16.5	0.9	0.6
Namibia	1992	2007	7	6270	8098	1.7	27	20	2.0	76	69	0.6	17	18	11.2	12.2	15.0	15.3	0.8	0.7
High-income*																				
Denmark	1901	2016	116	33786	45991	1.2	34	2	2.4	143	3	3.3	11	1	8.8	4.7	12.3	11.2	1.7	0.0
France	1975	2016	42	29515	38063	1.0	8	2	3.2	14	3	3.3	5	4	6.3	5.9	12.4	12.3	0.2	0.0
Japan	1980	2016	37	30582	38283	0.9	4	1	4.6	8	2	3.6	12	13	9.0	9.5	14.7	13.8	0.1	0.0
USA	1968	2016	49	37062	53399	1.4	14	3	3.0	21	6	2.7	3	4	5.2	5.6	11.7	11.9	0.3	0.1

* GDP for Liberia corresponds to 2001 and to 1990 for Denmark, France, Japan, and the U.S.A.

Complete list of countries classified by income group, and indicators of income and mortality for the initial and end year of analysis: GDP per capita levels (in international PPP 2011$) (GDP for Liberia corresponds to 2001 and to 1990 in Denmark, France, Japan, and the USA), and average growth rate (AGR); neonatal and under-5 mortality rates, and average rates of reduction (ARR); upper quartile of the conditional distribution of age at death (A_{75}); mean ($E_{[0]}$) and standard deviation ($S_{[0]}$) of age at death; and Generalized Gini Index ($G_{[0]}$).

Supplementary Information

Fig. S1 | Assessment of infant mortality fit from monthly-based age profiles and infant mortality from the Human Mortality Database (HMD) from Denmark, France, Japan, and the USA (Supplementary Methods)