

How Large Are African Inequalities? Towards Distributional National Accounts in Africa, 1990-2017

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October 2019



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This version:
Tuesday 1st October, 2019

Abstract

This paper makes a first attempt to estimate the evolution of income inequality in Africa from 1990 to 2017 by combining surveys, tax data and national accounts in a systematic manner. The low quality of the raw data calls for a lot of caution. Results suggest that income inequality in Africa is very high, and stands at par with Latin America or India in that respect. Southern and Central Africa are particularly unequal. The bulk of continent-wide income inequality comes from the within country component, and the between country component was even slightly reduced in the two last decades, due to higher growth in poorer countries. Inequality was rather stable over the period, with the exception of Southern Africa. Dualism between agriculture and other sectors and mining rents seem to be important determinants of inequality.

*We thank Thomas Blanchet, Léo Czajka, Ignacio Flores and Thanasak Jenmana for their useful advices and helpful comments.

1 Introduction

Despite strong economic growth in many African countries, human development and poverty indicators have not progressed as expected. This phenomenon has fueled a renewed interest for the study of inequality in Africa: it was seen as one of the main causes of the weak poverty-alleviation elasticity of growth.

Is Africa a high-inequality developing region? It has long been thought that Africa was too poor to be unequal, based on its expected position in a worldwide Kuznetsian inverted U-curve. Even if the Kuznets curve is no longer considered as a well-grounded regularity, African inequality levels are still debated today² Analyses on African inequality levels are typically made on the basis of household surveys, which provide a rich set of socio-economic information on inequality but also have several important limitations when it comes to comparing actual income inequality levels across countries. From one country to another, household surveys may inform on different types of welfare concepts (e.g. disposable income, taxable income or consumption) and may use different ranking concepts (individual, household or equivalence scales). Moving from one concept to another might radically modify the income distribution in a country and the level of inequality observed. While studying inequality across countries or regions, it is thus necessary to compare distributions as systematically as possible. In addition, household surveys tend to misreport top incomes due to sampling and non-sampling errors, which typically leads to underestimation of inequality levels. Average income or consumption levels reported in surveys are often at odds with values reported in the national accounts. In order to compare inequality levels between Africa and other regions, relying on household surveys only may be too limited (Alvaredo et al. (2016)).

Using a combination of sources is likely to provide a better approximation of Africa's true inequality levels and how it compares to other regions. This paper makes a first attempt in that direction, by estimating the evolution of pretax income inequality in the continent from 1990 to 2017, combining surveys, tax data and national accounts in a systematic manner. Our main finding is that Africa stands out as an extreme income inequality region by international standards: with a top 10% national income share of 54% and a bottom 50% share below 10%, Africa has the highest gap between average incomes of the top 10% and incomes of the bottom 50% (Figure 1)³. This overall

²According to Borhat and Naidoo (2017), the average Gini coefficient in Africa, based on household surveys, is 0.43 in 2014, whereas it is 0.39 in the rest of the developing world. There is high heterogeneity across countries however, and this high average figure is driven by seven highly unequal countries, (with a Gini above 0.55), located mostly in Southern Africa: Angola, Central African Republic, Botswana, Zambia, Namibia, Comoros, South Africa. In terms of trends, the reported average African Gini has declined (it was 0.48 in the early 1990s), but this fall is largely due to trends in relatively low-inequality countries.

³Appendix Figure A.1 presents Gini coefficient values for the same regions. The African Gini stands out as the second highest in the world at 0.67, after the Middle East (Gini of 0.69), but above Gini in Brazil and India (0.63 and 0.60, respectively)

high inequality level masks relatively large regional variations, which can partly be explained by a limited set of factors, namely differences in productivity and employment in the agriculture and the service sector. One can think of several structural mechanisms explaining why African inequality (as measured by the top 10% to bottom 50% average income ratio) appears to stem at a higher level than other high-inequality world regions, including India and Brasil. In particular, one should emphasize the role played by the high-income mining and extractive sector, together with the existence of a very low productivity sector in agriculture. The legacy of colonialism, both in terms of property structure (particularly in Southern Africa) and in terms of educational stratification, should also be stressed. We hope future research will be better able to disentangle the exact weight played by these different factors in accounting for the very high levels of African inequality.

The rest of this paper is organized as follows. We provide a brief overview of the literature on the distribution of income in Africa (section 2). We then explain and implement a statistical methodology for combining (noisy) household survey data with (scarce) income tax data and with (imperfect) national accounts (section 3). This allows us to compare inequality in Africa to the rest of the world, look at regional patterns within Africa, explore structural determinants of African inequality, assess the (absence of) link between redistribution policies and inequality, and finally describe individual countries trajectories during the two last decades (section 4). Let us warn the reader that this is still exploratory work. Data availability, quality and comparability are still very unsatisfactory.

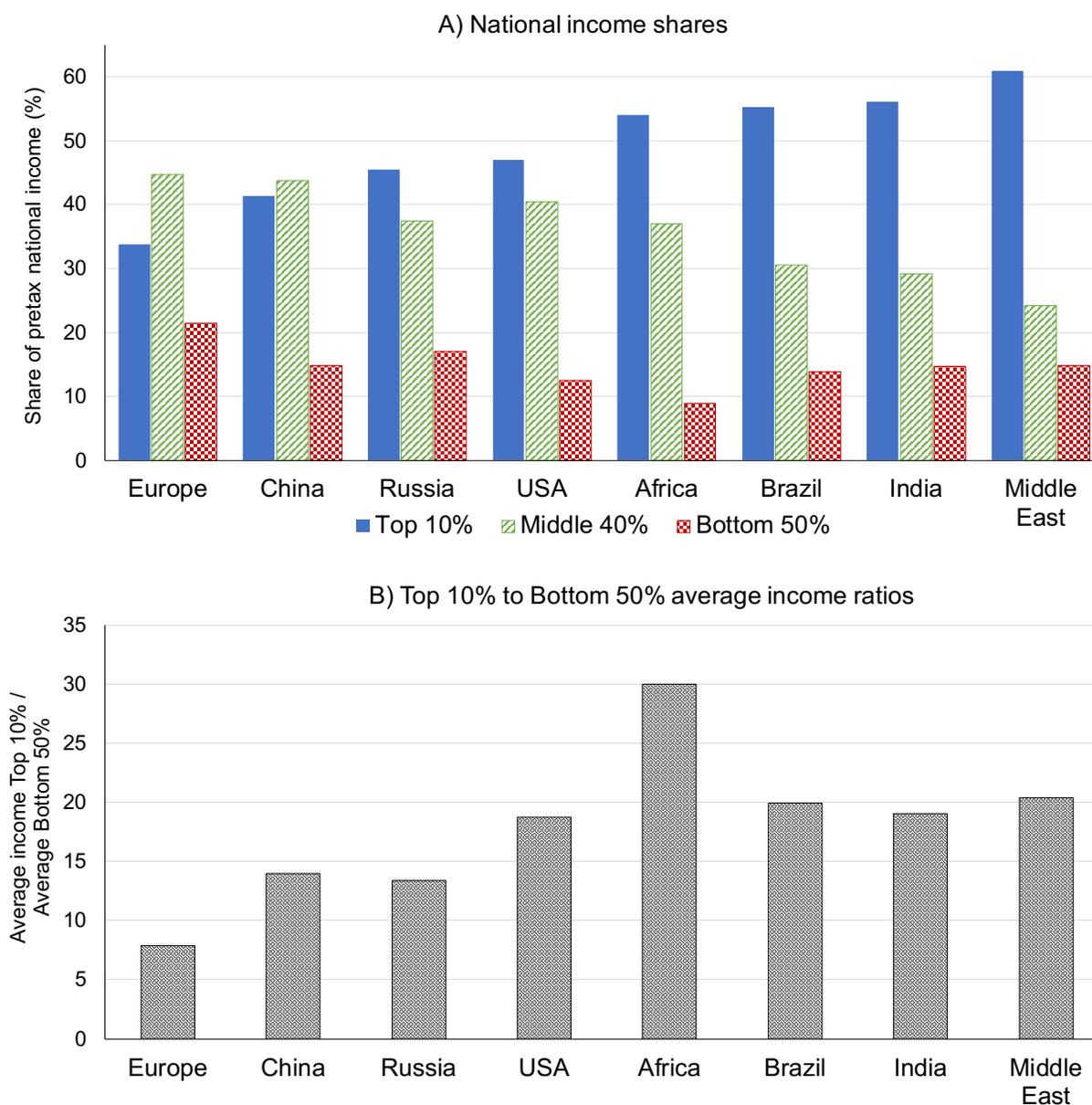
2 Existing literature on the distribution of income and growth in Africa

Research on the drivers of inequality in African countries is hindered by the lack of good-quality data, both on the distribution of welfare and on other economic or society indicators, but a few potential lines of explanations have been tackled.

A first strand of the literature explores the links between the so-called “sub-optimal” structural transformation experienced by the vast majority of African countries and the evolutions of welfare inequality. In theory, it can be expected that the growth of labor-intensive sectors, such as manufacturing or labor-intensive services activities will boost wage employment and reduce the spread of the income distribution (Bhorat and Naidoo, 2017). But unlike Europe during the Industrial Revolution or East Asia more recently, African economies didn’t experience a gradual shift from agriculture to manufacturing. Instead, the decline of the share of agriculture in the Gross Domestic Product (GDP) was absorbed by services or mining industries and less productive sectors⁴. This implied that the decrease in agricultural employment was entirely absorbed in services or in urban

⁴See the *World Development Indicators*, 2014; McMillan, Rodrik, and Inigo (2014) estimated that structural change in Africa between 1990 and 2005 made a negative contribution to overall economic growth by 1.3% per year on average.

Figure 1: Inequality levels across world regions, 2017



Source: Authors' computations based on WID.world (2019), see WID.world/methodology for region specific data series and notes. Distribution of pretax income per adult. Growth is assumed to be distribution-neutral between 2016 and 2017 in Africa.

unemployment, as mining industries are very capital-intensive. Polarization of the services sector increased because of the development of informal activities, with very poor working conditions and low income. Overall, it resulted in a gradual "urbanization of poverty", as informal employment or urban unemployment spread (Ravallion, 2002). The urban-rural gap didn't really decrease, because of the persistent urban bias of public spending, and because skilled urban residents were more able

to exploit the opportunities brought about by liberalization. Besides, inequalities increased within the rural and within the urban sectors even when they decreased between sectors (Cornia, 2017).

The impact of African growth models on income inequality was previously studied by looking at the joint evolution of sectoral value-added shares and Gini coefficients (Cornia, 2017). Gini fell where the value-added share of modern agriculture, labor-intensive manufacturing and modern services stagnated or rose (for example in Ethiopia, Cameroon, Madagascar); it increased in countries with stagnant land yields, a drop in manufacturing, a rise of the resource enclave and skill-intensive services, and urban informalization.

Focusing on agriculture, a strand of the literature argues that raising agricultural productivity could reduce inequality, through the increase of rural income, and because it would lead to a diversification of rural activities towards non-agricultural activities, and consequently favor industrialization (Gollin, 2010; Pingali, 2010; Estudillo and Otsuka, 2010). Indeed, a series of empirical papers underline the role of agricultural modernization in triggering growth, development, and reducing poverty and inequality (Imai and Gaiha, 2014; Bourguignon and Morrisson, 2002; Christiaensen, Demery, and Kuhl, 2011). This can be particularly relevant for Africa, which hasn't fully completed its agricultural transition yet. However, some authors underline that agricultural modernization has more impact on poverty than on inequality (Imai and Cheng, 2016; Herault and Thurlow, 2009). Some stress that equal land distribution is key to enable agriculture to reduce poverty and inequality Lipton (2009), Griffin, Khan, and Ickowitz (2002), and Manji (2006).

The influence of extractive industries on inequality has been extensively tackled. In theory, extractive industries favor income inequality, through both economic and institutional channels. According to Borat and Stanwix (2017), extractive industries yield limited employment creation, and only for skilled labor. Besides, the high cost of entry leads to monopolistic or oligopolistic market structures, that favor higher pricing and excessive profits. A boom in the resource price can lead to the appreciation of the local currency, which can then disadvantage employment-intensive and often export-reliant sectors, or attract the best workers, draining them from the other sectors (the so-called "Dutch disease"). This sector can also lead to the crowding out of non-resource investment (Papyrakis and Gerlagh, 2004), or hamper financial sector growth (Isham et al., 2003), and it tends to fuel an urbanization process without industrialization, by sustaining the existence of "consumption cities" (Gollin, Jedwab, and Vollrath, 2016).

The debate on the relationship between governance quality and resource dependence is not settled: a first strand of literature argues that resource abundance is a blessing for countries with good institutions and a curse for the others (Mehlum, Moene, and Torvik, 2006). But several works underline that resource dependence can have a negative impact on democratic consolidation (Jensen and Wantchekon, 2004). Other papers show the opposite causality: a country with weak institutions is more likely to become resource-dependent (Brunnschweiler and Bulte, 2008). According to Borat

and Stanwix (2017), the limited quantity of resources can either lead to political capture of resource rents, if the granting process is not transparent, or to monopolists with high influence over economic policies, in particular multinationals with the bargaining power to secure tax incentives. Besides, resources revenues are at high risk of being taken out through illicit channels, very often trade mispricing, even more with financial globalization and liberalization.

According to the same authors, in Africa many countries recorded a rise in the mining rents/GDP ratio and can now be qualified as “resource dependents”, meaning that 25% or more of export revenue is derived from natural resources⁵. This natural resource boom has helped to jump-start an infrastructure and construction boom, and in turn created many jobs that require low-level skills, but there are concerns about the temporary nature of these jobs. When they explore the link between resources rents and Gini coefficients, these authors find that there are a handful of resource-dependent countries with very high levels of inequality, which implies that resource-dependent countries have a higher risk of being highly unequal. However, focusing on the correlation between Gini coefficients and resource-dependency over all African countries yields weak or inconclusive empirical evidence.

The links between institutions, public policies and inequality, was also explored by the literature. In this regard, colonial legacy is a central issue (Walle, 2009). In colonial societies, a minority of settlers held a very large fraction of the wealth and lied at the top of income distribution (Alvaredo, Cogneau, and Piketty, 2019). High wages were paid in a tiny formal sector formed by administrations and a few firms specialized in the trading of natural resources exports (Cogneau, Dupraz, and Mesplé-Somps, 2018). This dualistic structure partly survived after independence, even after settlers’ departure, giving rise to an ‘oligarchic bourgeoisie’. Comparing five countries with household survey data carefully made comparable, Cogneau et al. (2007) find that income dualism between agriculture and other sectors accounts for a large part of the difference in inequality between them; dualism is higher in the three former French colonies (Cote d’Ivoire, Guinean and Madagascar) compared to the two former British (Ghana and Uganda); on the same data, Bossuroy and Cogneau (2013) show that intergenerational mobility between agriculture and other sectors is also lower in former French colonies, because of employment dualism, i.e. the concentration of non-agricultural occupations in large cities. Cogneau (2007) argues that the decentralized management of colonial empires also produced large spatial inequalities.

In terms of redistribution policies, available data studied by Odusola (2017) shows that fiscal space (tax revenue as a share of GDP) has been increasing over time. Institutions played a significant role in this increase: the Open Budget Index⁶ is highly correlated to fiscal space, which was also boosted

⁵See in particular the *World Development Indicators* and the United Nations Conference on Trade and Development (UNCTAD)

⁶The Open Budget Index is issued from the Open Budget Survey, which measures budget transparency, participation and oversight.

by debt relief. However, fiscal space remains low compared to the rest of the developing world, and the distributional effectiveness of fiscal policy remains highly questionable in most countries. Indeed, Odusola shows that the difference between the gross Gini (before taxes and transfers) and the net Gini (after taxes and transfers) decreased for most countries, implying that the redistribution system has lost efficiency. The author found other hints of tax regressivity, in particular a positive correlation between fiscal space and Gini coefficients. He concludes that African governments still lack fiscal space and economy-wide institutions to ensure social protection and have a positive distributional impact.

According to Borat and Steenkamp (2017), there has been a general increase in social protection expenditure according to available data, but social protection coverage, quality and level of assistance still remain very limited. It is more pronounced in Southern African countries. The expenditure increase was very variable across countries and was not correlated to economic growth. Current social protection expenditure is highly related to the quality of democratic governance (as captured by the Mo Ibrahim Index⁷) and to resource-dependence (non-resource dependent countries spend more on average).

The comprehensive review of social protection in Africa by the African Development Bank et al. (2011) has shown the positive impact of many specific transfer programs on poverty and inequality reduction, suggesting that social protection can be a key driver of inequality reduction. Borat and Steenkamp (2017) look at the correlation between inequality reduction (measured by the variation between pre-transfer and post-transfer Gini coefficients) and various characteristics of social protection. They find no clear impact of public social spending on inequality, but a positive impact of both pro-poor coverage of social protection, and transfer average amount on inequality reduction.

Regarding educational inequalities, the quality of education is still low, despite a general progress in primary schooling enrolment rates (Borat and Naidoo, 2017). In addition, except in some Southern and Northern African countries, there has been no substantial progress in terms of secondary education, and important enrolment differentials by income groups persist. Consequently, the stock of human capital is still inadequate, whereas the demand for skilled labor increased. This implies high wage premiums for a few skilled workers, which fuels income inequality.

⁷The overall Ibrahim Index of African Governance score aggregates four categories: safety and rule of law, participation and human rights, sustainable economic opportunity and human development. A score of "0" indicates autocracy, whereas a score of "100" indicates democracy (Mo Ibrahim Foundation, 2014).

3 Data and methodology

In this section, we present the data sources used to estimate income inequality in Africa and our methodology to combine them in a systematic way. Section 2.1 presents our data sources. Section 2.2 develops our method for converting consumption inequality estimates into income inequality estimates. Section 2.3 explains how we correct for under-representation of top incomes in surveys. Section 2.4 outlines how we reconcile our results with national accounts.

3.1 Data sources

3.1.1 Survey data

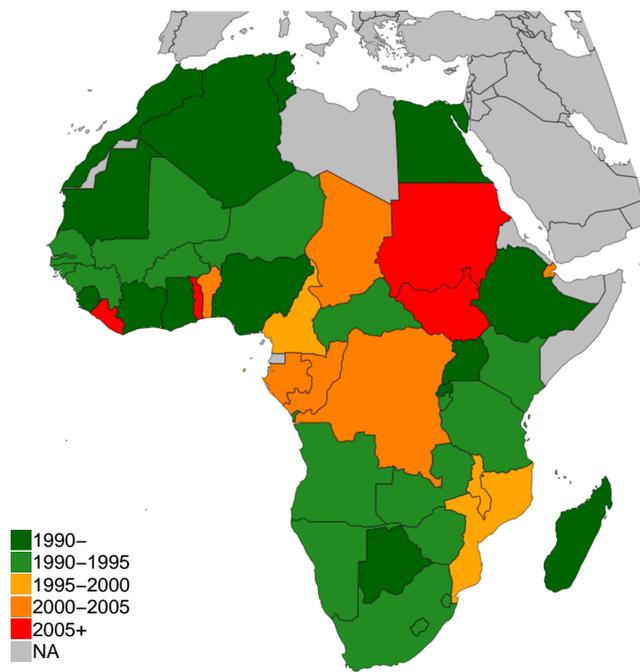
Our primary source of data consists in survey tabulations from the World Bank which are made publicly available on the PovcalNet website.⁸ These tabulations provide information on the distribution of consumption per capita for various groups of the national population. We use the Generalized Pareto Interpolation techniques developed by Blanchet, Fournier, and Piketty (2017) to harmonize these tabulations and obtain information on consumption at the percentile level. We complete our database with eight surveys from Côte d'Ivoire, which have been used by Czajka (2017) for his study on the evolution of income inequality in the country since the mid-1980s⁹. Finally, we use additional surveys conducted in Ghana (1988, 1998), Guinea (1994), Madagascar (1993) and Uganda (1992), which were compiled by Cogneau et al. (2007) and are especially useful to model the relationship between consumption inequality and income inequality (see section 2.2). In order to have a broader vision of the latter in developing countries, we complete our model by exploiting surveys available from Jenmana (2018) for Thailand (2001-2016) and from Chancel and Piketty (2017) for India (2005, 2011).

Figure 2 shows that there are large variations in data coverage across African countries. In Morocco, Nigeria and Madagascar, surveys have been more or less conducted on a regular basis since the early 1980s. In countries of central Africa, by contrast, only one or two surveys are available, generally after 2000. Overall, if we pool together our surveys and interpolate between years, we are able to cover about 60% of the continental population in the early 1990s, 80-90% from 1995 to the early 2010s.

⁸<http://iresearch.worldbank.org/PovcalNet/povOnDemand.aspx>

⁹See also: Cogneau, Houngbedji, and Mesplé-Somps (2016) and Cogneau, Czajka, and Houngbedji (2018)

Figure 2: Time and space coverage of survey data sources



Source: authors' computations. Surveys providing information on the national distribution of consumption or income before 1980 are only available in Morocco, Nigeria and Madagascar.

3.1.2 Fiscal data

Contrary to Europe, where fiscal data can be used to correct for the under-representation of top incomes in a number of countries (Alvaredo et al., 2018), publicly available tax tabulations are close to non-existent in Africa. We use South African tax tabulations covering the 2002-2014 period provided by Alvaredo and Atkinson (2010) and updates, as well as a similar tabulation covering the formal sector in 2014 Côte d'Ivoire available from Czajka (2017), to study to what extent accounting for the “missing rich” affects income inequality estimates. We complete our study of top incomes by extending our analysis to other developing countries thanks to Thai and Indian tax tabulations provided by Jenmana (2018) and Chancel and Piketty (2017) respectively. Given the lack of income tax data in many African countries, we make strong but transparent assumptions in order to correct survey data on the basis of comparable countries where both fiscal and survey data are available. As additional tax data becomes available, we will revise the series accordingly. In the meantime, given that top-end corrections associated to tax data have a comparable and sizable magnitude in most countries, we feel that it is more adequate to apply a simple and transparent correction method to countries with missing tax data rather than to make no correction at all. Our correction

procedure is described in a more detailed manner.

3.1.3 National accounts

We account for inequalities between African countries by using macroeconomic series available from the World Inequality Database (<http://wid.world>) and covering the 1950-2017 period. These series were constructed by Blanchet and Chancel (2016) by combining various sources of historical data. In line with the Distributional National Accounts approach to income inequality (Alvaredo et al., 2016), we use these series to make our estimates consistent with the national income per adult in purchasing power parity.

3.2 From survey consumption to survey income

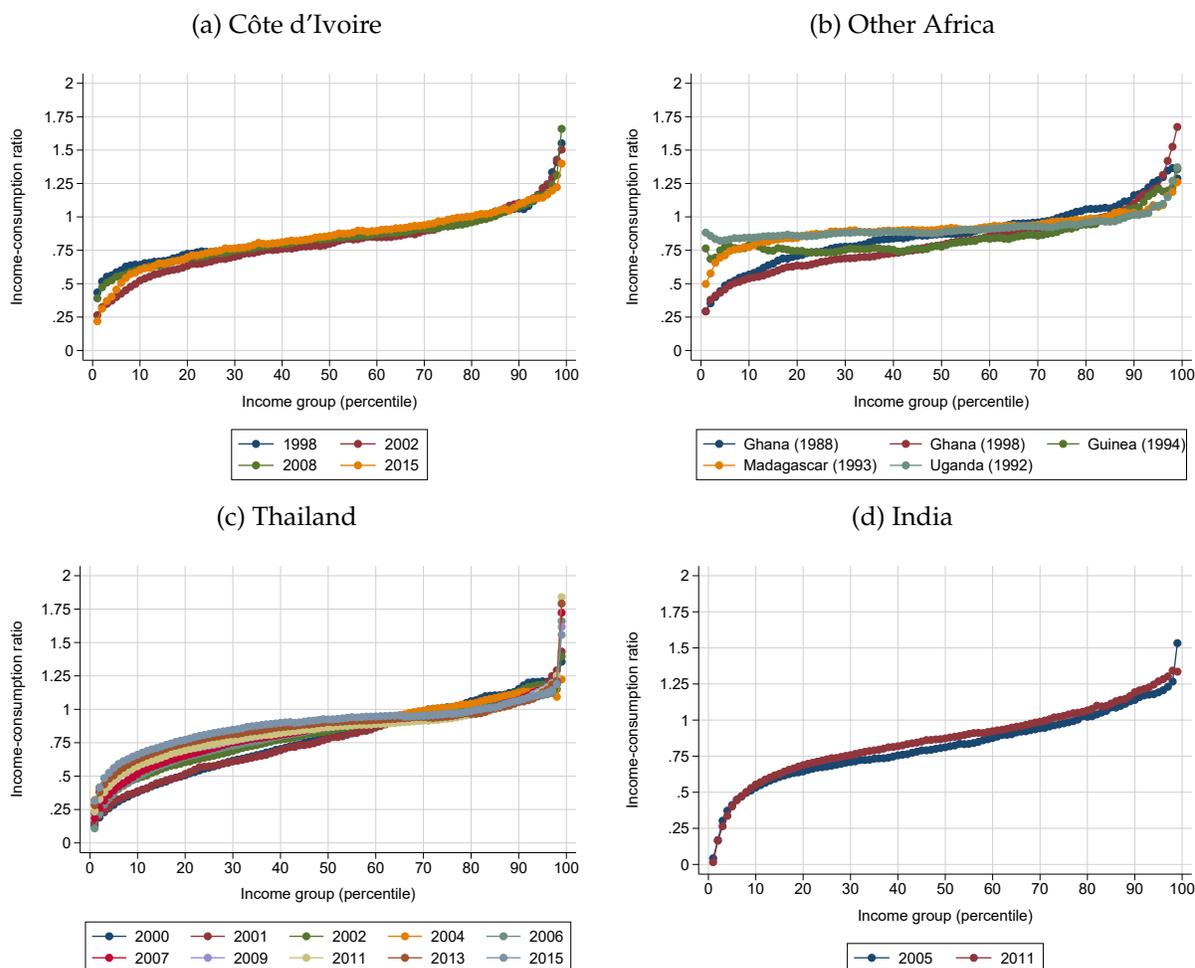
The first issue with available inequality statistics in Africa is that they rely almost exclusively on consumption. This makes systematic comparisons between developed and developing countries difficult, since inequalities of economic resources are most often measured by pre-tax or post-tax income in the former. From a theoretical perspective, it makes sense to believe that income inequalities are likely to be higher than consumption inequalities, as (i) high-income earners tend to spend a lower share of their earnings on household expenditures than poorer individuals; (ii) income has a transient component that some households are able to smooth in order to maintain a stable level of consumption; (iii) in many surveys income is measured less accurately than consumption and measurement errors can inflate inequality. This gap is likely to be large at the bottom of the distribution where (i) households incurring transient negative income shocks can be found; (ii) households with understated incomes can be found. It is also likely to be important at the top of the distribution, since (i) the very rich tend to save a large proportion of their current earnings; (ii) households benefitting from transient positive income shocks can be found; (iii) households with overstated incomes can be found. Yet, very little is known on how income-consumption profiles vary across countries and across time.

Our primary objective is to make consumption and income inequality estimates comparable. As such, we are interested in comparing income distributions rather than individual behaviours. If we know to what extent consumption is higher or lower than income for all quantiles of a given distribution, then we can exploit this relationship to “transform” consumption inequality estimates into income inequality estimates. In other words, our aim is to model income-consumption profiles $c_1(\cdot)$ of the form:

$$c_1(p) = \frac{Q^I(p)}{Q^C(p)}$$

Where $Q^I(\cdot)$ is the quantile function associated with a given distribution of income, $Q^C(\cdot)$ is the

Figure 3: Empirical consumption-income profiles in eight countries



Source: authors' computations.

quantile function associated with a given distribution of consumption, and $p \in [0, 1]$. If, given $c_1(p)$, we only have access to a consumption distribution $\hat{Q}^C(p)$, we can then impute a corresponding income distribution defined by $\hat{Q}^I(p) = c_1(p) * \hat{Q}^C(p)$.

We start by looking at the empirical shape of $c_1(p)$ for all percentiles in countries and years for which we have reliable data. Following our definition of $c_1(p)$, computing income-consumption ratios is straightforward: it simply consists in dividing the bracket average of each percentile of the income distribution by its consumption counterpart. In order to make profiles comparable, we systematically normalize average income or consumption to 1. Notice that since our aim is to use $c_1(p)$ as a multiplicative factor, the ratio of aggregate consumption to aggregate income is irrelevant: what matters is how $c_1(p)$ varies with p .

Figure 3 plots income-consumption profiles in Côte d'Ivoire, Ghana, Guinea, Madagascar, Uganda, Thailand and India for various years. In nearly all surveys, the relationship between income inequality and consumption inequality is distinctively S-shaped. Average income is generally substantially lower than average consumption for the bottom quintile of the population. The ratio of income to consumption then increases more or less linearly up to percentiles 80 and 90, before rising exponentially at the top of the distribution. This is fully consistent with the intuitive mechanisms outlined above: poorer individuals tend to smooth their consumption, while the very rich tend to save a significant proportion of their current earnings. As a result, consumption inequalities are generally lower than income inequalities.

In order to study more precisely how consumption-income profiles vary across surveys, we propose to formulate $c_1(\cdot)$ parametrically by using a scaled logit function of the form:

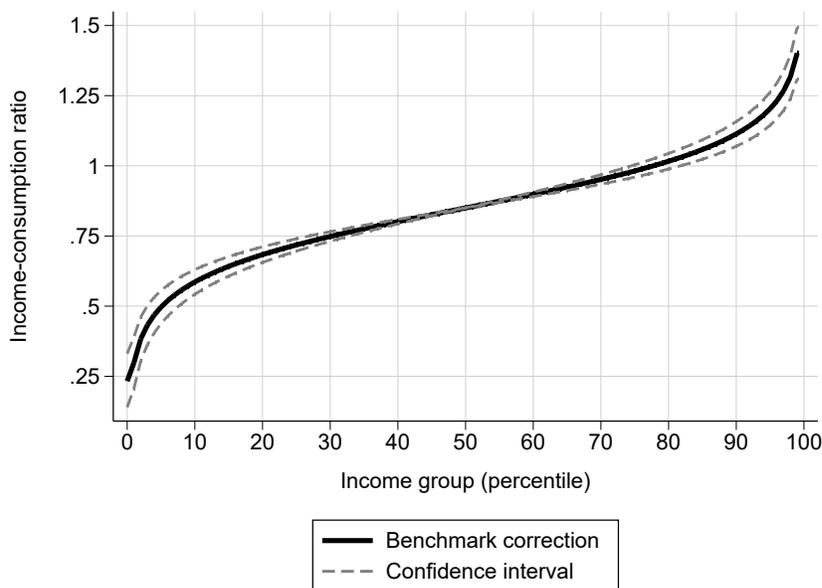
$$c_1(p) = \alpha + \beta \log\left(\frac{p}{1-p}\right) \quad (1)$$

For $p \in (0, 1)$. α is a constant which determines the starting point of the curve. It is completely irrelevant to our imputation problem, since multiplying the quantile function by α only affects the overall mean of the distribution. β is our parameter of interest: it determines how fast the ratio of income to consumption increases with p and is therefore a direct measure of the extent to which consumption inequalities are lower than income inequalities.

Table A.1 reports the results of $\hat{\alpha}$ and $\hat{\beta}$ estimated by least squares, along with the corresponding adjusted R-squared. In nearly all cases, our scaled logistic function provides an excellent fit of income-consumption profiles, explaining more than 90% of variations in the data. Strikingly, our coefficient of interest $\hat{\beta}$ is always positive and varies moderately across surveys. Consumption series underestimate income inequalities most in Thailand at the beginning of the 2000s ($\hat{\beta} = 0.16$), and underestimates them least in Madagascar and Uganda at the beginning of the 1990s ($\hat{\beta} = 0.05$ in Madagascar and $\hat{\beta} = 0.06$ in Uganda). Yet, beyond these two extremes, a majority of correction profiles range between 0.1 and 0.14.

Our objective is to provide a reasonable approximation of income inequality in Africa by transforming all available consumption distributions into income distributions. As such, it makes sense to define three theoretical profiles reflecting the variability in $\hat{\lambda}(\cdot)$ observed in the data. We thus define three scenarios which provide "confidence intervals" for our income inequality estimates. In our benchmark scenario (scenario A henceforth), we use $\hat{\beta}_A = 0.12$; in scenario B, we correct inequalities more moderately by imposing $\hat{\beta}_B = 0.1$; and we correct them more strongly in scenario C by using $\hat{\beta}_C = 0.14$. Figure 4 plots our three correction profiles (setting $\alpha = 0.85$ to make them easily comparable with observed profiles). Moving from consumption inequality to income inequality amounts to reducing the income of bottom 10% earners by 25% to 50%, while increasing

Figure 4: Theoretical income-consumption profiles



Source: authors' computations. Theoretical consumption-income profiles correspond to logistic functions of the form $Q_i(p) = \alpha + \beta_i \log \frac{p}{1-p}$ for $i \in A, B, C$. We set $\alpha = 0.85$ and $\beta_A = 0.12, \beta_B = 0.08, \beta_C = 0.14$.

the average earnings of the top 1% in similar proportions.

3.3 From survey income to fiscal income

The second correction we apply to our survey distributions consists in correcting the average income of top earners. It is well-known that the rich are under-represented in surveys, for both sampling and misreporting issues. Many researchers have attempted to correct for this bias by combining surveys with tax tabulations or microdata, which only cover a limited part of the population but generally provide better coverage of the very top of the distribution. While corrections based on fiscal data systematically yield higher inequality levels, little is known on the typical shapes of these corrections and their variations across countries.

Following the method used for consumption, our aim is to use existing data to define “plausible” profiles correcting income levels at the top of the distribution. In the African case, correcting for the under-representation of the rich in surveys is particularly challenging. To our knowledge, the only research paper combining surveys and tax data at the time of writing is Czajka (2017). The paper exploits recently released tax tabulations from Côte d’Ivoire, and shows that the average income of the top 1% is typically underestimated by about 75% in the private sector. In other developing countries, the correction profiles of top incomes obtained from matching surveys with

fiscal data vary greatly across studies. In Brazil, Morgan (2017) has found that the average fiscal income of the top 1% is typically 1.5 to 3 times higher than in surveys, with variations across years. Corresponding figure were found to be between 1.5 and 2.5 in Thailand (Jenmana, 2018) and as high as 3.5 in Lebanon (Assouad, 2017).

We look at variations in the underestimation of top incomes in Africa by bringing together surveys and fiscal tabulations from Côte d'Ivoire (Czajka, 2017) and South Africa (Alvaredo and Atkinson, 2010). For South Africa, we match the 2008, 2010 and 2012 surveys compiled in the Luxembourg Income Study (LIS) with the fiscal income series provided by Alvaredo and Atkinson (2010) and subsequent updates available from the World Inequality Database. We then use the method developed by Blanchet, Flores, and Morgan (2018) to combine surveys and tax data in order to get corrected survey income distributions. The method essentially compares the distributions of survey income and fiscal income, and finds a merging point where they cross. It then reweighs survey observations so that the information on top incomes in the survey matches that observed in the fiscal data.

Exactly as in the case of consumption and income, our objective is to study “survey-fiscal” profiles $c_2(\cdot)$ of the form:

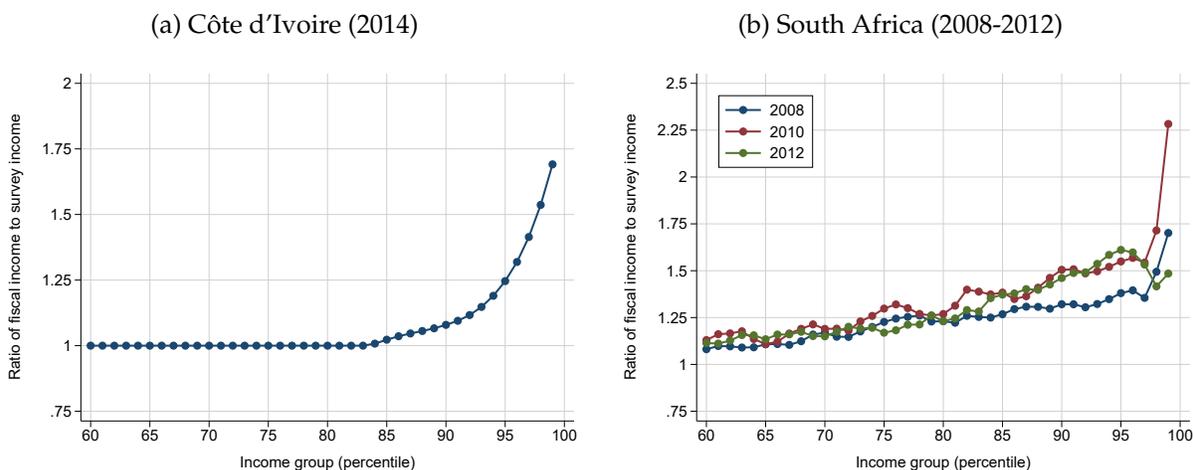
$$c_2(p) = \frac{Q^F(p)}{Q^I(p)}$$

Where $Q^I(p)$ is the quantile function associated with the distribution of income observed in the survey, and $Q^F(p)$ is the quantile function of the distribution obtained after correcting for the under-representation of top incomes. The South African profiles can be computed by dividing the average incomes observed in the corrected distributions by their corresponding values in the surveys. In Côte d'Ivoire, the ratio of fiscal income to survey income by percentile is obtained from Chancel and Czajka (2017).

Figure 5 plots the survey-fiscal profiles in our two countries of interest. In Cote d'Ivoire, the ratio of corrected income to survey income is close to 1 before the 90th percentile, and then increases exponentially. In South Africa, the correction starts much earlier (before the 80th percentile), but rises more moderately. In both countries, surveys tend to largely underestimate top incomes, especially at the very top of the distribution. Correcting for this bias amounts to increasing the average of the top 1% by between 50% and 125%.

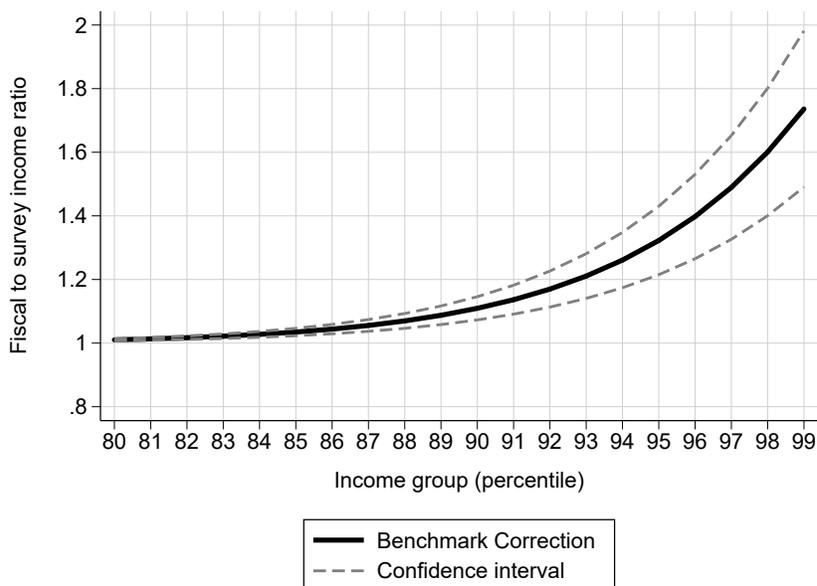
The correction profile of top incomes can be formally conceptualised as depending upon two dimensions: the size of the group which is corrected, and differences in the corrections applied to top earners within this group. One way to formulate these two dimensions parametrically is to

Figure 5: Empirical survey-fiscal profiles in Côte d'Ivoire and South Africa



Source: authors' computations.

Figure 6: Theoretical survey-fiscal profile



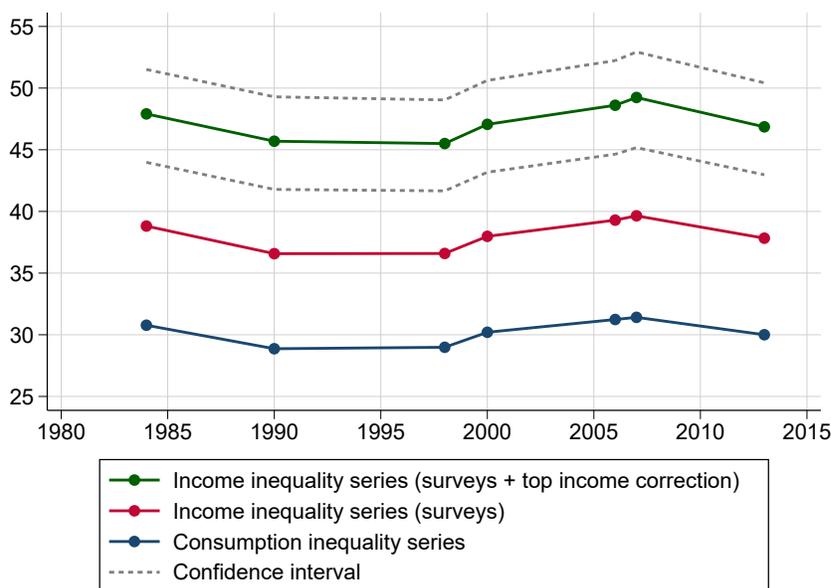
Source: authors' computations.

model survey-fiscal profiles by the quantile function of the Lomax (or Pareto Type II) distribution:

$$c_2(p) = \mu + \sigma(p^{1/\gamma} - 1)$$

For $p \in [0, 1]$. μ is a constant which determines the starting point of the curve; as in the case of

Figure 7: Top 10% income share in Morocco, 1984-2014: from consumption inequality to corrected income inequality



Source: authors' computations.

consumption-income profiles, it is irrelevant to our problem. Since it makes sense to let $c_2(p)$ take 1 before a certain percentile p_0 , one can set $\mu = 1 + \sigma$, so that $c_2(0) = 1$ and:

$$c_2(p) = 1 + \sigma p^{1/\gamma}$$

σ is the scale parameter. It controls the slope of the curve: the higher σ , the higher top incomes are underestimated by surveys. γ is the shape parameter: as it decreases, the slope becomes more convex, so that a smaller fraction of top incomes is corrected.

While it is difficult to find regularities in the correction of top incomes given the lack of comparable data across countries and across years, it is still our belief that some sort of correction is better than no correction at all. In our benchmark scenario, we set $\sigma = 0.9$ and $\gamma = 0.05$. We then let σ vary from 0.6 to 1.2. As figure 6 shows, this approximately corresponds to rescaling incomes exponentially above the 80th percentile (γ) and multiplying the average income of the top 1% by between 1.5 and 2 (σ). These bounds are in line with the different corrections observed in Côte d'Ivoire and South Africa. They are arguably sufficiently large to represent plausible variations in the correction of top incomes in Africa across countries and across time. If anything, this correction profile is likely to be a lower bound: in other developing countries such as Brazil, Lebanon or Thailand, it was not uncommon to find that the top 1% average was underestimated by a factor of

2 to 3 (Morgan, 2017; Assouad, 2017; Jenmana, 2018).

Figure 7 plots the evolution of the top 10% share in Morocco before and after correcting for conceptual discrepancies and underestimation of inequality at the top. According to consumption inequality estimates provided by PovcalNet, the highest decile received about 30% of total consumption every year between 1984 and 2013, with no clear trend over the period. Moving from consumption inequality to income inequality increases this value to 35-40%, and correcting top incomes increases it further to above 45% in our benchmark scenario. Hence, consumption-based measures from PovcalNet tend to underestimate the share of national income accruing to top 10% earners by as much as 40%.

3.4 From fiscal income to the national income

Under the assumption that our method for improving the measurement of income inequality is correct, the resulting dataset we obtain corresponds to the distribution of household income – that is, the sum of compensation of employees, mixed income and property income received by the household sector in the national accounts. To reach the national income and obtain figures on individual incomes which are consistent with macroeconomic growth, we have to make assumptions on the distribution of unreported income components. These mainly include the taxes on production received by the general governments and the retained earnings of corporations, which can constitute a significant fraction of the national income in both developed and developing economies Alvaredo et al. (2017).

In developed countries, and in some emerging economies, the levels of unreported income components can generally be observed from national accounts, and various methods can be used to impute these components indirectly on the basis of household surveys. Unfortunately, this is not the case for most African countries, where national accounts are still in their infancy. As a result, we do not have access to reliable data on unreported income. We choose to distribute the gap between surveys and the national income proportionally to individual income. This has the advantage of keeping the overall distribution unchanged, while making average incomes and growth rates more comparable across countries and over time. We stress that this is a conservative assumption: in most existing distributional national accounts studies, the imputation of unreported income leads to higher inequality levels, mainly because retained earnings are concentrated at the top the distribution (e.g. Blanchet, Chancel, and Gethin, 2018; Piketty, 2018; Jenmana, 2018; Morgan, 2017). As better national accounts data, survey microdata and tax data become more systematically available, our estimates can easily be improved and updated to account for such discrepancies.

4 The distribution of income and growth in Africa, 1990-2017

4.1 How unequal is Africa ?

4.1.1 Inequality in African countries

Is Africa a low or high inequality continent? The question, as simple as it may be, is difficult to answer due to the lack of comparability between data sources. Applying, to the extent possible, Distributional National Accounts methods to Africa yields estimates that are more in line with those recently developed for rich and emerging countries. We should stress at the onset that such estimates are still far from perfect and will greatly be improved thanks to the release of administrative data, as has already been the case in a few African countries (namely Côte d'Ivoire, Sénégal, Tunisia and South Africa).

Novel estimates suggest inequality remains very high in most African countries. The income earned by the top 10% of the distribution ranges from 37% in Algeria to 67% in Botswana (Figure 8), while the bottom 40% is at most 14% in Algeria, and is about 4% in South Africa (Figure 9).

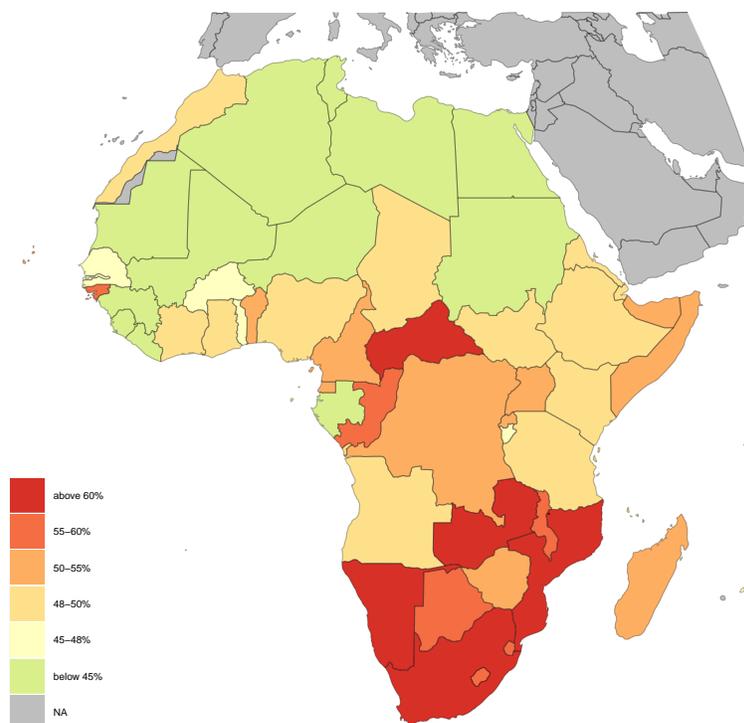
Significant regional differences appear across the African continent. Southern Africa is clearly the most unequal region: the share of national income earned by the top 10% is the highest in South Africa and Botswana, where it respectively amounted to 65% (in 2014) and 67% (in 2009), while the bottom 40% earned 4% of national income in these two countries.

Income inequality is on average less stringent in Central Africa, but it remains very high: 56% of national income accrued to the top 10% income earners in Congo in 2011 for instance, while the bottom 40%'s share was 7%. Eastern Africa's countries appear a bit less unequal, especially at the bottom: in Kenya in 2015 for example, the top 10% owned 48% of national income and the bottom 40% owned 9%.

Income inequality tends to decrease as we move towards the North and the West of the continent. In Sierra Leone in 2011, the top 10% owned 42% of national income, and the bottom 40% owned 12%, and its neighbors display similar income shares. The lowest inequality levels can be found in Northern Africa: in Algeria, the least unequal country in Africa, in 2011 37% of national income was captured by the top 10% of the distribution, while the bottom 40% owned 14%. However, one must here take into account data quality and data limitations, the publication of tax data and better households surveys might unveil higher inequality levels in North African countries, more comparable to the Middle East countries (in particular for Algeria, Lybia and Egypt).

Before discussing potential drivers of such regional differences, we focus on African-wide inequalities.

Figure 8: Top 10% income shares in Africa, most recent available years (2006-2017)

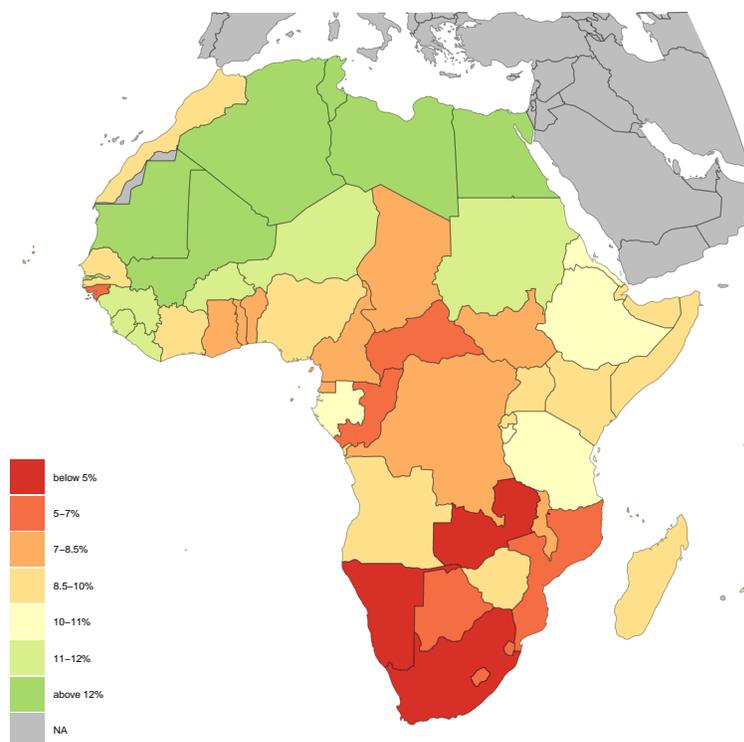


Source: WID.world (2019). Estimates combine survey, fiscal and national accounts data.

4.1.2 Inequality among Africans

Africa stands out as one of the regions with the highest level of income inequality. According to our methodology, top 10% of Africans capture 54% of national income vs. 9% for the bottom 50% (Figure 10). In an international perspective, the top 10% income share is of 34% in Europe (550m individuals), 41% in China (1.4bn individuals), 47% in the US (330m individuals), 55% in Brazil and the rest of Latin America (640m individuals), 56% in India (1.3bn individuals), 61% in the Middle East (420m individuals). A particularly striking fact of the pan-African distribution is the extent of the gap between the top 10% and the bottom 50% income share. Average incomes of the top 10% are about 30 times higher than those of the bottom 50%, well above the value found in other extreme inequality regions (the gap is around 20x in other extreme inequality regions such as the Middle East, India or Brazil). This particularity sheds light on the dual nature of the pan-African income distribution, with extremely low incomes at the bottom and relatively high incomes at the top. Preliminary work using similar correction methods for Pan-Asian and Pan-Latino American regions suggest that top 10% to bottom 50% average income ratios are around 20-25 in these regions.

Figure 9: Bottom 40% income shares in Africa, most recent available years (2006-2017)



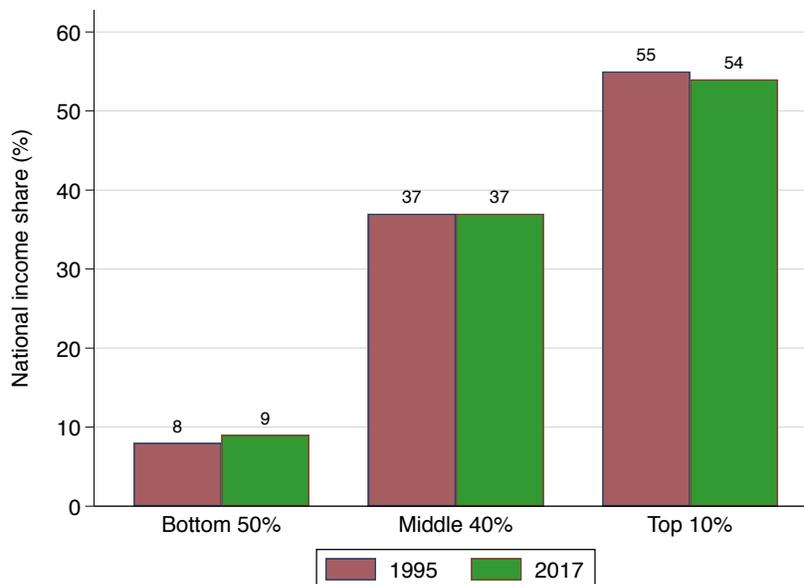
Source: WID.world (2019). Estimates combine survey, fiscal and national accounts data.

While there are uncertainties surrounding such comparisons, the high continental inequality level found in Africa as compared to other emerging world regions seems to hold with or without corrections applied to household surveys and described in Sections 3.2 and 3.3. Upcoming work on regional inequality levels in Asia and Latin America will allow us to better investigate this issue.

We find that the high level of inequality in Africa seems relatively stable over time. At the pan-African level, inequality has only slightly decreased over the past decades. The top 10% income share was 56% and the bottom 50% income share was 9% in 1995. As we discuss below, little movements in inequality levels at the pan-African level mask significant evolutions within countries.

What contributes to African inequality: is it mostly inequality within African countries or mostly inequality in average income levels? If we decompose overall inequality between what is imputable to national average income disparities (between-country inequality) and what is due to inequality within countries, it clearly stands out that inequality within countries explain most of pan-African inequalities. If there were no inequality between countries, but keeping constant current within-

Figure 10: Evolution of the pan-African distribution

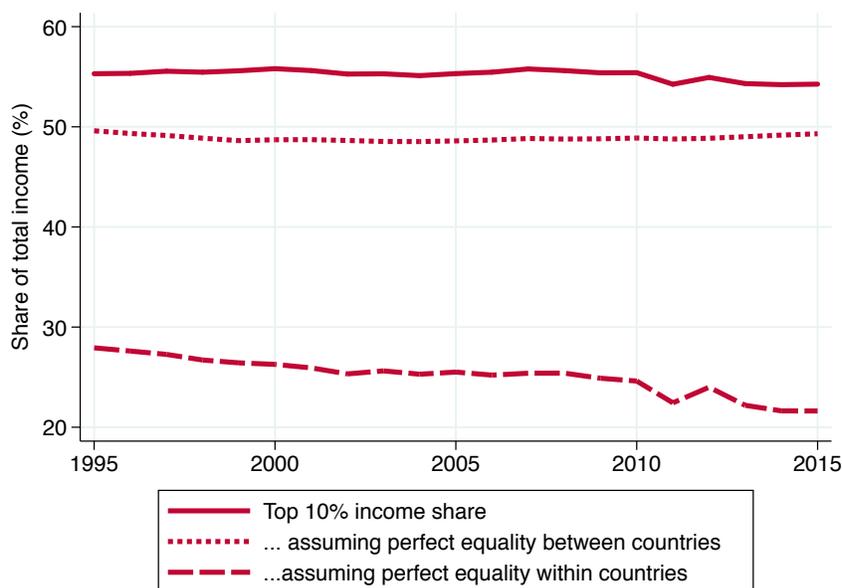


Source: WID.world (2019). Estimates combine survey, fiscal and national accounts data.

country inequality levels, the top 10% income share in Africa would be only slightly different from what it actually is (it would be 50%, vs. 55% in reality). Conversely, if within each country, all individuals were perfectly equal, but keeping national average income differences as they currently can be observed across Africa, the top 10% income share would substantially drop, to 21%. A Theil decomposition of African inequality levels shows that 25% of African inequality can be attributed to between-country differences and as much as 75% to within-country inequality.

Interestingly, the slight decline in overall inequality is entirely due to the dynamics of between-country inequality (Figure 11). Since 1995, there has been a tendency towards less average income inequalities between countries. This reduction is caused by several phenomena. Since the years 1990, several countries, located at the middle of the African ladder of national income per capita, such as Nigeria, Morocco, Ghana, Angola, Tunisia, Namibia or Lesotho, saw their average income increase significantly (we discuss the growth strategies followed by some of these countries below). On the other hand, the average income of Africa's richest countries (Algeria, South Africa, Libya for example), stagnated in the years 1990, and increased moderately in the years 2000. However, the poorest countries didn't experience any significant average income rise. This is why, assuming perfect equality within countries, the top 10% share decreased more than the bottom 50% increased. The dynamics of between and within country inequality in Africa contrast sharply to what can be observed at the global level, in Europe or in Asia. At the global level (see section 2), within-country

Figure 11: Decomposition of Pan-African Inequality: top 10% income share (1995-2015)



Source: WID.world (2019). Estimates combine survey, fiscal and national accounts data.

dynamics dominate between-country evolutions since 1980 and since the early 1990s: between-country convergence is not sufficient to cancel the impact of rising within country-inequality on global inequality. In Europe, contrary to Africa, most of the evolution in pan-European inequalities stems from within-country dynamics. Turning to Asia, the huge rise of inequality recorded in China and India (which amount to about two-thirds of the regional population) over the past four decades also meant that a very large share of total inequality explained by within-country evolutions.

The relative stagnation of within-country inequality at the pan-African level in fact masks a large array of inequality trajectories followed by African countries over the past decades. Part of this stagnation could also be due to noisy survey data and national income estimates, which would make it more difficult to detect marginal variations. Before assessing these individual trajectories, we analyze potential drivers of the significant variation of inequality levels across countries and their regional patterns underlined above.

4.2 How to account for different inequality patterns across Africa?

What causes such high inequality levels in Africa as compared to the rest of the world? This issue remains open and it is particularly challenging to address because of strong data limitations, as well as of the specificity and diversity of Africa's economic and political structures, shaped by its

recent history and colonial heritage. In the following section, our objective is not to explain the diversity of inequality levels found in Africa, but merely confront theories regarding growth and inequality to available data. We explore the role of structural economic factors and growth patterns on the one hand, and public redistribution policies on the other.

4.2.1 Growth models and income inequality

Unlike Europe during the Industrial Revolution or East Asia more recently, African economies did not experience a gradual shift from agriculture to manufacturing. Instead, the decline of the share of agriculture in the Gross Domestic Product (GDP) was absorbed by services (whether formal or informal) or mining industries. This type of structural transformation is considered to be a major symptom of Africa's shortcomings in terms of economic development by many researchers.

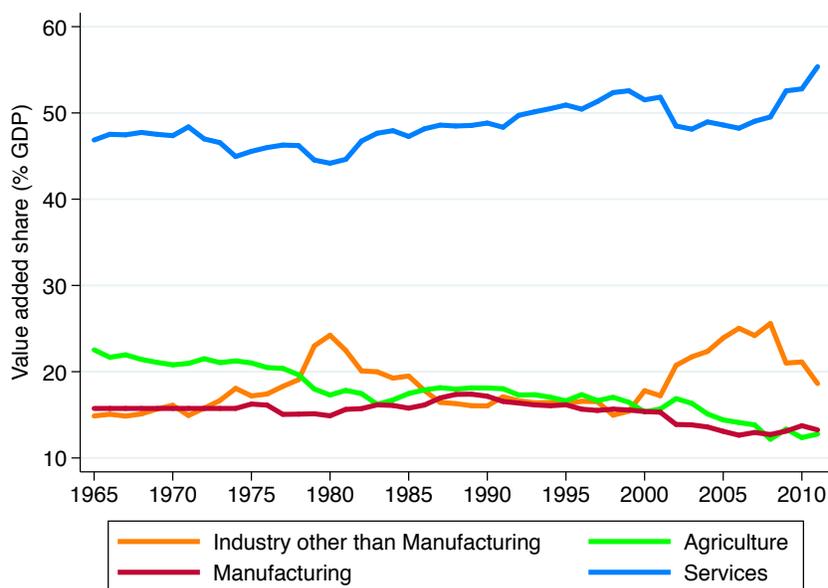
The impact of African growth models on income inequality was previously studied by looking at the joint evolution of sectoral value-added shares and Gini coefficients (Cornia, 2017). Gini coefficients appeared to be positively correlated with mining and services activities', suggesting a rise in inequality associated to mining activities. Conversely, such analyses highlighted a reduction of inequality with the rise of agriculture, labor-intensive services (trade, restaurants, hotels), transportation and construction activities.

We propose to re-examine these relationships with a new approach. On the one hand, we use the new African income inequality dataset, which addresses some of the limitations of existing distributional data (and yet which remains largely imperfect). Instead of focusing on Gini coefficients, which can mask important inequality dynamics, we also prefer to focus on top and bottom (or middle) income group shares. On the other hand, we focus on sectoral productivity and sectoral employment, rather than solely at value-added shares, as this provides a richer set of information on the structure of African economies.

In Africa, the decrease in agricultural employment is almost entirely absorbed in services or in urban unemployment, as mining industries are much less labor-intensive than manufacturing. Natural resources exports rather "fuel" an urbanization process without industrialization, by sustaining the existence of "consumption cities" (Gollin, Jedwab, and Vollrath, 2016). The process comes with a gradual "urbanization of poverty", as informal employment or urban unemployment spread (Ravallion, 2002).

We base our theoretical expectations on a quite standard Kuznets curve (Robinson, 1976). The three sectors are ranked according to their productivity of labor: agriculture has the lowest, and industry the highest, with services lying in between — keeping in mind that services is perhaps the most polarized sector between informal subsistence activities and skilled white collar jobs. For a given relative productivity of labor in agriculture, inequality should be the highest at intermediate

Figure 12: Evolution of sectorial value-added shares in Africa (1965-2012)



Source: Authors' computations based on Africa Development Indicators (World Bank)

levels of employment. For a given level of employment, inequality should also be higher the lower agricultural productivity. Besides, the agricultural sector itself can be equal or unequal depending on the way land is distributed¹⁰. Yet relative productivity and employment are not independent, and demand linkages complicate the Kuznets process, as the internal term of trade between agriculture and services endogeneously adjusts (Bourguignon, 1990). Besides, the degree of dualism in urban labor markets can make things even more complex: a large low productivity service sector can be a source of inequality as well as a small highly productive one. Last, according to previous research, extractive industries should in theory favor income inequality, even though for the moment empirical evidence is mixed (see section 2).

In order to better investigate the interplay between income inequality and economic structure, we analyzed the relationship between the top 10% income share and agricultural employment share, productivity of labor in agriculture relative to the non-agricultural sector, an indicator of informality¹¹, unemployment, and the share of mining rents on GDP. We first use principal component analysis to identify which components are mostly related to the top 10% income

¹⁰On relative agricultural productivity and land size distribution see Bourguignon and Morrisson (1998)

¹¹Our indicator of informality is constructed as the residual of a regression of the share of self-employed on the share of agriculture in employment, its square, and the unemployment rate.

share (see Figure A.2). We retain the two first components¹², and use automatic hierarchical tree classification to form groups of countries based on these components (see Figure A.3).

Three groups of countries emerge from the classification. The most unequal one — with an average top 10% share of 58% — is mostly composed of ten countries from Southern and Central Africa¹³. Most have a large services sector (42% of employment on average), as well as high levels of informal employment and open unemployment (21% on average). Mining rents are substantial in many countries (18% of GDP on average) and agricultural productivity is relatively low (except in Namibia and South Africa). The second group of 22 countries¹⁴ is less unequal, with an average top 10% share of 51%. These are countries where agriculture makes a very large share of employment (72% on average) but has very low relative productivity, and where conversely the service sector is small and shows relatively high remunerations, which often corresponds to administration workers. The third and last group of 16 countries displays even lower inequality levels, with an average top 10% share of 45%.¹⁵ Like the first group, these countries are relatively urbanized and the service sector gathers 43% of employment. In contrast with the first group however, mining rents are limited, as well as open unemployment, and the relative productivity of agriculture is much higher (more than threefold the one of the first group and of the second group). The second and third groups roughly correspond to two different sides of the Kuznets curve: high share of agriculture in employment combined with rather low productivity, or conversely intermediate share combined with higher productivity.

These findings confirm the important role of dualism in explaining inequality, and they are again suggestive that mining rents are also inequality increasing. Yet they have two important limitations. First, the analysis might partly capture regional effects: the most unequal categories are also the ones with a majority of countries from Southern and Central Africa. Inequality variation across African regions are also the results of various historical factors, such as specific colonial legacies, past land distribution, or the lasting impact of strong racial inequalities in Southern Africa. Historical causes and structural economic factors should not be confused, and distinguishing their respective role should be undertaken in future research.

Moreover, sectoral structure is not able to account for within-country inequality variation over time. A potential reason for that might be the lack of substantial variation over time of these structural factors in many countries. Most importantly, the quality of income shares, employment and productivity data remains questionable: the lack of historical data, combined with potentially

¹²These two components account for 56% of the information in the correlation matrix.

¹³Angola, Botswana, Congo, Gabon, Lesotho, Mozambique, Namibia, South Africa, Swaziland, Zambia

¹⁴Burkina Faso, Burundi, Cabo Verde, Cameroon, Central African Republic, Chad, Comoros, Djibouti, DR Congo, Ethiopia, Guinea, Guinea-Bissau, Madagascar, Malawi, Mauritania, Morocco, Niger, Rwanda, Senegal, Tanzania, Uganda, Zimbabwe

¹⁵Algeria, Benin, Cote d'Ivoire, Egypt, Gambia, Ghana, Kenya, Liberia, Mali, Mauritius, Nigeria, Sao Tome, Sierra Leone, Sudan, Togo, Tunisia

recurrent measurement error, prevents satisfying analysis of within-country evolutions. The collection of good quality welfare and economic data should thus be given full priority.

Re-examining the link between African growth models and inequality can have several implications for public policies. They support the call for a renewed interest in agricultural productivity enhancement. Its impact on inequality is indeed also tightly linked to the distribution of land rights. Property rights and land access reforms are thus needed and the proper articulation of these two types of policy is a central issue. Access to land is a particularly serious problem in Zimbabwe and South Africa for instance, where the agrarian reform failed to solve the issue.

However, the impact of agricultural productivity enhancement on inequality might itself be mitigated by unequal urbanization process and by the development of a large informal services sector. This is why making services growth more egalitarian should also be a major concern of policy-makers. In particular, two major causes of high inequality in this sector should be tackled: its high degree of informality, and the poor performances of the schooling system, especially at the secondary and tertiary levels, which enable the development of high human capital inequality.

Other dimensions of the “African” structural transformation, such as the growth of extractive industries, are also likely to contribute to unequal growth. Among the most resources-dependent countries, some have indeed experienced an inequality increase, like Angola or Chad, others, like Algeria followed the opposite pattern. This issue is thus complex, and should be tackled in more depth. To that purpose, better quality data is needed: data cover only a short time period in Chad, Angola, or Nigeria for instance, and thus limits our understanding of the impact of extractive industries on inequality.

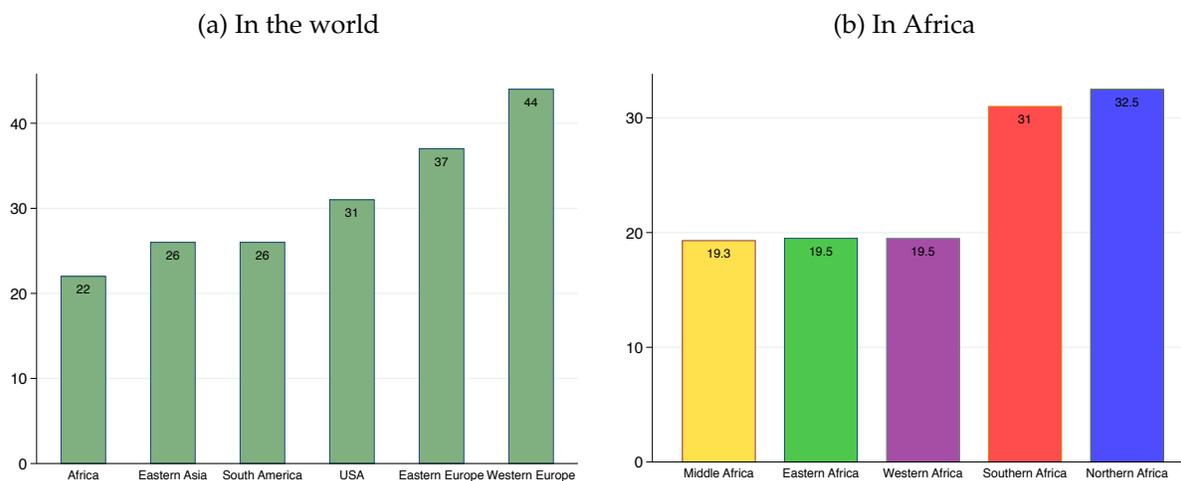
Working on the redistribution of the unequal benefits of growth should also be on the agenda, especially in a continent where tax and redistribution systems are mostly regressive. There is indeed no easy way to develop equitable growth patterns, especially in sectors with high capital intensity like extractive industries.

4.2.2 Redistribution Policies and Inequalities in Africa

Taxation and redistribution policies can have a significant impact on the distribution of income. Most African countries have a substantial margin of improvement in this respect, in terms of fiscal space and tax progressivity on the one hand, and on the redistributive efficiency of government expenditure on the other hand.

In terms of government revenue, Africa is clearly lagging behind all developed and many developing world regions (Figure 13). However, this average figure hides a wide variety of situations (Figures 13 and 14).

Figure 13: General Government Revenue in 2018 (% of GDP)



Source: World Economic Outlook (International Monetary Fund)

Note: General government revenue consists of taxes, social contributions, grants receivable, and other revenue

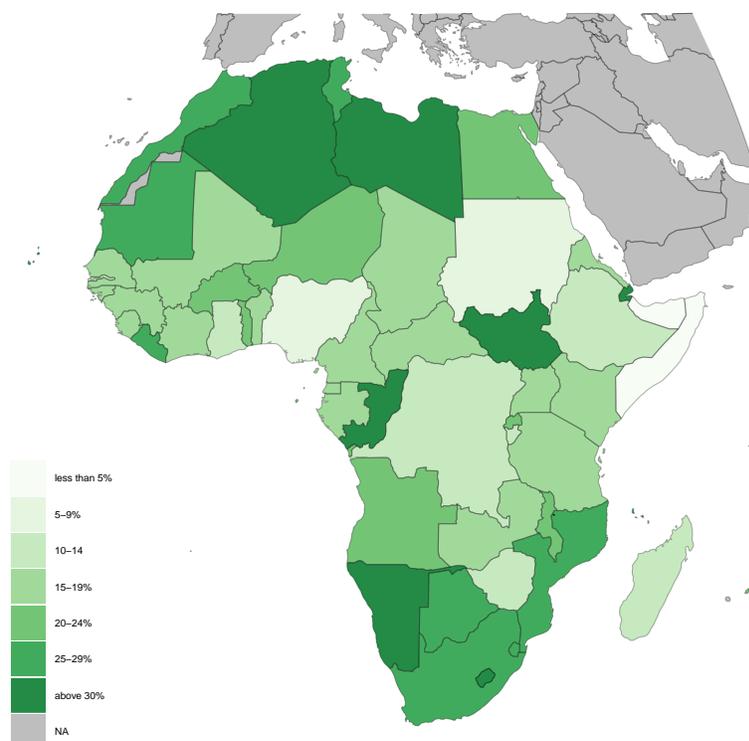
A first group of countries, in Central and Eastern Africa, is characterized by a very low government revenue (below 15% of GDP). In a second group, composed of most Western and Eastern African countries, the total government revenue lies between 15 and 25%, around the region's average. The majority of Southern and Northern countries constitutes a third group, where government revenue is above 25% of GDP, and for a handful of countries (such as Namibia and Algeria), it is higher than 35% of GDP.

In spite of this diversity, most African countries have low government revenue, and consequently limited fiscal space. This hinders their capacity to provide good quality public services, such as health services or education, and to finance social protection and social transfers, and thus limits importantly their influence over income inequality. Fiscal space can be improved through several channels, among which more efficient fight against tax evasion, better inclusion of informal activities into the tax system, or progressive taxation.

Looking simultaneously at the income captured owned by the bottom 40% (Figure 15) and total revenue, it appears that there is no clear correlation between governments' affluence and low income inequality.

Oduola (2017) studied the relationship between Gini coefficients and tax revenue, and concluded that most African tax systems were actually regressive. Indeed, looking at personal income tax top rates (Figure 16), one can see that they are lower than in the developed world in most African countries. For a quarter of the countries for which data is available, personal income top tax rates amount to 25% or less: this concerns many Western African countries, as well as Angola, Guinea, Egypt, and Madagascar, among others. For half of countries studied, top personal income tax rates

Figure 14: General government revenue in Africa in 2018 (%GDP)



Source: World Economic Outlook (International Monetary Fund)

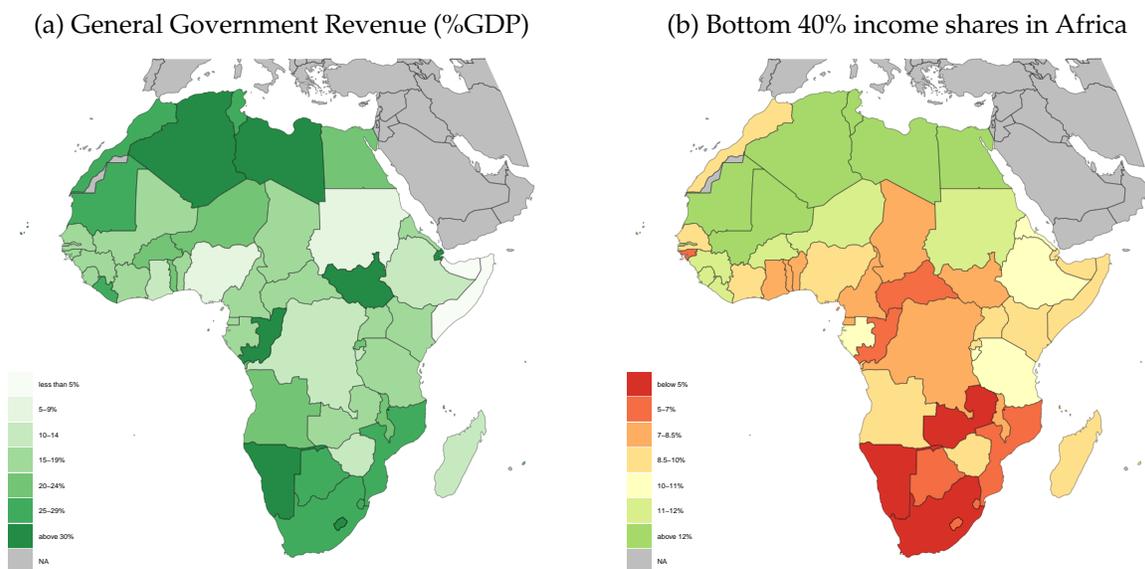
Note: General government revenue consists of taxes, social contributions, grants receivable, and other revenue

lie between 30 and 40%. Eight countries only, located in Central, Southern and Northern Africa, have top rates equal to 40% or higher, that is to say similar to top rates in rich countries (such as France, the United Kingdom, and the United States since the beginning of the 2010s).

The lack of historical data on African tax systems prevents us from analyzing precisely the relationship between the evolution of taxation and inequality. Figure 16 provides a map of personal income tax top rates in 2018. No particular relationship can be found between personal top income tax rates and inequality.

The contribution of governments' revenue to inequality reduction also heavily relies upon the efficient use of public resources, but most African governments have still a substantial margin of improvement in this respect. Public resources can be used to provide good quality public services and to fund of a social assistance and protection system that reaches the poorest parts of the population. From what can be seen on Figure 17, social protection and assistance coverage, as well as public services funding, still remain limited in many of the region's countries. In the majority of countries, total government expenditure on education is below 5% of GDP, its average

Figure 15: General Revenue and Income Inequality



Source: World Economic Outlook (International Monetary Fund) and authors' computations.

Note: General government revenue consists of taxes, social contributions, grants receivable, and other revenue. Data is from 2018 for government revenue, from most recent available years for bottom 40% shares.

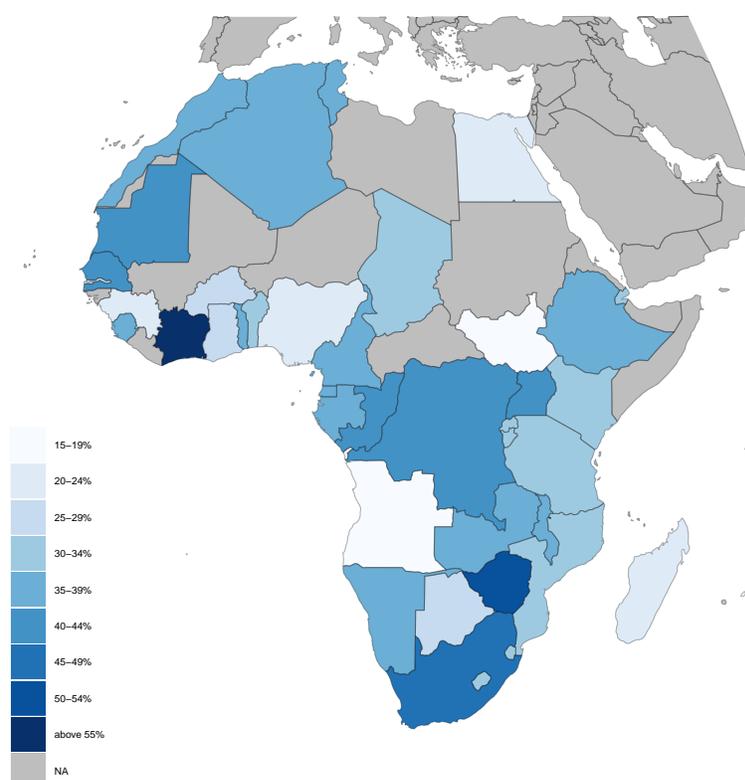
level in developed countries. This is particularly true in Central and Eastern Africa, but also in comparatively rich countries like Egypt or Algeria. Regarding social protection, only a fifth of countries where data is available, mostly located in the South and the North, provide social insurance, social safety nets, or unemployment benefits to more than 45% of their population. This figure was 54% in Brazil in 2015, and 63% in China in 2013 for instance¹⁶.

Redistribution remains very limited in African countries. This issue should be given full attention, because redistribution systems can be key to reduce income inequality. On the one hand, progressive taxation can strongly influence post-tax and pre-tax income inequality, especially at the top of the distribution. Its impact on post-tax income distribution is straightforward, but its role in shaping pre-tax income distribution is often overlooked. Progressive tax systems can limit capital accumulation, and consequently reduce future capital income flows. Besides, they can have a strong influence on top income earners incentives to bargain for high remunerations: when top income tax rates are high, it becomes less worth it for top income earners to bargain for high wages (Piketty, Saez, and Stantcheva, 2014).

On the other hand, public services and social protection, when they are accessible to the poorest part of the population, can have a significant incidence on the distribution of income, especially at the middle and the bottom. This impact can be either direct, through the increase of poor people's

¹⁶Source: *World Development Indicators*

Figure 16: Personal Income Tax Top Rates in 2018



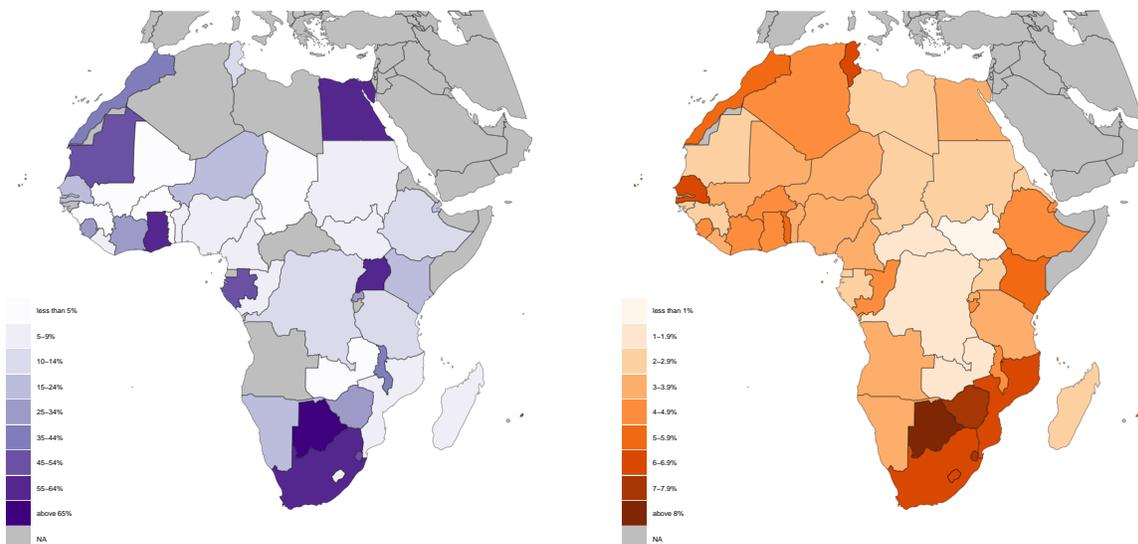
Source: Source: Deloitte, Guide to fiscal information: Key economies in Africa, 2018; Ernst & Young, 2018-19 Worldwide Personal Tax and Immigration Guide, 2019

income by social transfers for instance, or indirect, through the formation of human capital thanks to the provision of medical care, education, and other public services. In previous research, African Development Bank et al. (2011) provided a comprehensive review of social protection in Africa, and demonstrated that it can have a significant impact on poverty and inequality. In South Africa, among other examples, the implementation of a comprehensive system of social grants has helped to reduce post-tax inequality, while the implementation of cash transfers reduced poverty incidence by 4.3% in Namibia. These encouraging results underline the strong potential of social protection, and the urgent need to improve its quality in Africa.

The quality of public services can strongly impact income inequality through its influence on education and health inequalities. Recent works have shed light again on the interplay between educational and economic inequalities in developed countries (Chetty et al., 2014). This issue is particularly interesting in Africa, where despite a substantial rise of primary enrolment rates in the last decades, the low post-primary education enrolment rates and the poor quality of educational systems remain important (Bhorat and Naidoo, 2017). They are likely to fuel strong human capital inequalities, that result in high wage premiums for the few qualified workers, and thus increase

Figure 17: Social protection and public services in Africa

(a) Coverage of social transfers (% population) (b) Government expenditure on education (% GDP)



Source: World Development Indicators (World Bank)

Note: - Coverage of social protection and labor programs (SPL) shows the percentage of population participating in social insurance, social safety net, and unemployment benefits and active labor market programs. Estimates include both direct and indirect beneficiaries.

- Government expenditure on education (current, capital, and transfers) corresponds to all expenditure by the general government (general government usually refers to local, regional and central governments), including expenditure funded by transfers from international sources to government.

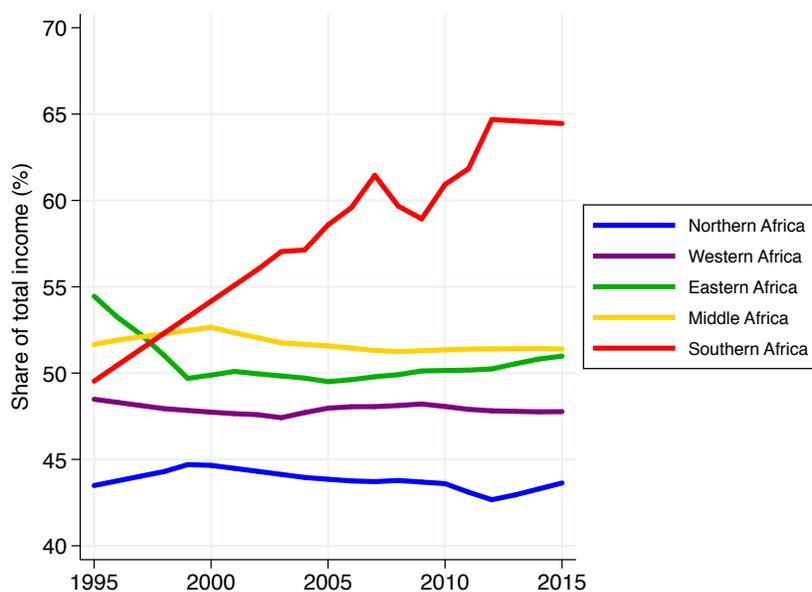
- Data is from the most recent year available for each country.

income inequality, especially in the service sector. The provision of good quality education, not only at primary level, should thus be given full priority.

Overall, it appears that most African countries have still significant progress to make regarding income redistribution, from the increase of fiscal space to the improvement of the progressivity of taxes, the implementation of efficient social protection systems and the provision of good quality public services. These issues are all the more pressing as existing research suggests that improvements along these margins could be key drivers of inequality reduction in Africa.

Beside these potential common factors, current inequality levels are shaped by each country's specific economic, political, and social history. We focus now on these individual trajectories and their diversity, bearing in mind data limitations.

Figure 18: Evolution of regional top 10% income shares in Africa



Source: Authors' computations.

Note: Weighted averages using total population as weight

4.3 Heterogeneous trajectories: income inequality dynamics since the mid-1990s

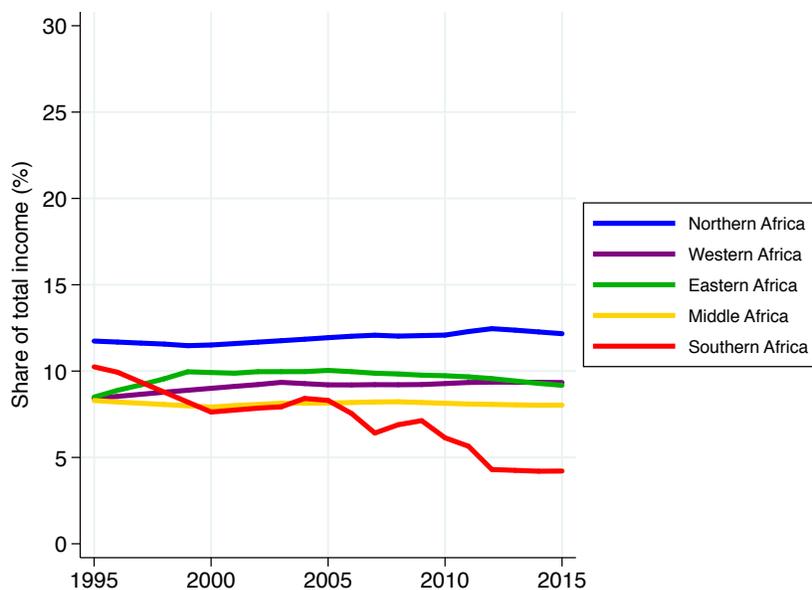
There is no unique African trend in terms of inequality, not even clear regional common trends. On the contrary, income distributions evolved in a wide variety of ways between countries, which underlines the role of national institutions and policies in shaping inequality. Given the important differences in terms of data quality between African countries, and the lack of harmonization of data collection instruments and welfare concepts, and the irregularity of survey availability¹⁷, comparing inequality trends is a perilous exercise. We stress that these results must be interpreted with great caution.

Looking at average regional trends (Figures 18 and 19), it appears that inequality increased dramatically in Southern Africa, while it decreased importantly in Eastern Africa in the years 1990 (before stabilizing in the years 2000), and stagnated in Northern, Middle, and Western Africa, despite small fluctuations.

In Southern Africa, the drastic rise of the income share received by the top 10% was largely, though not exclusively, driven by a strong increase of the income of the very top of the distribution (the share received by the top 1% increased by 6 percentage points between 1995 and 2015). This rise

¹⁷In the following section, countries with an asterisk (*) are countries where data only enables to discuss trends from 1995 to 2005, countries with two asterixes (**) are countries where data only enables to discuss trends after 2005.

Figure 19: Evolution of regional bottom 40% income shares in Africa



Source: Authors' computations.

Note: Weighted averages using total population as weight

occurred at the expense of both the middle and the bottom of the distribution, whose income shares decreased. Indeed, the regional performance over SDG 10.1 for the period between 1995 and 2015 is highly negative (on average, bottom 40% in this region grew by 70 percentage points less than the average) and is the worst across African regions.

This trend was very much driven by South Africa, (by far the most populated country of the region), which experienced a strong increase in income inequality (see Box 1) – despite important poverty reduction rates in the post-Apartheid context.

In the rest of the region: Botswana, Lesotho, Swaziland* and Namibia**, inequality decreased. Top shares fell significantly in the three last countries (by around 6 percentage points), while inequality at the bottom declined at various paces: very importantly in Lesotho, that has one of the highest score over SDG 10.1 (the bottom 40% grew by 88 points more than the average), to a lesser extent, but substantially in Botswana and Swaziland, and much less in Namibia (where the bottom 40% grew by 10 points faster than average).

In Eastern Africa, the regional top income share decreased significantly in the years 1990, while the bottom 40% grew faster than the average from 1995 to 2005. Since the beginning of the years 2000 however, the distribution has remained rather stable: inequality has decreased only within the top and has even grown slightly at the bottom.

Table 1: Sustainable Development Goal 10.1 – Regional Patterns

	Difference between bottom 40% growth and average growth (p.p.)		
	1995-2015	1995-2005	2005-2015
Eastern Africa	47.2	40.2	-4.7
Middle Africa			5.2
Northern Africa	19.2	7.9	8.7
Southern Africa	-69.9	-19.2	-54.3
Western Africa	6.7	4.4	0.7

Source: WID.world (2019)

Note: SDG 10.1 indicators are expressed in percentage points. Regional estimates are derived from the pan-regional distributions, they are not averages of national indicators.

This general trend can be explained by the decrease of inequality in two of the most populated countries of the region, Kenya and Ethiopia. The overall decline was drastic in Ethiopia, where the top 10% share lost 10 percentage point (mostly because of a much slower growth of very top incomes), and the overall gap between the bottom 40% and the average was 48 points; it was more moderate in Kenya (see also Box on inequality trends in Kenya).

Zimbabwe followed a similar trend than Ethiopia. Otherwise inequality rose everywhere in the region, in particular at the bottom of the distribution, to various extents. The increase was very modest in Madagascar (where the top 10% share increased by 3 points). It was more significant at the bottom in Tanzania, Djibouti**, and Uganda (where the bottom 40% grew 15 p. p. less than the average), and even more important in Malawi** (-15 points).

In Rwanda, inequality grew importantly at the bottom (the bottom 40% grew 22 p. p. less than the average), but the top shares also increased moderately. The most worrying evolutions occurred in Mozambique** and Zambia, where the bottom 40% grew 40 and 60 p. p. slower than average, whereas the top 10% shares gained around 7 percentage points.

The decrease of inequality in Northern Africa (where the bottom 40% grew 18 percentage points faster than average from 1995 to 2015), results from the combination of opposite trends. On the one hand, inequality decreased significantly in Algeria (see Box 3) and in Tunisia: in the two countries, the bottom 40% grew 33 and 54 percentage points, respectively, faster than the average, and the top 10% shares lost approximately 5 points. It is worth noting that the decline of top income shares was much more driven by the very top of the distribution in Tunisia. On the other hand, inequality stagnated in Morocco, and increased modestly in Egypt.

The small apparent decrease of inequality in Western Africa (where the overall shared prosperity premium is of 25 points), hides a wide diversity of trajectories. Inequality rose importantly at the bottom in Guinea-Bissau, Ghana and Cote d'Ivoire (the later two are amongst the region's most populous), with bottom 40% income groups growing 20 percentage points more slowly than the average, and even more so in Benin** (30 points more slowly). But the top 10% earners saw their income share rise substantially only in Benin and Guinea-Bissau.

Otherwise, inequality declined. The top 10% share fell by about 4 points in Senegal and Mauritania, but inequality at the bottom was reduced only in Mauritania (the bottom 40% grew 21 percentage points faster than the average). By contrast, Nigeria* experienced a reduction of inequality at the bottom similar than in Mauritania, but top shares stagnated. In Niger, inequality fell substantially both at the top (the top 10% share lost 7 points) and the bottom (the bottom 40% grew 35 percentage points faster than the average).

The inequality reduction was greater in Gambia, Guinea and Mali*, whose bottom 40% grew 60 to 80 percentage points faster than the average, and where top 10% shares lost between 7 and 10 percentage points. The strongest inequality declines were in Burkina Faso and Sierra Leone, where the gap between the bottom 40% and the average are the highest in Africa (respectively 93 and 117 points), and where top income shares decreased by about 12 percentage points. However, in Sierra Leone, this decrease was partly due to a fall in average income (by 13%).

For Central Africa there are few data, over relatively short time span. Despite different patterns, a common feature is that no country displayed a strong trend in inequality, up or down, especially at the top.

For most countries, available data covers the years 2000 and 2010 only. In Cameroon**, Chad** and Congo**, inequality increased at the bottom (average growth was 13 percentage points to 19 p.p faster than for the bottom 40%), but top income shares were relatively constant. Inequality

Table 3: Sustainable Development Goal 10.1 – Summary

	Difference between bottom 40% growth and average growth (p.p.)		
	1995-2015	1995-2005	2005-2015
Algeria	32.5	19.6	9.6
Angola		-26.1	
Botswana	56.4	-9.8	71.8
Cameroon			-19.3
Cote d'Ivoire	-21.2	-22.1	8.2
Egypt	-7.1	-5.5	-0.6
Ethiopia	48.3	75.1	-46.8
Gabon			10.4
Ghana	-24.1	-13.7	-4.5
Kenya	12.6	-8.6	25.7
Madagascar	-0.0	10.4	-8.4
Mali		70.6	
Nigeria		19.2	
South Africa	-74.4	-22.7	-57.8
Zambia	-59.6	-24.7	-20.9

Source: WID.world (2019)

Note: SDG 10.1 indicators are expressed in percentage points. Grey cells correspond to positive shared prosperity premium with decreasing average income.

stagnated in Sao Tome and Principe** and decreased, markedly but not substantially, in Gabon**: bottom 40% grew around 12 p.p faster than the average and top 10% shares fell by about 4 points, but average income decreased in Gabon.

The two countries that can be observed between 1995 and 2005 are Angola* and Central African Republic*. In Angola, inequality increased significantly, at both ends of the distribution. In Central African Republic (CAF), inequality was reduced but average incomes decreased.

Box 1. Income inequality in South Africa since the 1990s

The abolition of the Apartheid regime in Africa put an end to extreme forms of political and social injustices which characterized the regime. The rise of democracy in 1994, was also followed by a strong reduction in the poverty headcount (from 32% in 1993 to 19% in 2014).

In terms of income disparities, however, the past two decades have been marked by a notable increase in inequality. The share owned by the top 10% income earners increased from 45% in 1993 to 65% in 2014, largely driven by the top 1% share, which rose from 10% to 19%. Both the middle and the bottom of the distribution saw their income share decrease, and the bottom of the

distribution fell largely behind: the country's performance over SDG income inequality indicator between 1995 and 2015 is the lowest one in the continent (overall growth was 74 p.p. faster than bottom 40% growth).

Top income earners probably benefited from the good conditions of the world market for agricultural commodities and of the increase in the value of minerals other than gold in the years 2000, as well as from the development of the financial sector. On the other hand, the legacy of the apartheid hindered inequality reduction, even after the end of the regime: for instance former exclusive property rights might have had long-lasting consequences on the distribution of property. This is maybe why only a small Black elite benefited from the end of the Apartheid regime, leading to a decline in the inter-racial gap more than offset by an increase of inequality within the Black population.

South Africa is thus another example of country where the analyse of poverty dynamics and of the distribution of macroeconomic growth should be studied jointly.

Box 2. Income inequality in Kenya since the 1990s

Inequality strongly declined in Kenya between 1992 and 2015: the top 10% income earners captured 66% of national income in 1992 and this share was down to 48% in 2015. This fall was mostly driven by very top income earners, whose share was more than divided by two over the period (from 32% in 1992 to 15% in 2015). The middle of the distribution benefited substantially from it: their share of national income increased from 29% to 42%, while the bottom 40% share almost doubled from 5% to 9%.

Inequality decreased unevenly over time: it fell drastically in the first half of the 1990s, stabilized between 1997 and 2005 (the bottom 40% between 1995 and 2005 grew more slowly than the average), and then fell again significantly between 2005 and 2015 (the bottom 40% grew 26p.p. faster than the average).

These short-run fluctuations seem closely related to Kenya's agitated political history. The period from 1991 to 1993, when top income shares plunged the first time, was one of the country's most explosive periods, with the regime of Daniel Arap Moi stirring ethnic tensions to ensure his reelection in 1992. Riots and expropriations took place in 1992 and 1993, as well as an economic recession with high inflation and almost null growth. Violent riots also followed the 2007 presidential elections, leading to population displacements, infrastructures destructions, and investors' flight. This might have had long-term consequences on Kenyan growth, and thus hindered the growth of top incomes at the turn of the 2010s.

An alternative explanation would hold structural change in the Kenyan economy responsible. Top income shares and the value added share of services fell simultaneously, and top income shares remained stable when the services value added share increased. These evolutions are in

line with the existing literature on structural change and inequality in Africa, according to which the growth of the services sector tends to favor inequality. On the other hand, Simson (2018) shows that in Kenya, as well as in neighbouring Tanzania and Uganda, real wages of civil servants were significantly adjusted downwards in the 1980s and 1990s, to allow public employment (in particular teachers and nurses) and services to expand under strong budgetary constraints. In contrast with the usual elite capture story, the so-called bureaucratic oligarchy of high-pay civil servants conceded to these changes and lost ground at the top of the income distribution.

Box 3. Income inequality in Ivory Coast since the mid-1980

Available data for Côte d'Ivoire shows a reduction in income inequality starting in 1988. Overall, between 1985 and 2014, the top 10 percent of earners' share decreased from 54 percent to 48 percent of national income, largely but not exclusively driven by the top 1 percent's share, which went from 20 percent to 17 percent. Both the bottom and the middle of the distribution were affected.

This decrease is far from linear. Inequality decreased markedly between 1988 and 1993, when the top 10 percent's share went from 54 percent to 45 percent of national income. A possible cause was the spreading effect of the price commodity shock that hit Côte d'Ivoire at the end of the 1980s, eventually affecting everybody and reducing income disparities. Income per capita recovered after 1994 with the devaluation of the franc CFA, a bounce back in commodity prices, and large inflows of foreign aid and private investment. Economic growth then may have given little benefit to the poorest food crops farmers and informal workers, so inequality rose, especially at the bottom, from 1993 to 1998.

In 1999, Robert Guëi overthrew President Konan Bédié in a military coup, the beginning of civil unrest that would end only in 2011. In 2002, with Laurent Gbagbo in power in Abidjan, a surging rebellion in the north led to partition of the country and intervention by French and UN forces. At first, between 1998 and 2002, the political turmoil seemed to fuel inequality by deepening the gap between urban and rural incomes as civil servants' wages and farmers' poverty rates increased. Overall, between 1995 and 2005, the income of bottom 40 percent of earners grew by 22 percentage points less than average income. Between 2002 (before partition) and 2008, the poorest northern areas were more affected by the civil war than the southern ones, if only due to the disruption of cotton production and exports. From 2008 to 2014, inequalities started to decrease but slowly, either because the growth that returned after the 2011 political stabilization was evenly distributed, or because inequalities started increasing again at that point but did not reach their previous levels. Thus, the growth rate of the bottom 40 percent of earners between 2005 and 2015 was slightly higher—8 percentage points—than the average (Cogneau, Hounghbedji, and Mesplé-Somps, 2016; Cogneau, Czajka, and Hounghbedji, 2018).

5 Conclusion

Existing data on economic inequality in Africa is scarce and raises many challenges. We endeavored to address one of the main challenge, namely the strong underestimation of inequality by consumption welfare indicators. Resulting estimates, though they are far from perfect, are at least more comparable with the rest of the world.

The pan-African income distribution built from these estimates appear to be particularly unequal compared to other world regions. Within-countries inequality accounts for a large part of pan-African inequality, and indeed many African countries rank among the most unequal in the world. However interesting regional differences appear: Southern Africa houses the most unequal countries of the continent, while inequality tends to decrease when we move towards the North and the West.

We tried to understand the influence of the productivity and employment shares of agriculture and services, (the two main sectors of the region in terms of employment), on inequality, using a simple model. It appears that the interplay between agricultural productivity, services productivity, and allocation of employment between these two sectors, could partly account for inequality variations across countries. We remain very cautious regarding these findings, as available data does not enable to properly understand their role using econometric estimation techniques, and as these factors must be better disentangled from regional effects in further research.

The evolution of inequality since 1990 is even harder to measure, because data reliability becomes even more questionable as we go back in time, but a wide diversity of trends appear. There has been a very timid decrease of inequality in the pan-African distribution, that is entirely accounted for by the decrease in between-countries inequality. Within-country inequality shows no clear trend on the aggregate level, because there is a very wide variety of trajectories, that cannot even be summed up in clear regional patterns.

Our model relying on services and agriculture is not able to explain these individual evolutions over time, whether we use standardized or original survey estimates. Understanding potential drivers of the evolution of inequality over time in Africa remains an open issue. We stress that further research on the subject needs African countries to cooperate to produce more reliable, transparent, and harmonized distributional data.

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Appendix

Appendix 1. Methodology and data sources

Table A.1: Logistic fit of
income-consumption profiles

Survey	$\hat{\alpha}$	$\hat{\beta}$	Adj. R ²
African countries			
Cote d'Ivoire, 1998	0.85	0.10	0.96
Cote d'Ivoire, 2002	0.81	0.13	0.99
Cote d'Ivoire, 2008	0.84	0.11	0.96
Cote d'Ivoire, 2015	0.85	0.12	0.99
Ghana, 1988	0.87	0.13	0.99
Ghana, 1998	0.81	0.14	0.96
Guinea, 1994	0.85	0.07	0.74
Madagascar, 1993	0.91	0.06	0.94
Uganda, 1992	0.92	0.04	0.72
Other countries			
India, 2005	0.82	0.14	0.99
India, 2011	0.86	0.14	0.98
Thailand, 2000	0.78	0.16	0.97
Thailand, 2001	0.77	0.16	0.98
Thailand, 2002	0.82	0.14	0.98
Thailand, 2004	0.86	0.11	0.96
Thailand, 2006	0.82	0.13	0.95
Thailand, 2007	0.83	0.13	0.95
Thailand, 2009	0.84	0.12	0.96
Thailand, 2011	0.84	0.12	0.93
Thailand, 2013	0.87	0.11	0.91
Thailand, 2015	0.89	0.10	0.93

Source: authors' computations. *Interpretation:* the best logistic fit for the ratio of consumption to income by percentile in 1998 Cote d'Ivoire yields a functional form of $Q(p) = 0.86 + 0.11 \log \frac{p}{1-p}$, with an adjusted R-squared of 97%.

Table A.2: Top 10% and Bottom 50% income shares before and after correction, 2017

Country	Top 10%				Bottom 50%			
	Original Survey	Corrected Survey	Lower Bound	Upper Bound	Original Survey	Corrected Survey	Lower Bound	Upper Bound
Algeria	22.8%	37.3%	33.8%	40.6%	31.3%	20.7%	22.7%	18.9%
Angola	32.3%	48.9%	45.1%	52.3%	21.6%	13.1%	14.6%	11.7%
Benin	37.6%	54.7%	50.9%	58.2%	19.3%	11.5%	12.9%	10.2%
Botswana	41.5%	58.9%	55.1%	62.2%	15.8%	8.7%	9.9%	7.6%
Burkina Faso	29.6%	46.4%	42.5%	50.0%	27.0%	16.5%	18.5%	14.8%
Burundi	31.0%	47.8%	43.9%	51.3%	24.8%	15.1%	16.9%	13.5%
Cabo Verde	37.1%	54.6%	50.7%	58.1%	19.6%	11.3%	12.8%	10.0%
Cameroon	35.0%	51.7%	47.9%	55.1%	19.1%	11.3%	12.7%	10.1%
Central African Republic	46.2%	64.6%	60.8%	68.0%	15.3%	8.0%	9.3%	7.0%
Chad	32.4%	48.9%	45.1%	52.4%	21.3%	13.0%	14.5%	11.6%
Comoros	33.7%	49.9%	46.2%	53.3%	19.8%	12.0%	13.4%	10.7%
Congo	37.9%	55.6%	51.7%	59.1%	18.3%	10.5%	11.9%	9.3%
Cote d'Ivoire	31.9%	48.5%	44.7%	52.0%	22.6%	13.8%	15.4%	12.3%
DR Congo	34.6%	51.8%	47.9%	55.3%	21.0%	12.3%	13.9%	11.0%
Djibouti	32.3%	49.1%	45.3%	52.7%	22.7%	13.8%	15.4%	12.3%
Egypt	27.8%	44.9%	40.9%	48.6%	29.5%	18.3%	20.4%	16.4%
Equatorial Guinea	34.4%	51.4%	47.5%	54.8%	20.6%	12.2%	13.7%	10.9%
Eritrea	31.4%	48.1%	44.3%	51.6%	24.7%	15.0%	16.8%	13.5%
Ethiopia	31.4%	48.1%	44.3%	51.6%	24.7%	15.0%	16.8%	13.5%
Gabon	27.7%	42.8%	39.3%	46.2%	24.1%	15.4%	17.0%	13.9%
Gambia	28.7%	45.2%	41.4%	48.8%	26.2%	16.2%	18.1%	14.6%
Ghana	32.2%	48.6%	44.8%	52.0%	21.0%	12.8%	14.3%	11.5%
Guinea	26.4%	42.1%	38.4%	45.5%	27.3%	17.4%	19.3%	15.8%
Guinea-Bissau	42.0%	59.9%	56.0%	63.3%	18.4%	10.2%	11.6%	8.9%

Kenya	31.6%	48.2%	44.4%	51.7%	23.1%	14.0%	15.6%	12.5%
Lesotho	40.9%	57.6%	53.9%	60.9%	14.8%	8.4%	9.5%	7.4%
Liberia	27.1%	42.6%	39.0%	46.0%	26.2%	16.6%	18.4%	15.0%
Libya	27.5%	43.6%	39.9%	47.2%	27.9%	17.6%	19.5%	15.8%
Madagascar	33.5%	50.3%	46.4%	53.7%	22.2%	13.3%	15.0%	11.9%
Malawi	38.1%	57.8%	53.6%	61.6%	22.4%	12.2%	14.0%	10.7%
Mali	25.7%	40.6%	37.0%	43.9%	27.5%	17.7%	19.6%	16.1%
Mauritania	24.9%	39.9%	36.3%	43.2%	27.7%	18.0%	19.9%	16.4%
Mauritius	29.0%	45.6%	41.7%	49.1%	26.5%	16.5%	18.3%	14.8%
Morocco	31.9%	48.8%	44.9%	52.4%	24.2%	14.6%	16.4%	13.1%
Mozambique	45.5%	64.2%	60.3%	67.6%	17.0%	8.9%	10.3%	7.7%
Namibia	47.2%	64.0%	60.4%	67.1%	12.8%	6.9%	7.9%	6.0%
Niger	27.0%	42.6%	38.9%	46.0%	26.9%	17.1%	18.9%	15.4%
Nigeria	32.7%	49.3%	45.5%	52.8%	21.6%	13.0%	14.6%	11.7%
Rwanda	35.6%	53.4%	49.4%	56.9%	22.1%	12.8%	14.5%	11.3%
Sao Tome and Principe	24.2%	38.7%	35.2%	41.9%	29.0%	19.0%	20.9%	17.3%
Senegal	31.0%	47.2%	43.5%	50.6%	23.3%	14.3%	16.0%	12.9%
Seychelles	33.7%	51.6%	47.5%	55.2%	22.2%	13.0%	14.7%	11.6%
Sierra Leone	26.9%	42.2%	38.6%	45.6%	27.2%	17.3%	19.1%	15.6%
Somalia	33.4%	50.4%	46.6%	53.9%	23.2%	13.9%	15.6%	12.4%
South Africa	50.5%	65.1%	65.1%	65.1%	10.7%	6.3%	5.3%	7.2%
South Sudan	33.2%	49.1%	45.5%	52.5%	18.7%	11.4%	12.7%	10.2%
Sudan	26.7%	42.0%	38.4%	45.3%	26.0%	16.6%	18.4%	15.1%
Swaziland	40.0%	56.9%	53.2%	60.3%	16.8%	9.5%	10.8%	8.4%
Tanzania	31.0%	48.1%	44.2%	51.7%	25.5%	15.4%	17.2%	13.7%
Togo	31.6%	47.6%	43.9%	51.0%	21.1%	12.9%	14.4%	11.6%
Tunisia	25.6%	40.7%	37.1%	44.1%	27.8%	17.9%	19.8%	16.2%
Uganda	34.2%	51.5%	47.6%	55.0%	22.3%	13.1%	14.8%	11.7%

Zambia	44.4%	61.5%	57.8%	64.7%	13.4%	7.3%	8.4%	6.4%
Zanzibar	31.0%	48.1%	44.2%	51.7%	25.5%	15.4%	17.2%	13.7%
Zimbabwe	33.7%	50.7%	46.8%	54.2%	21.7%	12.9%	14.5%	11.4%
Africa	40.3%	54.3%	51.4%	57.0%	13.9%	8.8%	9.6%	8.1%
Eastern Africa	37.3%	53.2%	49.6%	56.5%	19.3%	11.5%	12.9%	10.2%
Middle Africa	47.9%	60.4%	57.5%	63.0%	10.4%	6.5%	7.3%	5.8%
Northern Africa	28.9%	44.6%	41.0%	48.1%	24.2%	15.4%	17.1%	13.8%
Southern Africa	49.9%	64.7%	64.3%	65.1%	11.0%	6.4%	5.6%	7.1%
Subsaharan Africa	42.8%	56.6%	53.8%	59.1%	15.1%	9.3%	10.2%	8.5%
Western Africa	35.5%	50.8%	47.3%	54.0%	18.7%	11.6%	13.0%	10.5%

Table A.3: Data sources

Country	Distributional data	National accounts data	Method
Angola	HH consumption surveys: 1995, 2000, 2008	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Burkina Faso	HH consumption surveys: 1994, 1998, 2003, 2009, 2014	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Burundi	HH consumption surveys: 1992, 1998, 2006, 2013	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Benin	HH consumption surveys: 2003, 2011, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Botswana	HH consumption surveys: 1985, 1993, 2002, 2009, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
DR Congo	HH consumption surveys: 2005, 2008	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Central African Republic	HH consumption surveys: 1992, 2003, 2008	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Congo	HH consumption surveys: 2005, 2011	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Cote d'Ivoire	HH consumption surveys: 1985, 1986, 1987, 1988, 1992, 1993, 1995, 1998, 2002, 2008, 2015; Tax data:2014	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using tax data and national accounts
Cameroon	HH consumption surveys: 1996, 2001, 2007, 2014	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Cabo Verde	HH consumption surveys: 2001, 2007	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Djibouti	HH consumption surveys: 2002, 2012, 2013, 2017	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Algeria	HH consumption surveys: 1988, 1995, 2011	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Egypt	HH consumption surveys: 1990, 1995, 1999, 2004, 2008, 2010, 2012, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2015). World Bank growth rate after 2015.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Ethiopia	HH consumption surveys: 1981, 1995, 1999, 2004, 2005, 2010, 2015	World Bank levels (2017). Previous year growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Gabon	HH consumption surveys: 2005, 2017	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Ghana	HH consumption surveys: 1987, 1988, 1991, 1998, 2005, 2012, 2016	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Gambia	HH consumption surveys: 1992, 1998, 2003, 2010, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Guinea	HH consumption surveys: 1991, 1994, 2002, 2003, 2007, 2012	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Guinea-Bissau	HH consumption surveys: 1991, 1993, 2002, 2010	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Kenya	HH consumption surveys: 1992, 1994, 1997, 2005, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Comoros	HH consumption surveys: 1995, 2004, 2013	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Liberia	HH consumption surveys: 2007, 2014, 2016	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Lesotho	HH consumption surveys: 1986, 1993, 1994, 2002, 2010	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Morocco	HH consumption surveys: 1984, 1990, 1998, 2000, 2006, 2007, 2013	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Madagascar	HH consumption surveys: 1980, 1993, 1997, 1999, 2001, 2005, 2010, 2012	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Mali	HH consumption surveys: 1994, 2001, 2006, 2009	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Mauritania	HH consumption surveys: 1987, 1993, 1995, 2000, 2004, 2008, 2014	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Mauritius	HH consumption surveys: 2006, 2012	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Malawi	HH consumption surveys: 1997, 2004, 2010, 2016	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Mozambique	HH consumption surveys: 1996, 2002, 2007, 2008, 2014	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Namibia	HH consumption surveys: 2003, 2009, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Niger	HH consumption surveys: 1992, 1994, 2005, 2007, 2011, 2014	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Nigeria	HH consumption surveys: 1985, 1992, 2003, 2009	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Rwanda	HH consumption surveys: 1984, 2000, 2005, 2010, 2013, 2016	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Seychelles	HH consumption surveys: 1999, 2006	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Sudan	HH consumption surveys: 2009	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Sierra Leone	HH consumption surveys: 1989, 2003, 2011	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Senegal	HH consumption surveys: 1991, 1994, 2001, 2005, 2011	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
South Sudan	HH consumption surveys: 2009	UN SNA levels (2017) and UN SNA growth rates (1990-2016). Previous year growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Table A.3: Data sources

Sao Tome and Principe	HH consumption surveys: 2000, 2010	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Swaziland	HH consumption surveys: 1994, 2000, 2009	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Chad	HH consumption surveys: 2002, 2003, 2011	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Togo	HH consumption surveys: 2006, 2011, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

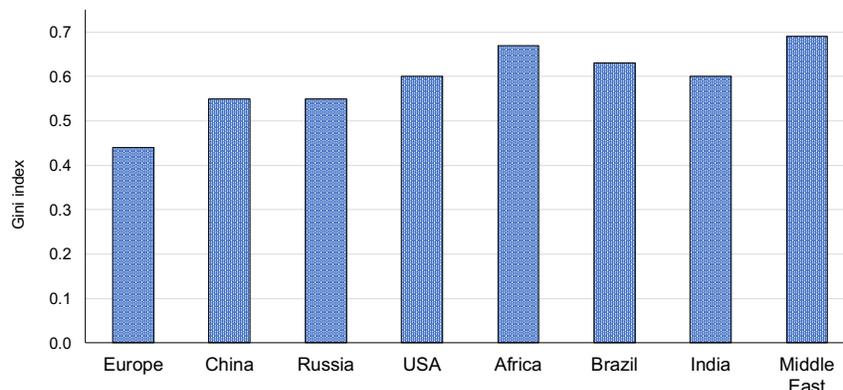
Table A.3: Data sources

Tunisia	HH consumption surveys: 1985, 1990, 1995, 2000, 2005, 2010, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Tanzania	HH consumption surveys: 1991, 2000, 2007, 2011	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Uganda	HH consumption surveys: 1989, 1992, 1996, 1999, 2002, 2005, 2009, 2012, 2016	UN SNA levels (2017) and UN SNA growth rates (1990-2016). Previous year growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
South Africa	HH consumption surveys: 1993, 1996, 2000, 2005, 2008, 2010, 2014; Tax data: 1990-1993, 2002-2012	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using tax data and national accounts

Table A.3: Data sources

Zambia	HH consumption surveys: 1991, 1993, 1996, 1998, 2002, 2004, 2006, 2010, 2015	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts
Zimbabwe	HH consumption surveys: 1991, 1996, 2011	World Bank levels (2017) and UN SNA growth rates (1990-2016). World Bank growth rate after 2016.	Correction of surveys using stylized correction profile (see section 3.2 and 3.3) and national accounts

Figure A.1: Inequality levels across the world (Gini coefficient)



Source: Authors' computations based tax, survey and national accounts data. Distribution of per adult pre-tax income. See WID.world for region or country series and methodological notes.

Appendix 3. Sustainable Development Goal 10.1 Indicators in Africa

The shared prosperity premium (or SDG 10.1 indicator), is the difference between the bottom 40% income growth rate and the average income growth rate.

The average income of the bottom 40% can be written the following way:

$$Income_{Bottom40} = \frac{Bot40 * Average}{0.4} \quad (2)$$

With Bot40 the income share owned by the bottom 40% of the distribution, and Average the national average income. Thus, the shared prosperity premium can be expressed the following way:

$$SDG_{10.1} = \frac{Average_2}{Average_1} * \left(\frac{Bot40_2}{Bot40_1} - 1 \right) \quad (3)$$

$Average_t$ is the average income per capita in period t, $Bot40_t$ is the bottom 40% share in period t, t = 1, 2.

We computed SDG 10.1 indicators from 1995 to 2015 using our standardized inequality estimates and average national income per adult data, for all countries where data is available. These are presented in Table A.4, which is the complete version of Table 1.

Table A.4: Sustainable Development Goal 10.1

	Difference between bottom 40% growth and average growth (p.p.)		
	1995-2015	1995-2005	2005-2015
Algeria	32.5	19.6	9.6
Angola		-26.1	
Benin			-29.8
Botswana	56.4	-9.8	71.8
Burkina Faso	92.8	31.5	32.8
Burundi	-1.2	10.3	-12.9
Cameroon			-19.3
Central African Republic		40.6	
Chad			-12.7
Comoros	5.0	-19.5	37.0
Congo			-13.4
Cote d'Ivoire	-21.2	-22.1	8.2
Djibouti			-6.0
Egypt	-7.1	-5.5	-0.6
Ethiopia	48.3	75.1	-46.8
Gabon			10.4
Gambia	58.3	7.7	46.9
Ghana	-24.1	-13.7	-4.5
Guinea	68.3	30.8	23.5
Guinea-Bissau	-23.8	5.5	-30.4
Kenya	12.6	-8.6	25.7
Lesotho	88.9	86.4	-8.3
Madagascar	-0.0	10.4	-8.4
Malawi			-14.9
Mali		70.6	
Mauritania	21.5	-3.4	25.9
Morocco	2.8	-3.0	5.5
Mozambique			-38.4
Namibia			9.7
Niger	35.0	-8.6	48.9
Nigeria		19.2	
Rwanda	-21.8	-50.4	50.5

Table A.4: Sustainable Development Goal 10.1

	Difference between bottom 40% growth and average growth (p.p.)		
	1995-2015	1995-2005	2005-2015
Sao Tome and Principe			3.1
Senegal	2.0	6.1	-3.5
Sierra Leone	116.8	79.9	17.5
South Africa	-74.4	-22.7	-57.8
Swaziland		36.6	
Tanzania	-8.5	-12.9	8.8
Tunisia	54.2	21.3	20.3
Uganda	-14.9	-14.7	3.1
Zambia	-59.6	-24.7	-20.9
Zimbabwe	31.5	11.2	13.9

Source: WID.world (2019)

Note: SDG 10.1 indicators are expressed in percentage points. Grey cells correspond to positive shared prosperity premium with decreasing average income.

Income share data is not always available for the years 1995, 2005, and 2015, but we present indicators in this time frame to ensure comparability across countries. For 1995-2005 indicators, we kept only countries where data is available in 1995 or before, and in 2005 or after, and interpolated income shares for 1995 and 2005 using closest available years. For 2005-2015 indicators, we kept countries where data is available in 2005 or before, and where there is at least one post-2009 observation. 2005 income shares were therefore interpolated using closest available years, and 2015 shares were either interpolated, when data was available after 2015, or extrapolated from the latest post-2009 year available. For 1995-2015 indicators, we kept only countries where data is available in 1995 or before, and where there is at least one post-2009 observation, and proceeded the same way. Indicators based on actual available years are presented in the Tables A.5, A.6 and A.8.

SDG 10.1 indicators are relatively sensitive to their input data. We analyze below the impact of our data standardization method on this indicator.

This metric depends on two data inputs: average income (or consumption) and income (or consumption) shares estimates. Our standardization procedure (see Section 2) changes both inputs compared to our primary data source, consumption surveys: first, we transform income shares using theoretical underestimation profiles, to account for the facts that consumption distribution is usually smoother than with income, and that surveys underestimate inequality compared to fiscal data. Second, we use WID national average income per adult data instead of surveys' mean

consumption per capita data.

In tables A.5, A.6 and A.8, Column A contains SDG 10.1 indicators constructed using survey data only: consumption shares estimates and average consumption per capita. Column B indicators are built with standardized income share estimates and surveys' average consumption per capita. Column C contains our benchmark indicators, based on standardized income share estimates and WID national income per adult data.

The difference between indicators from Columns B and A is the following:

$$SDG10.1_B - SDG10.1_A = \frac{Average_2^{Survey}}{Average_1^{Survey}} * \left(\frac{Bot40_2^{Standard}}{Bot40_1^{Standard}} - \frac{Bot40_2^{Survey}}{Bot40_1^{Survey}} \right) \quad (4)$$

The gap is driven by the difference in bottom 40% share growth rates: it captures the impact of our inequality estimates standardization procedure. This procedure significantly discounts bottom 40% income share estimates, but the differences between income shares over time ($Bot40_2 - Bot40_1$) are kept relatively constant. This is why, when the bottom 40% income share increases over time, it tends to increase even more in percentage with our standardized estimates. When the bottom 40% share decreases over time, this decline tends to be more pronounced in percentage with standardized estimates. There can be exceptions to this rule, in particular when the share of the very bottom of the distribution increases more than the rest of the bottom 40%, or when a decrease in the share of the top 10% benefits only to the middle of the distribution and has no impact on the bottom 40%. Besides, the impact of the difference between shares' growth rates is exacerbated when average income grows over time, and mitigated when it decreases (Equation 4).

This general trend in income shares explains why SDG 10.1 tends to be higher in Column B when bottom 40% average income grows faster than average income (green cells): bottom 40% average income grows even faster with standardized estimates because shares grow increase in percentage. Conversely, when SDG 10.1 is negative, either because bottom 40% income increases more slowly than average income, either because it decreases more quickly, or falls when average income increases, the gap tends to be wider with standardized estimates. Indeed, with standardized estimates, bottom 40% average income either decreases faster, either increases more slowly. This is why most negative SDG 10.1 (red cells) are even smaller in Column B. When SDG 10.1 is positive but average income is decreasing (grey cells), the gap between the two income growth rates also tends to be wider in Column B, because bottom 40% income grows more quickly and decreases more slowly with standardized estimates¹⁸.

¹⁸Income shares estimates for Cote d'Ivoire and South Africa are issued from Czajka (2017) and Alvaredo and Atkinso (2010), respectively, because available data in these two countries enable more refined corrections. Therefore, they do not result from the same standardization methodology, but the issues addressed and the methodological principles are very close. They are available on WID.world.

The difference between SDG 10.1 from Columns C and B is the following:

$$SDG10.1_C - SDG10.1_B = \left(\frac{Bot40_2^{Standard}}{Bot40_1^{Standard}} - 1 \right) * \left(\frac{Average_2^{WID}}{Average_1^{WID}} - \frac{Average_2^{Survey}}{Average_1^{Survey}} \right) \quad (5)$$

It is driven by the difference in average growth rates. These growth rates sometimes differ substantially. This divergence can be imputed to the different welfare concept they are based on (consumption or income), the different data sources they come from (surveys or national accounts data, measurement variability over time being a well-known issue for surveys), or the scale chosen (per capita or per adult). The impact of this difference on the gap between SDG 10.1 indicators is exacerbated when the bottom 40% income share increases by more than 100%, it is mitigated if the bottom 40% share grows by less than 100%, and it is reversed if the bottom 40% share declines over time (Equation 5).

The difference between SDG 10.1 from Columns C and A is the following:

$$SDG10.1_C - SDG10.1_A = \frac{Average_2^{WID}}{Average_1^{WID}} * \left(\frac{Bot40_2^{Standard}}{Bot40_1^{Standard}} - 1 \right) - \frac{Average_2^{Survey}}{Average_1^{Survey}} * \left(\frac{Bot40_2^{Survey}}{Bot40_1^{Survey}} - 1 \right) \quad (6)$$

It can be decomposed as the combination of the two previous effects: the impact of the standardization of income shares estimates (the share effect), captured by the difference between Columns B and A, and the impact of the use of national income per adult data as average (the average effect), captured by the gap between Columns C and B.

The respective importance of share and average effects vary importantly across countries: none of them strongly dominates the other, and there is no apparent correlation with the size or direction of the gap between indicators C and A. The share effect accounts for more than 50% of the total difference in a small majority of cases (around 60% of all observations). However, the gap between $SDG10.1_C$ and $SDG10.1_B$ (the mean effect), reaches much higher sizes, in terms of percentage point: its extreme values are 70 and -115 percentage points, versus 50 and -31 for the gap between $SDG10.1_B$ and $SDG10.1_A$.

The income of the bottom of the distribution is arguably better captured by survey aggregates than by national accounts, though the question remains open. It would therefore be interesting to compare our benchmark indicator to an indicator based on survey aggregates for bottom 40% income, and average income from national accounts. Unfortunately, it is not possible to build such an indicator at the per adult scale for the time being, because the most existing welfare survey data for Africa is expressed in per capita terms, and demographic information is not detailed enough to

convert it in per adult terms.

Though SDG 10.1 appears to be relatively sensitive to data inputs, it should be noted that this variation affects only the size of the indicator and not its direction. Thus, the answer to the simple question, “Did the income of the bottom 40% grew faster than the average?” remains the same. Besides, important size differences concern only a handful of countries, and seem to improve over time.

However, the sensitivity of SDG 10.1 to data inputs underlines the importance of transparency about data and methodology used for inequality measurement. It most of all stresses the need to reach an agreement on welfare concepts and methodological tools used to measure economic inequality, and the urge to access welfare data of better quality in emerging countries.

In the following tables are presented:

- Column (A): Inequality estimates and average incomes from consumption surveys
- Column(B): Inequality estimates from WID.world, average incomes from consumption surveys
- Column(C): Inequality estimates and average incomes from WID.world
- Column(C)-(A): Difference between indicators column C and column A
- Share effect (in %) = (column B - column A)/(column C - column A)
- Average effect (in %) = (column C - column B)/(column C - column A)

Table A.5: SDG 10.1 from surveys and WID.world data inputs (2000s-2010s)

	Difference between bottom 40% growth and average growth (p.p)					
	(A)Survey	(B)WID-Survey	(C)WID	(C)-(A)	Share Effect	Average Effect
Angola(00-08)	32.3	39.9	72.0	39.7	19	81
Benin(03-15)	-29.4	-30.5	-32.7	-3.3	34	66
Botswana(02-09)	27.7	30.0	30.8	3.1	75	25
Burkina Faso(03-14)	34.6	39.5	42.0	7.4	66	34
Burundi(06-13)	-16.3	-18.0	-16.2	0.1	-1643	1743
CAF(03-08)	-36.1	-47.4	-41.8	-5.7	199	-99
Cabo Verde(01-07)	18.9	25.6	26.8	7.9	86	14
Cameroon(01-14)	-24.1	-23.7	-22.1	2.0	19	81
Chad(03-11)	-20.7	-21.6	-22.3	-1.6	59	41
Comoros(04-13)	14.5	30.0	44.8	30.2	51	49
Congo(05-11)	-10.8	-11.8	-11.8	-1.0	100	0
Cote d'Ivoire(02-15)	-3.9	-3.0	-3.7	0.2	411	-311
Djibouti(02-13)	-13.2	-15.1	-16.6	-3.4	56	44

Table A.5: SDG 10.1 from surveys and WID.world data inputs (2000s-2010s)

	Difference between bottom 40% growth and average growth (p.p)				Share Effect	Average Effect
	(A)Survey	(B)WID-Survey	(C)WID	(C)-(A)		
Egypt(04-15)	1.2	0.1	0.1	-1.1	100	0
Ethiopia(04-15)	-29.0	-32.8	-49.2	-20.2	19	81
Gabon(05-17)	8.9	19.1	12.2	3.3	307	-207
Gambia(03-15)	63.5	80.8	54.2	-9.4	-184	284
Ghana(05-12)	0.1	1.4	1.3	1.2	107	-7
Guinea(02-12)	36.0	50.4	40.1	4.0	355	-255
Guinea-Bissau(02-10)	-32.8	-40.6	-41.1	-8.3	95	5
Kenya(05-15)	16.9	23.4	25.7	8.8	73	27
Lesotho(02-10)	-9.1	-11.3	-13.2	-4.1	55	45
Liberia(07-14)	13.5	15.5	13.0	-0.4	-464	564
Madagascar(01-12)	14.2	17.2	18.4	4.2	72	28
Malawi(04-10)	-16.4	-19.6	-22.6	-6.2	50	50
Mali(01-09)	22.3	28.6	38.8	16.4	38	62
Mauritania(00-14)	23.7	29.9	29.2	5.5	112	-12
Mauritius(06-12)	-0.6	-0.7	-0.8	-0.2	69	31
Morocco(00-13)	6.6	7.7	7.0	0.4	250	-150
Mozambique(02-14)	-31.5	-39.1	-40.8	-9.4	81	19
Namibia(03-15)	3.9	17.5	15.0	11.1	122	-22
Niger(05-14)	39.3	60.4	48.8	9.5	222	-122
Nigeria(03-09)	-6.4	-9.4	-10.8	-4.4	68	32
Rwanda(00-13)	16.0	19.9	23.3	7.3	54	46
Sao Tome and P.(00-10)	4.1	4.6	6.6	2.5	20	80
Senegal(01-11)	-2.7	2.8	2.5	5.3	105	-5
Sierra Leone(03-11)	17.5	23.6	23.5	6.0	100	-0
South Africa(00-14)	-32.8	-85.9	-58.5	-25.8	206	-106
Swaziland(00-09)	-4.5	0.7	0.7	5.2	101	-1
Tanzania(00-11)	5.2	-2.3	-1.5	-6.7	112	-12
Togo(06-15)	-8.4	-6.5	-7.5	0.9	205	-105
Tunisia(00-10)	16.3	24.0	23.8	7.5	103	-3
Uganda(02-16)	4.9	10.6	10.3	5.4	106	-6
Zambia(02-15)	-50.2	-54.9	-76.4	-26.2	18	82

Table A.5: SDG 10.1 from surveys and WID.world data inputs (2000s-2010s)

Difference between bottom 40% growth and average growth (p.p)					
(A)Survey	(B)WID-Survey	(C)WID	(C)-(A)	Share Effect	Average Effect

Source: WID.world (2019)

Note: SDG 10.1 are expressed in percentage points. Grey cells correspond to positive shared prosperity premium with decreasing national average income per capita.

Table A.6: SDG 10.1 from surveys and WID.world data inputs (1990s-2000s)

	Difference between bottom 40% growth and average growth (p.p.)					Average Effect
	(A)Survey	(B)WID-Survey	(C)WID	(C)-(A)	Share Effect	
Botswana(93-02)	-33.8	-32.5	-20.9	12.9	10	90
Burkina Faso(94-03)	15.3	27.5	21.0	5.8	212	-112
Burundi(92-06)	4.3	1.5	0.9	-3.4	83	17
CAF(92-03)	184.9	235.2	120.2	-64.8	-78	178
Cameroon(96-01)	4.7	7.9	6.2	1.5	214	-114
Cote d'Ivoire(92-02)	0.6	-2.6	-2.6	-3.2	102	-2
Egypt(90-04)	2.1	0.7	0.9	-1.2	113	-13
Ethiopia(95-04)	47.0	72.8	67.7	20.8	125	-25
Gambia(98-03)	7.5	6.4	3.7	-3.8	29	71
Ghana(91-05)	-24.0	-26.8	-18.5	5.5	-53	153
Guinea(91-02)	102.7	83.8	35.6	-67.1	28	72
Guinea-Bissau(91-02)	72.3	86.7	143.5	71.2	20	80
Kenya(92-05)	22.6	31.9	50.6	28.0	33	67
Lesotho(94-02)	42.9	50.8	112.0	69.1	11	89
Madagascar(93-01)	-11.8	-9.2	-9.8	2.0	131	-31
Malawi(97-04)	37.6	72.2	142.4	104.9	33	67
Mali(94-01)	58.0	83.2	91.7	33.7	75	25
Mauritania(93-00)	27.7	55.4	43.3	15.6	177	-77
Morocco(90-00)	-1.2	-1.6	-1.9	-0.7	67	33
Mozambique(96-02)	31.2	35.3	42.4	11.2	36	64
Niger(92-05)	-22.0	-29.1	-25.6	-3.6	195	-95
Nigeria(92-03)	24.7	24.1	29.9	5.2	-11	111
Senegal(91-01)	77.5	93.2	74.7	-2.7	-573	673
Seychelles(99-06)	-0.7	-0.7	-0.6	0.1	60	40
South Africa(93-00)	4.6	-27.2	-31.7	-36.3	88	12
Swaziland(94-00)	93.4	108.6	48.3	-45.1	-34	134
Tanzania(91-00)	-3.5	-4.9	-7.4	-3.8	36	64
Tunisia(90-00)	-3.3	-4.0	-3.8	-0.5	135	-35
Uganda(92-02)	-7.4	-11.3	-16.3	-8.9	43	57
Zambia(91-02)	184.8	189.8	252.3	67.5	7	93

Source: WID.world (2019)

Note: SDG 10.1 are expressed in percentage points. Grey cells correspond to positive shared prosperity premium with decreasing national average income per capita.

Table A.8: SDG 10.1 from surveys and WID.world data inputs (1990s-2010s)

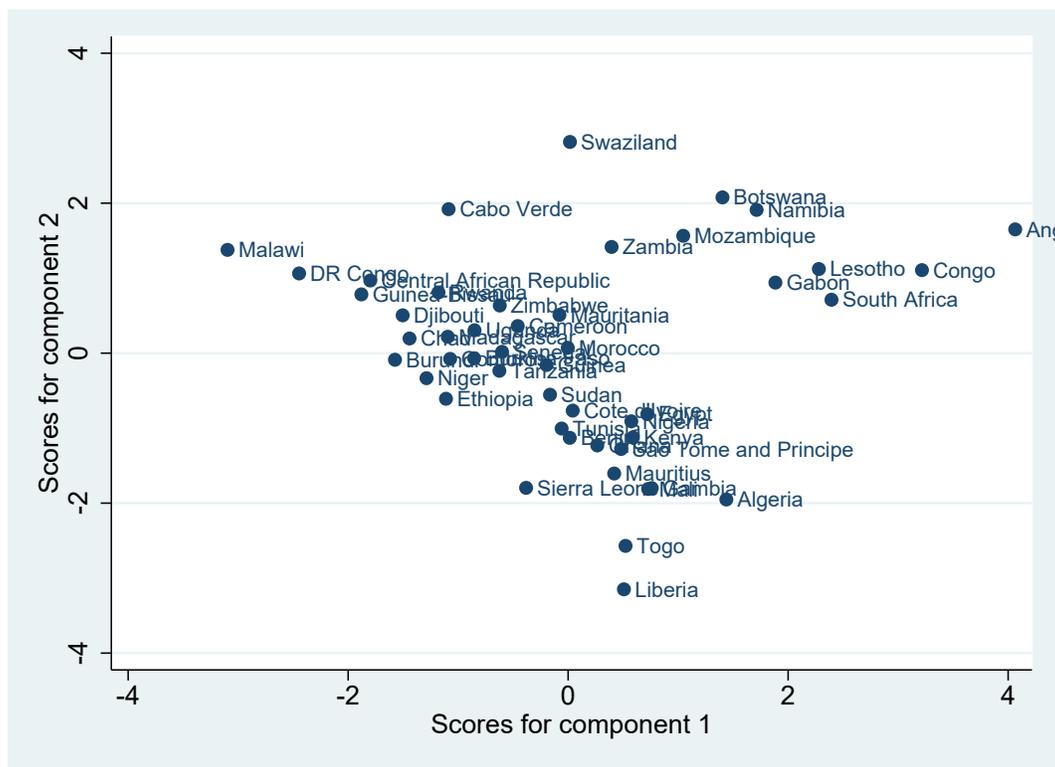
	Difference between bottom 40% growth and average growth (p.p.)					
	(A)Survey	(B)WID-Survey	(C)WID	(C)-(A)	Share Effect (%)	Average Effect (%)
Algeria(95-11)	31.7	37.8	31.4	-0.3	-2444	2544
Botswana(93-09)	-8.2	-3.7	-2.5	5.8	78	22
Burkina Faso(94-14)	84.0	112.5	91.5	7.5	381	-281
Burundi(92-13)	-13.9	-18.6	-10.1	3.8	-122	222
CAF(92-08)	99.3	97.3	43.8	-55.5	4	96
Cameroon(96-14)	-29.5	-25.8	-18.7	10.8	35	65
Cote d'Ivoire(92-15)	-3.2	-5.3	-6.3	-3.1	66	34
Egypt(90-15)	3.8	1.0	1.1	-2.7	107	-7
Ethiopia(95-15)	14.4	34.6	48.3	33.9	60	40
Gambia(98-15)	125.1	153.4	60.0	-65.2	-43	143
Ghana(91-12)	-34.0	-36.2	-23.5	10.5	-21	121
Guinea(91-12)	272.9	292.1	98.6	-174.3	-11	111
Guinea-Bissau(91-10)	29.5	27.8	46.4	16.9	-10	110
Kenya(92-15)	36.9	53.9	94.1	57.2	30	70
Lesotho(94-10)	38.1	43.6	112.2	74.2	7	93
Madagascar(93-12)	2.7	7.8	8.9	6.1	83	17
Malawi(97-10)	25.1	51.7	117.9	92.8	29	71
Mali(94-09)	115.4	164.6	245.8	130.3	38	62
Mauritania(93-14)	71.2	123.8	94.7	23.6	223	-123
Morocco(90-13)	3.8	3.9	4.1	0.3	33	67
Mozambique(96-14)	5.9	0.1	0.1	-5.8	100	-0
Niger(92-14)	2.5	6.6	4.7	2.3	183	-83
Nigeria(92-09)	18.3	14.0	19.9	1.6	-278	378
Senegal(91-11)	83.9	114.0	82.7	-1.2	-2504	2604
South Africa(93-14)	-21.4	-97.4	-77.2	-55.7	136	-36
Swaziland(94-09)	83.8	117.4	49.2	-34.6	-97	197
Tanzania(91-11)	-4.2	-12.2	-12.0	-7.8	103	-3
Tunisia(90-10)	15.0	23.0	21.7	6.7	119	-19
Uganda(92-16)	-5.5	-5.7	-8.0	-2.4	5	95
Zambia(91-15)	66.7	56.7	104.9	38.1	-26	126

Source: WID.world (2019)

Note: SDG 10.1 are expressed in percentage points. Grey cells correspond to positive shared prosperity premium with decreasing national average income per capita.

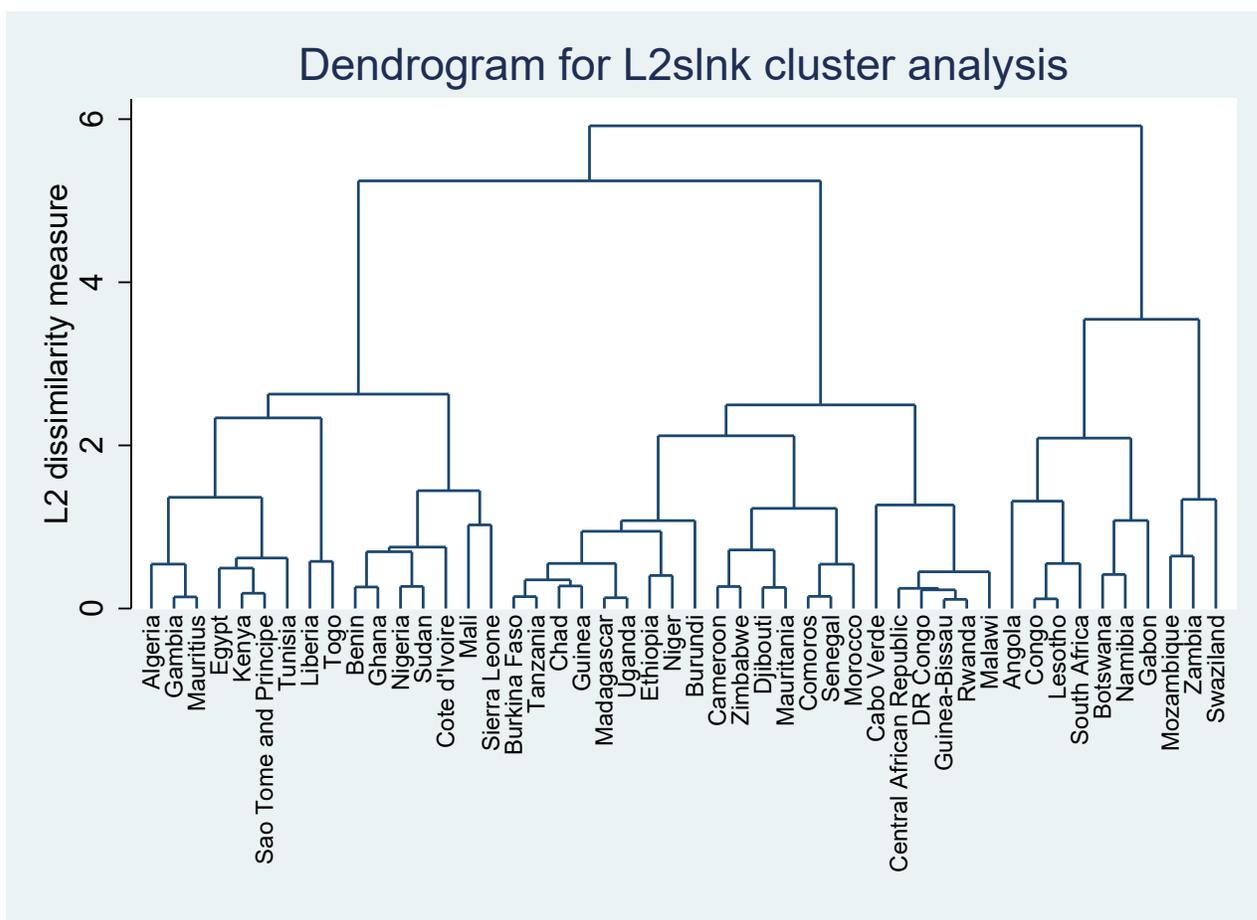
Appendix 4. Typology of countries based on structural features

Figure A.2: Principal component analysis



Notes: The graph shows the mapping of countries according to the first and second components of a principal component analysis based on the correlation matrix of the following variables: top 10% income share, agricultural employment share, productivity of labor in agriculture relative to the non-agricultural sector, unemployment rate, an indicator of informality(residual of a regression of the share of self-employed on the share of agriculture in employment, its square, and the unemployment rate), and the share of mining rents on GDP. This mapping was used to classify countries in three groups (see dendrogram below). The first group of countries lies in the right upper quadrant of the graph, the second group in the left upper quadrant, and the third group at the bottom.

Figure A.3: Classification of countries



Notes: The dendrogram shows how countries are grouped together according to the two first components of the principal component analysis.