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## Inequality in current and lifetime income

#### Abstract:

Standard theory of intertemporal choice predicts that people smooth out life-cycle changes in income by borrowing and saving, such that their standard of living in any given year depends more on lifetime income than on that year's income. Yet, contemporary empirical studies of income inequality are typically based on observations of income for one or a few years. This empirical simplification is due to the simple fact that researchers seldom have access to data on long-run or lifetime income. In this paper, we exploit a unique data set with nearly career-long income histories to provide novel evidence on inequality in current and lifetime income. We apply these results to assess the role of so-called life-cycle bias in empirical analysis of income inequality that uses current income variables as proxies for lifetime income. Our results suggest that income inequality increases over the working lifespan, and inequality in current income at age 60 is more than twice the inequality in lifetime income. Another key finding is that inequality in lifetime income is much lower than what cross-sectional estimates of inequality suggest. This means that we may need to reconsider how unequal individuals' living standard actually is.

Keywords: Income inequality, lifetime income, current income, life-cycle

JEL classification: D31; I31

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#### Sammendrag

Fordelingen av inntekt opptjent over en kort periode, typisk ett år, danner ofte utgangspunktet for å tallfeste inntektsulikheten i samfunnet. Empiriske analyser av inntektsulikhet basert på årlige tverrsnittsdata hviler imidlertid på en underliggende forutsetning om at årsinntekt er en god indikator for individenes økonomiske velferd. Både økonomisk teori og empiriske analyser tyder imidlertid på at en rekke personer har lave inntekter i en kort periode uten at det nødvendigvis er et uttrykk for at de har økonomiske problemer. En alternativ og mer tilfredsstillende metode for måling av inntektsulikhet er å benytte summen av inntekter opptjent over hele den yrkesaktive aldersperioden.

I denne artikkelen utnytter vi registerdata med informasjon om inntekt over livsløpet til å studere inntektsulikhet. Vi finner at ulikheten i årlig inntekt øker betydelig over livsløpet. Når individer er seksti år er ulikheten i årlig inntekt mer enn dobbelt så høy som ulikheten i livstidsinntekt. Våre funn tyder på at tradisjonelle empiriske studier basert på årlig inntekt for et tverrsnitt av befolkningen overvurderer inntektsulikheten i samfunnet.

#### 1. Introduction

More than half a century ago, Friedman (1957, p38) stated that "the identification of low measured income with 'poor' and high measured income with 'rich' is justified only if measured income can be regarded as an estimate of expected income over a lifetime or a large fraction thereof". Because individuals can borrow and smooth out life-cycle changes in income, he argued that the standard of living in any year depends more on lifetime income than on that year's income. To gauge inequality in living standard, the distribution of lifetime income is therefore likely to be more relevant than the distribution of current income.

Yet, contemporary empirical studies of income (or earnings) inequality are typically based on observations of income (or earnings) for one or a few years. This empirical simplification is due to the simple fact that researchers seldom have access to data on long-run or lifetime income. To provide evidence on inequality in current and lifetime income, we use a unique data set covering the entire Norwegian population with nearly career-long income histories for certain cohorts. We apply these results to assess the role of so-called life-cycle bias in empirical analysis of income inequality that uses current income variables as proxies for lifetime income.

The insights from our analysis can be summarized by two broad conclusions. First, we find evidence of substantial life-cycle bias in estimates of inequality based on current income. The life-cycle bias is minimized when individuals' incomes are measured at age 34-35. Later in the life, the bias increases steadily and inequality in current income at age 60 is more than twice the inequality in lifetime income. One implication is that cross-sectional estimates of income inequality are likely to be sensitive to the age composition of the sample. As a consequence, it is necessary to pay close attention to differences in age composition when comparing cross-sectional estimates of income inequality across countries, subgroups, or time.

Second, inequality in lifetime income is much lower that what cross-sectional estimates of inequality suggest. This means that we may need to reconsider how unequal individuals' living standard actually

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<sup>&</sup>lt;sup>1</sup> There are, however, a few studies that use income data over a longer period. A notable example is Björklund (1993), who uses a long, panel data on income from Sweden to compare the distributions of annual and long-term income. He finds that the inequality in long-run income is around 35 to 40 percent lower than in the cross-sections of annual income. Burkhauser and Poupore (1997) report reductions in income inequality indices of about 15 to 20 percent when the accounting period was extended from 1 to 6 years. Using income data for a whole decade, both Gittleman and Joyce (1999) and Aaberge et al. (2002) find that cross-sectional inequality indices overstate inequality in long-run income by 25 to 30 percent. The focus of these studies is how income mobility reduces inequality in long-run income, and they do not study the issue of life-cycle bias in studies of inequality that use current income as a proxy for lifetime income.

is. To understand better why inequality in lifetime income exceeds inequality in current income, we decompose the life-cycle bias into two components: income mobility (i.e. changes over time in individuals' positions in the current income distributions) and heterogeneous (but non-intersecting) age—income profiles. We find that income mobility reduces inequality in lifetime income by about 25 percent, while heterogeneous age—income profiles contributes to upward (downward) life-cycle bias in estimates of income inequality when current incomes are measured late (early) in the working lifespan. Overall, the equalizing effect of income mobility tends to dominate, implying that inequality in current income exceeds inequality in lifetime income even when current incomes are measured early in individuals' careers.

Our paper complements a growing literature in economics aimed at providing better-informed analyses of estimation biases in a wide range of research that uses current income variables as proxies for long-run income. Closely related to our paper is Haider and Solon (2006), who demonstrate that the association between current and lifetime income varies systematically over the life-cycle. As a consequence, regression models of intergenerational mobility that use current income as a proxy for lifetime income will produce inconsistent estimates of the intergenerational elasticity. The inconsistency varies as a function of the age at which current income is observed, and is therefore called life-cycle bias.<sup>2</sup>

Our point of departure is that life-cycle variation in the association between current and lifetime income, might not be representative for life-cycle variation in the association between *inequality* in current and lifetime income: The former depends on the covariance between individuals' current and lifetime income; the latter depends on how the covariance between individuals' incomes and their position in the distribution of income varies with the age at which current income is measured. Our empirical analysis demonstrates that this difference is it not merely a theoretical peculiarity, but empirically important for research on income inequality that uses current income variables as proxies for lifetime income.

This paper unfolds as follows. Section 2 presents a conceptual framework that relates inequality in current and lifetime income and illustrates the possible role of life-cycle bias. Section 3 describes our data and reports summary statistics. Section 4 provides evidence on inequality in lifetime and current income, before assessing the life-cycle bias. Section 5 concludes.

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<sup>&</sup>lt;sup>2</sup> The empirical analysis of Haider and Solon has been replicated and extended for Sweden (Böhlmark and Lindquist, 2006), Germany (Brenner, 2010), and Norway (Nilsen et al., 2012).

## 2. Conceptual framework

This section uses a framework of compensating differences, originally proposed by Mincer (1958), to relate inequality in distribution of current and lifetime income and illustrate the possible role of lifecycle bias.

Following Willis and Rosen (1979), suppose that individuals choose between two levels of schooling, labeled college (A) and high school (B), to maximize the present value of lifetime income. Assume that credit markets are perfect and the environment is perfectly certain, but occupations differ in the amount of schooling required. If an individual chooses college, his or her current income stream is

(2.1) 
$$y_{A,t} = \begin{cases} 0, & t \le s \\ \overline{y}_A e^{\gamma_A(t-s)}, & t > s \end{cases},$$

where s is the number of years it takes to get a college degree, t represents age (measured as years since high-school graduation),  $\overline{y}_A$  is initial income, and  $\gamma_A$  is the growth rate in income. If the individual chooses high school, his or her current income stream is

$$(2.2) y_{B,t} = \overline{y}_B e^{\gamma_B t}, t \ge 0,$$

where  $\bar{y}_{\scriptscriptstyle A}$  is the initial income and  $\gamma_{\scriptscriptstyle A}$  is the growth rate in income.

Suppose that additional schooling entails opportunity costs in the form of foregone income (but no direct cost such as tuition). Assume an infinite horizon, and an exogenously determined interest rate r, with  $r > \gamma_A > \gamma_B > 0$ . Then, the present value of income is

$$(2.3) Y_A = \int_0^\infty y_{A,t} e^{-rt} dt = \frac{\overline{y}_A e^{-rs}}{r - \gamma_A}$$

if college is chosen, and

$$(2.4) Y_B = \int_0^\infty y_{B,t} e^{-rt} dt = \frac{\overline{y}_B}{r - \gamma_B}$$

if high school is chosen. To induce a worker to choose college, foregone income while in school must be compensated by higher future income, such that  $Y_A > Y_B$ . In the long-run competitive equilibrium, the relationship between lifetime income and schooling is such that: (i) the supply and demand for workers of each schooling level are equated, and (ii) no worker wishes to alter his or her schooling level.

In the basic framework of compensating differences, individuals are ex ante identical. Suppose that inequality is measured by the Gini coefficient (G), and let  $p \in (0,1)$  denote the equilibrium proportion of workers with high school. Then, income inequality in equilibrium is given by

(2.5) 
$$G_{t} = \frac{p(1-p)|y_{A,t} - y_{B,t}|}{py_{B,t} + (1-p)y_{B,t}}$$

if current income is measured at age t, and by

(2.6) 
$$G = \frac{p(1-p)|Y_A - Y_B|}{pY_B + (1-p)Y_A}$$

if lifetime income forms the basis for the analysis.

In the spirit of Haider and Solon (2006), we define  $G_t - G$  as the *life-cycle bias* in income inequality from using current income at age t as a proxy for lifetime income. Equilibrium equates lifetime incomes such that individuals are indifferent between schooling levels, implying that G is equal to zero. Since foregone income while in school must be compensated, the equilibrium  $G_t$  will generally be non-zero and vary as a function of the age at which  $y_t$  is observed. In our case, the age at which  $G_t$  is zero and life-cycle bias is eliminated is uniquely given by

(2.7) 
$$t^* = \frac{\log \overline{y}_B - \log \overline{y}_A + s\gamma_A}{\gamma_A - \gamma_B},$$

which is obtained by inserting (2.1) and (2.2) in  $y_{A,t} = y_{B,t}$ .

More realistic models of income allow for ex ante heterogeneous individuals, such as in initial wages, growth rates, and the interest rate (see e.g. Willis and Rosen, 1979). Even so, the crucial insight of the basic framework of compensating differences still applies: to induce a worker to undertake additional schooling, foregone income while in school must be compensated by higher future income. This generates changes in the population variance of income around the central tendency of income growth, causing life-cycle bias in analysis of inequality using cross-section data.

To circumvent the issue of life-cycle bias, the data used in analysis of income inequality would ideally consist of complete longitudinal life histories of income. Unfortunately, such ideal data are seldom available. In the remainder of the paper, we provide novel evidence on inequality in current and lifetime income from nearly career-long income histories, and assess the role of life-cycle bias in empirical analysis of income inequality. Our analysis also provides an estimate of  $t^*$ . Restricting the sample to individuals around this age may be used as an imperfect, short-cut method for approximating inequality in lifetime from cross-section data.

#### 3. Data and definitions

#### **3.1 Data**

Our empirical analysis uses a longitudinal dataset containing records for every Norwegian from 1967 to 2006. The variables captured in this dataset include demographic information (sex, year of birth) and income. We focus on the 1942-1944 cohorts in order to ensure availability of data on income for more or less the entire working lifespan. In particular, these cohorts are between 23 and 25 years old in 1967 and between 62 and 64 in 2006.<sup>3</sup> Our analytical sample is restricted to males, given their role of breadwinner and primary wage-earner for these cohorts. Also, we exclude individuals whose information on annual income is missing. The final sample used in the analysis consists of 51 552 individuals.

Our measure of income includes all taxable income after deductions. Taxable income is the sum of pretax market income (from wages and self-employment), cash benefits (such as unemployment benefits, disability benefits, and family benefits), and capital income. We use the consumer price index to make incomes from different years comparable. We define *current income* as the average real income over a two-year period. The use of two years of income observations is a common approach in

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<sup>&</sup>lt;sup>3</sup> Although the formal retirement age is 67 years, many individuals retire age around age 65.

the empirical literature on income inequality to reduce the problem of measurement error (Jenkins and Van Kerm, 2006). Following Haider and Solon (2006), our measure of *lifetime income* is the annuity value of the discounted sum of real income

$$Y = \frac{y_T + \sum_{t=1}^{T-1} y_t \prod_{j=1+t}^{T} (1 + r_j)}{1 + \sum_{t=1}^{T-1} \prod_{j=1+t}^{T} (1 + r_j)},$$
(3.1)

where  $r_t$  denotes the real interest rates on income-transfers from period t-l to t.

It should be noted that the Norwegian income data have several advantages over those available in many other countries. First, there is no attrition from the original sample because it is not necessary to ask permission from individuals to access their tax records. In Norway, these records are in the public domain. Second, our income data pertain to all individuals, and not only to jobs covered by social security. Third, we have nearly career-long income histories for certain cohorts, and do not need to extrapolate the income profiles to ages not observed in the data. And fourth, our income measure is neither top-coded nor bottom-coded.

Table 1. Descriptive statistics for current and lifetime income (NOK)

	Full sample		Subsample by lifetime income 4 <sup>th</sup>		
	Mean	St. Dev.	quartile	quartile	quartile
Current income					
Age 26-27	177 883	64 753	138 304	177 659	217 912
Age 38-39	245 814	101 431	167 555	236 789	342 124
Age 50-51	272 246	423 089	145 380	233 691	476 223
Age 62-63	330 922	689 784	167 061	261 474	633 681
Lifetime income	249 523	64 753	160 577	232 852	371 811
Observations	51 :	552	12 888	25 776	12 888

*Notes*: The sample consists of males born 1948-1950. Current income is defined as the average income over the two-year period. Lifetime income is defined as the annuitized value of real income from age 24 to 63. Age refers to the 1949 cohort.

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<sup>&</sup>lt;sup>4</sup> The annual real interest rates on borrowing and savings are computed from Norwegian official statistics on interest rates on loans and deposits in commercial banks over the period 1967–2006.

#### 3.2 Descriptive statistics

Table 1 reports descriptive statistics. The first two columns display the mean and standard deviation in current and lifetime income for the whole sample. We can see that average income increases over the life cycle. Average current income is most similar to average lifetime earnings when individuals are in their 40s. The increase in average current income over the life cycle is accompanied by an increase in the variance of current income. This is an important observation, because life-cycle variation in inequality is due to changes in income variation around the central tendency of income growth.

In the last three columns of Table 1, we partition the sample into three groups according to the distribution of lifetime income. We can see that individuals with high lifetime income (cf. Column 5) experience, on average, rapid income growth over the life-cycle as compared to individuals with medium (cf. Column 4) or low (cf. Column 3) lifetime income. The life-cycle bias in inequality based on current income at a particular age depends on how well the differences in current income approximate the differences in lifetime income. The table suggests that the income gap late in individuals' careers tends to overstate the lifetime income gap. Taken at face value and assuming no income mobility, this would indicate that there is an upward (downward) life-cycle bias in estimates of ncome inequality when incomes are measured late (early) in the working lifespan.

#### 3.3 Inequality measures

To achieve rankings of intersecting Lorenz curves, many empirical studies employ the Gini coefficient. To examine the extent to which the empirical results depend on the choice of inequality measure, the conventional approach is to complement the Gini coefficient with measures from the Atkinson or the Theil family. However, the Gini coefficient and measures from the Atkinson or the Theil family have distinct theoretical foundations which make it inherently difficult to evaluate their capacities as complimentary measures of inequality.

As demonstrated by Aaberge (2000, 2007), an alternative approach for examining inequality in the distribution of income is to rely on the moments of the scaled conditional mean curve M, defined by

$$M(u) = \frac{E[Y | Y \le F^{-1}(u)]}{\mu} = \begin{cases} \frac{1}{u\mu} \int_{0}^{u} F^{-1}(t)dt, & 0 < u \le 1\\ 0, & u = 0 \end{cases}$$

where F is the observed cumulative distribution function with mean  $\mu$ , and  $F^{-1}$  its left inverse. Specifically, the first, second, and third order moments of the scaled conditional mean curve prove to make up a fairly good summarization of the conditional mean curve as well as the Lorenz curve. The  $k^{th}$  order moment of the scaled conditional mean curve for F, denoted  $C_k(F)$ , is defined by

(3.2) 
$$C_k(F) = k \int_0^1 u^{k-1} (1 - M(u)) du, \quad k = 1, 2, \dots$$

Aaberge (2007) characterizes the complementary nature of the first three moments of the conditional mean curve; they all satisfy the Pigou-Dalton transfer principle but differ in their sensitivity to inequality in the lower, central, and upper part of the distribution. The first moment ( $C_1$ ) is equal to the Bonferroni coefficient, which is particularly sensitive to inequality in the lower part of the distribution. The second moment is equal to the Gini coefficient ( $C_2 = G$ ), which is known to assign a lot of weights to inequality in the central part of the distribution. The third moment is an inequality measure ( $C_3$ ) that is particularly sensitive to inequality in the upper part of the income distribution. We will examine the sensitivity of the empirical results to the choice of inequality measure by complementing the information provided by G with its two close relatives  $C_1$  and  $C_3$ . Hence, we meet the most common criticism of the Gini-coefficient, namely that is insensitive to inequality in the tails of the distribution (see e.g. Wiles, 1974).

#### 3.4 Mobility measures

To understand better why inequality in current incomes differs from inequality in lifetime incomes, we need an appropriate measure of income mobility. Let  $F_Y$  and  $F_Z$  denote the distribution of the observed lifetime income Y and the distribution of the reference lifetime income Z in the hypothetical situation of no income mobility. The latter distribution is formed by assigning the lowest income in every period to the poorest individual in the first period, the second lowest to the second poorest, and so on. Accordingly, the design of the distribution of Z does not alter the marginal period-specific

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<sup>&</sup>lt;sup>5</sup> For a given u, M(u) is the ratio of the mean income of the poorest 100u per cent of the population and the overall mean. By inserting for the Lorenz curve in (3.2) it follows straightforwardly that the scaled conditional mean curve is a representation of inequality that is equivalent to the Lorenz curve.

<sup>&</sup>lt;sup>6</sup> Since  $C_1$ ,  $C_2$  and  $C_3$  have a common theoretical foundation and prove to complement the Gini coefficient with regard to sensitivity to changes in the lower and upper part of the income distribution, Aaberge (2007) calls these three measures Gini's Nuclear Family.

distributions. Since  $F_Y$  can be attained from  $F_Z$  by a sequence of period-specific Pigou-Dalton income transfers, we know that

$$C_{k}(F_{\gamma}) \geq C_{k}(F_{\gamma})$$

with strict equality insofar  $F_Y$  is identical to  $F_Z$ .

In line with Shorrocks (1978), we can then measure income mobility as

(3.3) 
$$M_{k}(F) = \frac{C_{k}(F_{Z}) - C_{k}(F_{Y})}{C_{k}(F_{Z})} = \frac{\sum_{t=1}^{T} b_{t} \frac{\mu_{t}}{\mu} C_{k}(F_{Y_{t}}) - C_{k}(F_{Y})}{\sum_{t=1}^{T} b_{t} \frac{\mu_{t}}{\mu} C_{k}(F_{Y_{t}})}$$

where

$$b_{t} = \frac{\prod_{j=t+1}^{T} (1+r_{j})}{1+\sum_{s=1}^{T-1} \prod_{j=s+1}^{T} (1+r_{j})}, \quad t = 1, 2, ..., T-1; b_{T} = 1.$$

It follows straightforward that  $0 \le M_k \le 1$ . Mobility measures based on (3.3) can be interpreted as the relative reduction in inequality in lifetime income due to changes in individuals' positions in the short-term distributions of income. The state of no mobility  $(M_k = 0)$  occurs when individuals' positions in the short-term income distributions are constant over time.

## 4. Empirical results

This section begins by reporting estimates of inequality in lifetime and current income. This allows us to assess the life-cycle profile in income inequality, and identify the age at which life-cycle bias is minimized. We next decompose the life-cycle bias into income mobility and heterogeneous income profiles. We conclude this section by demonstrating how the essential features of our findings could be revealed from the life-cycle association between the levels of lifetime and current income.

#### 4.1. Inequality in current and lifetime income

Figure 1 graphs the estimated inequality in lifetime income and current income, age 24–63. There are three clear patterns in our results, independent on whether the inequality measure focuses on income

differences that take place in the lower  $(C_1)$ , middle  $(C_2 = G)$ ), or upper  $(C_3)$  part of the income distribution.

First, we see that inequality in current income is quite stable until individuals are in their late 30s, after which inequality increases steadily. This pattern reflects that individuals with high lifetime income experience more income growth over the life-cycle.

Second, we find evidence of systematic life-cycle bias in estimates of inequality based on current income. Recall that  $C_k(F_{Y_t}) - C_k(F_Y)$  measures the *life-cycle bias* in income inequality from using current income at age t as a proxy for lifetime income. Figure 1 shows that the life-cycle bias is minimized when individuals' incomes are measured at age 34-35. Later in the life, the bias increases steadily and inequality in current income at age 60 is more than twice the inequality in lifetime income.

Third, inequality in lifetime income is much lower than what cross-sectional estimates of inequality suggest. For example, Aaberge and Mogstad (2011) study the evaluation of income inequality in Norway over the period 1967-2007. In each year, their sample consists of the entire population of males aged 20–65. The cross-sectional estimates of the Gini coefficient increase from around .26 in 1967 to about .36 in 2006. In contrast, Figure 1 shows that the inequality in lifetime income over this period is less than .19.

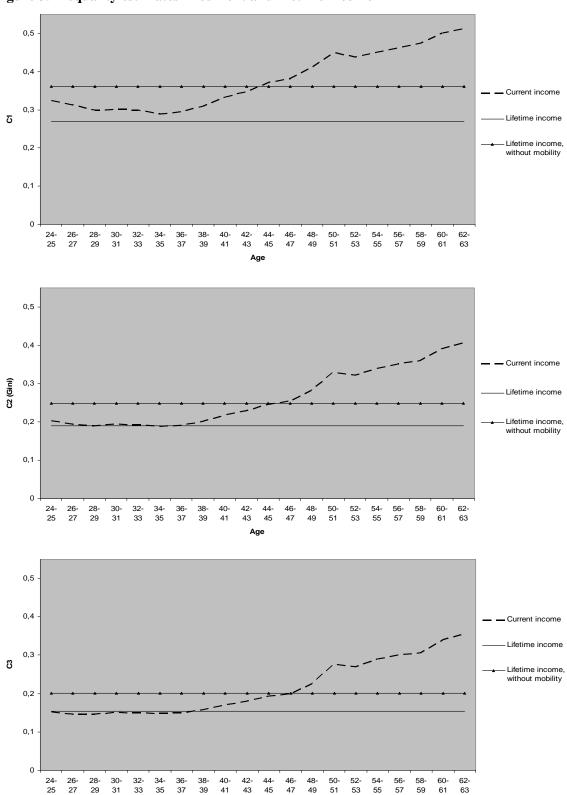
#### 4.2. Decomposition of life-cycle bias

To understand why inequality in current incomes differs from inequality in lifetime incomes, we decompose the life-cycle bias into two components: income mobility (i.e. changes over time in individuals' positions in the current income distributions) and heterogeneous (but non-intersecting) age—income profiles. From (3.3), we get that

(4.1) 
$$C_{k}(F_{Y}) - C_{k}(F_{Y}) = C_{k}(F_{Y}) - C_{k}(F_{Z})(1 - M).$$

Recall that  $C_k(F_Z)$  can be expressed as a weighted average of inequality in current income over the life-cycle. In the absence of income mobility (M=0),  $C_k(F_{Y_t}) - C_k(F_Y)$  will therefore be either zero at all ages or vary in sign over the life-cycle. Because of income mobility, however, inequality in current income may exceed inequality in lifetime income at all ages.

Figure 3. Inequality estimates in current and lifetime income



Age Note: The sample consists of males born 1948-1950. Current income is defined as the average income over the two-year period. Lifetime income is defined as the annuitized value of real income from age 24 to 63. Age refers to the 1949 cohort. Lifetime income without mobility is defined as assigning the lowest income in every period to the poorest individual in the first period, the second lowest to the second poorest, and so on.

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Figure 1 displays  $C_k(F_Y)$ ,  $C_k(F_Z)$ , and  $C_k(F_{Y_k})$  for k=1,2,3. The figure shows that income mobility reduces inequality in lifetime income by about 25 percent, while heterogeneous (but non-intersecting) age—income profiles contributes to upward (downward) life-cycle bias in estimates of income inequality when current incomes are measured late (early) in the working lifespan. Overall, the equalizing effect of income mobility tends to dominate, implying that inequality in current income usually exceeds inequality in lifetime income even when current incomes are measured early in individuals' careers.

#### 4.3 Life-cycle variation in the association between current and lifetime income

To evaluate the appropriateness of the textbook errors-in-variables model, Haider and Solon (2006) analyzed life-cycle variation in the association between current and lifetime income. They found that the slope coefficient in a regression of current income on lifetime income

$$\lambda_{t} = \frac{cov(y_{t}, Y)}{var(Y)},$$

is generally not equal to one, as the textbook errors-in-variables model assumes, varying instead systematically over the life cycle. An important implication is that using current income as a proxy for lifetime income can generate substantial life-cycle bias in regression analysis. Haider and Solon propose a possible remedy: Under the assumptions of the generalized errors-in-variables model, life-cycle bias is eliminated by using current income around the age when  $\lambda_t$  equals one. The procedure of measuring income around a certain age has over the last few years been much used to correct for life-cycle bias applied research, especially in analysis of intergenerational mobility.<sup>7</sup>

We begin by replicating the analysis of Haider and Solon. Our estimates of  $\lambda_t$  start out at a value less than one at the outset of the career, then increase over the life cycle, and exceed one later in the career. In our data set,  $\lambda_t$  is closest to one around when individuals' incomes are measured in their late 40s.<sup>8</sup> However, this finding does not justify using current income in the late 40s as a proxy for lifetime

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<sup>&</sup>lt;sup>7</sup> For studies of intergenerational mobility taking this approach to correct for life-cycle bias, see the review of Black and Devereux (2011).

<sup>&</sup>lt;sup>8</sup> In comparison, Haider and Solon (2006) find that  $\lambda_t$  comes close to 1 when individuals are in their early 40s (and mid 30s), whereas Bohlmark and Lindquist (2006) report that  $\lambda_t$  is approximately 1 when individuals are aged 46-53 (and around age 33). It should be noted, however, that these two studies measure income in logs rather than levels. When measuring income in logs, our estimate of  $\lambda_t$  is equal to 1 around age 40.

income in the measurement of inequality. To see this, note that the life-cycle variation in the association between the Gini coefficient in current and lifetime income can be expressed as

$$G(y_t) - G(Y) = \frac{2cov[y_t, F(y_t)]}{E(y_t)} - \frac{2cov[Y, F(Y)]}{E(Y)}$$

In general,  $\lambda_t = 1$  does not imply  $G(y_t) = G(Y)$ . The former depends on the covariance between individuals' current and lifetime income; the latter depends on how the covariance between individuals' incomes and their position in the distribution of income varies over the life-cycle. Figure 1 demonstrates that this difference is it not merely a theoretical peculiarity, but empirically important: Life-cycle bias in inequality is minimized at age 34-35, whereas current income at the age when  $\lambda_t$  is close to one grossly overstates the extent of inequality in lifetime income

#### 5. Conclusion

Research on the income inequality has a long history in economics. In particular, considerable effort has been directed towards examining how cross-sectional estimates of inequality may overstate inequality in long-run incomes because of income mobility. In contrast, much less attention has been devoted to the life-cycle bias that may arise from the widespread use of current earnings as a proxy for lifetime earnings in measurement of income inequality.

This paper provided evidence on inequality in current and lifetime earnings. We used these results to assess the importance of life-cycle bias in the common practice of using current earnings as proxy for lifetime earnings. Our results showed substantial life-cycle bias in estimates of inequality based on current income. An important implication is that the cross-section estimates of inequality are likely to be sensitive to the age composition of the sample. This means that it is necessary to pay close attention to differences in age composition when comparing estimates of inequality across countries, subgroups, or time. Another key finding is that inequality in lifetime income is much lower than what cross-sectional estimates of inequality suggest. This means that we may need to reconsider how unequal individuals' living standard actually is.

However, caution is in order. Inequality in current and lifetime income for the Norwegian cohorts born 1942-1944 may differ from those for other cohorts or other countries. In particular, we advise readers to exercise due caution in importing our estimates of life-cycle bias to other income or earnings data.

Nevertheless, our study illustrates, at the very least, the importance of addressing life-cycle bias in empirical analysis of inequality using current earnings as proxy for lifetime earnings.

One important issue remains unresolved: What is causing the observed life-cycle variation in inequality? On the one hand, our findings are consistent with age-earnings profiles of education groups that fan out with age, as documented and analysed by for example Bhuller et al. (2011). On the other hand, the rise in inequality in the later part of the life-cycle is consistent with a large body of evidence suggesting an increase in skill premiums since the early 1980s. Clearly, our results suggest a need for more work in this area, with particular attention to separating age and time effects in the life-cycle profile of income inequality.

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