**Measuring the Labor Income Share of Developing Countries**

Learning from Social Accounting Matrices

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**Abstract**:

This paper is the first to address the challenges of measuring the labor income share of developing countries. The poor availability and reliability of national account data as well as the fact that self-employed, whose labor income is hard to capture, account for a major share of the workforce and often work in the informal sector render its computation difficult. I consult social accounting matrices as additional source of information to construct a labor share data set that is backed up with micro-economic evidence. First descriptive results show a significant downward trend in labor shares of developing countries since the early 1990s.

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# Introduction

In the recent past, inequality has increased within most countries, not only in high income regions but also across the developing world (Atkinson and Bourguignon 2015; Goldberg 2007; Niño-Zarazúa et al. 2014; UNDP 2013 and others). Despite this worrying and serious development, we know far too little about inequality, its magnitude, causes and consequences. For a long time, the distribution of income between individuals and even more so between the factors of production has been left largely unattended by the economic discipline. But studies like the recent Oxfam publication predicting that 'by 2016, the top 1% will have more than 50% of total global wealth' (Hardoon 2015: 22) evermore enhance the interest of economists. Not least because of the widely known book by Thomas Piketty (2014), also the distribution question of capital and labor has been put back on the international agenda.

The labor income share reflects how much of national value added accrues to labor (as opposed to capital). It is a highly informative macroeconomic variable to explore when analyzing the roots of inequality. As has been emphasized by Atkinson (2009), shifting the focus from the personal to the factor level can deliver a more integrated understanding of the determinants of inequality, such as imbalances between different sources of income. By decomposing inequality into factor shares and their concentration, the underlying causes of income disparities can be more easily grasped. For example, profits, rents and other income from capital are much more concentrated than labor and inclined to go to rich households, as revealed by Piketty (2014) and others. A declining labor share is thus likely to indicate an increase in inequality among individuals (ILO 2014c). This illustrates that the personal income distribution can be directly determined from factor income shares once the owners of capital, land and labor are known (Ray 1998).

Dynamics in the factor income distribution are of particular relevance for developing countries, especially in their effort to fight poverty. Regressive redistribution of factors and their remuneration will be felt strong in these countries due to weak social safety nets and limited capital access of the poor. The main asset of the poor certainly is labor, usually in form of agricultural self-employment (Fields 2014). As such, the labor share can serve as indicator in designing policies for social protection and tax systems as these usually target at the factor income distribution (minimum wage policies, tax concessions for investments etc.).

The measurement of the labor share, however, is notoriously difficult for the low and lower middle income countries. Most studies rely on the relation of compensation of employees to GDP from national account statistics when measuring the labor share. A key problem of this simple definition is the fact that compensation of employees does not include the labor income of the self-employed, which account for the major fraction of the labor force in developing countries (Gollin 2002). An additional difficulty arises from the fact that most of the self-employed in developing countries are located in the informal sector. Eventually, and most importantly, developing countries give reason for concern about the scope, detail and quality of their national accounts (United Nations 2012). The adjustment of the labor share hence requires more prudent handling in the case of less developed countries. Furthermore, the fact that the economic structure of developing countries fundamentally differs from the ones of high income economies makes separate assumptions and estimates indispensable. For example, self-employed in OECD countries are more likely to have consciously decided to enter self-employment while it may be a business out of necessity for workers in the developing world.

To date, there is no compiled and adjusted panel data set available that addresses developing countries and their peculiarities when measuring the labor share. As a consequence, research findings on the development of the labor share in lower income countries remain incomplete which is a clear gap in the economic literature. For instance, Piketty's (2014) highly illuminating results remain limited to OECD countries and a few major emerging economies since the lower income countries do not dispose of a (well-documented) tax base that could be used as data source.

By exploring the challenges associated with the measurement of the labor share in developing countries, identifying social accounting matrices (SAMs) as favorable solution approach and drawing on them to construct a data set of labor shares of poor countries, this paper intends to take a first step towards closing this gap. SAMs are micro-funded representations of an economy that provide detailed data on all the economic transactions that take place within a country. Insights are used to counter-check the reliability of macro-economic data and to formulate assumptions required for measuring the labor share. By this means, this paper provides the first macro-level labor share data set for developing countries that is based on and backed up by micro-economic evidence.

The final data set covers about 100 developing countries from 1990-2011. First descriptive statistics show that labor's relative income in developing countries has been declining since the early 1990s by about 10 percentage points, indicating that global labor income has been lagging behind overall productivity increases. This effect holds true for most of the developing regions and seems to be independent of a country's development stage. The hope that globalization will not only lead to economic growth in developing countries but also reduce worldwide and within-country inequality through the promotion of the factor labor hence seems not to materialize.

The paper is organized as follows: Section 2 presents the concept of the labor share and elaborates on its measurement from national accounts. Existing data sets are reviewed in section 3. Section 4 discusses the challenges associated with measuring the labor share of developing countries, whereupon section 5 introduces SAMs as solution approach that serve as information source and robustness check. Insights are applied to construct and validate the labor share in the following step (section 6). Section 7 presents some properties of the data set and section 8 concludes.

# The Labor Share: Concepts and Measurement

The labor share reflects how much of national income is earned by labor and hence measures the factor income distribution of a country. Assuming that value added, or production output, is given by Y = f(K, L), where K is capital (including land capital) and L labor used in production, the income distribution between production factors is given by:

where w is wage, i the interest rate, r rent and P the price level. The labor share LS then can be expressed as:

The standard approach to describe the relationship between factors of production and their production output continues to be the Cobb-Douglas production function (Cobb and Douglas 1928). Due to its assumption of constant output elasticities and factor remuneration according to productivity, the Cobb-Douglas production technology predicts factor shares to be consistent over time. Changes in the labor share may only originate from shifts to sectors that use one of the production factors with relative intensity, non-neutral technological progress, or a deviation of a factor's remuneration from its marginal product caused by labor adjustment costs or bargaining between capital and labor (Bentolila and Saint-Paul 2003). Early empirical estimations of the factor income distribution using the Cobb-Douglas production function found the labor share to range between 0.6 and 0.75 in the United States and other high income countries, giving rise to the standard value of two thirds for labor's share in production output (Douglas 1967, 1976). The same studies approved the long-term stability of factor shares, which also became one of the 'stylized facts' of economic growth put forward prominently by Kaldor (1957).

The labor share can be computed from national accounts statistics. The empirical literature usually starts out from the relation of compensation of employees, taken from the primary distribution of income accounts, to total value added produced in the respective country (GDP):

Data is provided by the United Nations System of National Accounts (UN SNA) and is accessible through the National Accounts Official Country data. This simple measurement, however, tells only half the story and is often referred to as the '*naive labor share*'. As pointed out by Krueger (1998) and Gollin (2002), compensation of employees merely covers wage earners in the corporate sector and ignores self-employment[[1]](#footnote-1). The challenge with self-employed income is that it is composed of income from labor as well as from capital so that its labor component needs to be filtered out in a first step. Labor income, however, is not easy to identify, especially at the national level. If it is not corrected for self-employment, income of the self-employed would be mistakenly treated as only consisting of non-labor income and would be added to the denominator of the equation but not to the numerator, resulting in a downward bias of the labor share. Furthermore, in a dynamic perspective, ceteris paribus shifts in the composition of employment would automatically change the labor share. For example, when formerly self-employed enter wage employment − in developing countries, this typically is the movement away from subsistence agriculture − their labor income suddenly appears in employee compensation statistics, raising the labor share, even though labor income has effectively not changed (or only very little). It is therefore essential to adjust this measure such that it does not only reflect the share of national income that is earned by employees but the entire share that accrues to labor input, regardless of how it was earned. Gollin (2002) presented three possible approaches in this regard. His article soon became the standard reference.

Gollin's first two adjustments make use of the item 'mixed income' listed in the UN SNA. Mixed income refers to the remuneration of the self-employed and − as the term already suggests − includes income from labor and capital (UN 2008). By using this item and filtering out its labor income component, which is then added to employee compensation, a meaningful measure of the labor share can be obtained.

In his *first adjustment*, mixed income of the self-employed is completely added to compensation of employees, assuming income of the self-employed to be only composed of labor income:

As this procedure ignores income from other factors of production than labor, it is likely to overestimate the labor share.

The *second adjustment* assumes mixed income to consist of the same mix of labor and non-labor income as the rest of the economy:

This approach is more straightforward but disregards that capital and labor shares might vary substantially across sectors and with the size and structure of businesses.

Gollin's *third adjustment* requires only data on the employment structure of a country, which is for example available in ILO's Key Indicators of the Labor Market database. Relying on the assumption that self-employed earn the same labor income as employees, it imputes the average wage bill of employees to the self-employed. Only income of the self-employed that exceeds the mean wage sum is counted as income from capital:

This adjustment does not take into consideration that self-employed and wage earners might work in different sectors, realizing different labor productivities. If self-employment mainly occurs in subsistence farming and other low-productive activities, as it is typically the case in developing countries, this equation systematically overestimates the labor share. On the other hand, it underestimates the labor share in countries where the self-employed belong to the high income earners of the economy. To account for such systematic differences, some studies (Arpaia et al. 2009; OECD 2012) impute the income of the self-employed on the basis of the average wage of employees in each sector, using EU KLEMS data. While the naive labor share reflects the share of the total wage bill of employees in GDP, the labor share computed using this third adjustment relates the wage per employee to the value added per employed person. It follows that the adjusted labor share rises (falls) if the wage per employee grows faster (slower) than the productivity per employed person. It is hence closely linked to the unit labor costs which relate average wages to overall performance, that is to productivity (Marterbauer 2003).

When applying these adjustments, Gollin (2002) finds labor shares to be more or less constant across time and space and therefore suggests to adhere to models using a Cobb-Douglas production technology. The size of the labor share is shown to range between 0.6 and 0.85. His results are, however, based on a small sample of 31 high and low income countries observed at only one point in time and are heavily challenged by other recent studies.

# Data Review

Aside from Gollin (2002), various international organizations as well as researchers have taken up the analysis of trends in labor shares. The majority of the empirical literature (Arpaia et al. 2009; Bentolila 2003; Blanchard 1997; Daudey and García-Peñalosa 2010; EC 2007; Ellis and Smith 2007; Guscina 2006; Hutchinson and Persyn 2012; ILO 2011; IMF 2007; Jaumotte and Tytell 2007; OECD 2012; Richardson 1998; Slaughter 2001; and others) is restricted to OECD countries, where data quality as well as coverage is high and data on employment structure and mixed income available. They rely on the naive labor share or, additionally, Gollin's third adjustment, as it is for example provided in the European Commission's AMECO database.

Piketty (2014) adopts an entirely different approach and uses tax data to study the distribution of top incomes and capital gains.[[2]](#footnote-2) His study is restricted to around 30 countries, mainly from the Western Hemisphere, that provide income tax data. When he splits mixed income between capital and labor income, he relies on Gollin's second adjustment.

There are some more broader studies (e.g. Bernanke and Gürkaynak 2001; Diwan 2001; Guerriero and Sen 2012; Harrison 2005; Jayadev 2007; Karabarbounis and Neiman 2013; Rodriguez and Jayadev 2010) that conduct worldwide analyses including a number of developing countries. Like the recent study of Karabarbounis and Neiman (2013), they mainly base their analysis on unadjusted labor shares. In fact, about two thirds of low and middle income countries report the item 'compensation of employees' to UN SNA on a regular basis since 1990 but a systematic recording of unincorporated businesses remains missing. Just a handful of countries report the necessary data on the number of self-employed or their income with the result that applying one of Gollin's adjustments (primarily the first or third) comes with the consequence that only a few developing countries remain in the sample. For example, when Rodriguez and Jayadev (2010) adjust their broad labor share dataset of 135 countries for mixed income, their sample size sharply reduces to 59 (mainly high income) countries. Some other studies (Decreuse and Mareek 2013; Ortega and Rodriguez 2001) use the UNIDO dataset INDSTAT3 as an alternative to the UN SNA data but it is even more restrictive as it is limited to the corporate manufacturing sector.

With adjusted labor share data for 127 low, middle and high income countries for at least 20 years, the recently published Penn World Table (PWT) version 8.0 is the broadest analysis of labor share trends. By using data on total agricultural value added (taken from the Socio-Economic Accounts of the World Input-Output Database) as a proxy for mixed income, Inklaar and Timmer (2013) are able to increase the number of sample countries by about 60. Time and country coverage is further extended by interpolating and keeping labor shares constant over time. They develop a 'best estimate' labor share that chooses the most appropriate from the three Gollin adjustments in light of the country-specific data situation.[[3]](#footnote-3) Their approach is primarily mechanical meaning that it is based on a country's data availability and plausibility rather than on economic conditions.[[4]](#footnote-4)

In contrast to Gollin's results, all OECD studies find the labor share to be decreasing in most of the high income countries over time, regardless of how labor's share in national income is measured. The global studies confirm this decline and further provide evidence of a significant negative trend for the developing world. The labor share is also found to be less than the ubiquitous 'two thirds', both in rich and poor countries.[[5]](#footnote-5)

To my knowledge, there is no study explicitly exploring the development of the labor share in low and middle income countries. This is a clear gap in empirical research as the special economic structures and endowments of developing countries require a distinct model as well as a separate empirical investigation. Instead, the developing world is exclusively covered in global studies, where the same assumptions are applied for industrialized as well as low income economies. It remains some risk that this approach results in incorrect labor share data. Self-employment systematically differs between the two economic worlds: While it may mean business management and entrepreneurial spirit in the Western hemisphere, it is likely to refer to subsistence farming and precarious employment in developing nations. Similarly, cross-country studies omit verifying the assumptions on which their measurement relies with national or regional data, which might be another source of trouble. My dataset improves upon this literature by being the first to consider region-specific peculiarities as well as differences owed to a country's development status when constructing the labor share. It also considers the specific data situation in these countries. To do so, this study investigates self-employment sectors and their typical characteristics in low and lower middle income countries consulting national social-accounting matrices.

# Measurement Challenges

Data review has shown that constructing a broad data set on the labor share of developing countries is strongly hampered by the limited availability of national account data in these countries. Data on self-employed income is lacking for the majority of lower income economies and even basic figures such as the number of self-employed are scarce. This is not surprising in view of the fact that the most prevailing forms of self-employment in developing countries are micro and small enterprises and subsistence farmers whose economic activities are difficult to capture. The naive labor share hence cannot easily be adjusted for self-employment as suggested by Gollin (2002). This is only possible against the backdrop of losing the bulk of observations, most probably the least developed countries which do not have the institutional capacities to collect the data.

Data constraints therefore require to select proxy variables for either self-employed income or their share in the workforce when compiling a comprehensive data set. This, in turn, involves making assumptions about the composition of self-employed, their characteristics or value added. To impute the labor income of the self-employed and decide for one of Gollin's adjustments, it is additionally needed to formulate assumptions concerning the self-employed factor intensities and productivities. The PWT database uses total value added in agriculture as proxy for self-employed labor income, assuming that most of self-employed income stems from agricultural production with labor being by far the most important input factor (Inklaar and Timmer 2013). This proxy is plausible, yet it disregards capital, especially land, as agricultural factor and leaves aside labor income from other forms of self-employment. One should also bear in mind that this variable is not able to reflect industrialization of agriculture, meaning a constant or increasing agricultural value added accompanied by a decreasing number of farmers.

Choosing a proxy in the context of developing countries takes place against the background that − in contrast to advanced economies − an average of about two thirds and up to 90 per cent of the working population is self-employed, with most of them belonging to the informal sector (ILO 2014b). Self-employment hence is the rule in lower income countries and systematically differs from the one of developed economies: It primarily describes workers who are forced into self-employment as they do not find an employment on the regular job market, whereas self-employment in OECD countries rather refers to (formerly employed) people who wish to set up their own business and become their own boss (Field 2013; World Bank 2013). It is essential to differentiate between these two main circumstances and determine whether it is a conscious decision or rather the lack of alternatives that drives people into self-employment. In fact, the formal labor market in poor countries is relatively underdeveloped offering only limited opportunities. Furthermore, a high share of the workforce is poorly educated, causing difficulties to meet the job profiles. By implication, self-employed people in developing countries are likely to have received little or no education at all which, in turn, also suggests low labor productivities. Another particularity of self-employment in developing countries lies in its vulnerability and informality. Mostly, a formal contract and support by a social system is missing and its remuneration is entirely dependent on self-made profits (ILO 2014b). The risk of being active in the informal sector is especially high in Africa, Asia, Latin America and the Caribbean (ibid.). In developing nations, self-employment thus usually coincides with informality and hence often remains statistically unobserved. UN SNA standards demand to record the so-called shadow economy but due to its very nature national accounting often fails to do so as it can only be estimated (OECD 2004).

The processing of data is furthermore challenged by the unreliable scope, detail and quality of national accounting in developing economies (UN 2012). As demonstrated by Jerven (2012), especially national income accounting in Sub-Saharan Africa is fraught with inaccuracies and inconsistencies which impairs cross-country comparability. In this case, for example, it is not immediately obvious what is covered by the items 'compensation of employees' and 'mixed income' of the self-employed as they are not always standardized with international guidelines (i.e. SNA 2008). Dealing with international income data is thus likely to be subject to a quality-coverage-tradeoff, as it is common among broad cross-country data analyses: The more countries (especially developing countries) are covered, the greater the chance of poorer quality data and lack of comparability.

To meet these particular demands and challenges, the following section more thoroughly investigates labor income in general and self-employed income in particular before turning to the formulation of assumptions. This is done by consulting SAMs as additional sources of information. This prevents the exclusive reliance on macro-level national account data of previous studies.

# Social Accounting Matrices

A social accounting matrix is a square matrix that represents flows of all transactions that take place in an economy. Rows display the income of an account while columns denote its expenditures. SAMs draw a comprehensive picture of the economy and thus can reveal a country's economic structure much better than national accounts can do. They are constructed by matching and complementing national accounts, input-output tables, labor force surveys, household surveys et cetera. They bridge between income data collected at the household level and macro-economic data and therefore imply a higher reliability than national accounts. Already Pyatt and Round (1984) stressed the role SAMs could play in improving the quality of national accounts.

As SAMs disaggregate by sector and type of employment, they make the sectoral composition of labor visible. By this means, SAMs can be used to identify the sectors of the self-employed as well as the corresponding productivities and factor intensities. This information can be used as basis for a suitable estimation of the labor income of the self-employed. Unfortunately, SAMs are not available at large scale but the International Food Policy Research Institute (IFPRI) and the UN Development and Analysis Division (UN DESA) freely provide data for several developing countries. In addition, there are a number of country case studies available from various sources. Usually, the cross-entropy approach is applied to develop SAMs. This method exploits scattered and inconsistent data in a highly flexible and efficient way and thus can deal with the poor data situation of developing countries (Robinson 2001). Since SAMs are usually only constructed at large time intervals, the data mostly covers one observation per country.

SAMs can help to understand the substance of self-employed in two ways, namely by providing quantitative as well as qualitative information.

**5.1 Basic Descriptives on the Labor Share**

First, national factor income shares are extracted from SAMs, thereby compiling a small pool of data on the labor share. To ensure comparability, I restrict the data to SAMs from IFPRI and UN DESA that both apply the cross-entropy approach. Together, they provide 51 SAMs for 45 developing countries. Unfortunately, the size of the data pool is too small to conduct large-scale data analyses across time and space. Nevertheless, this data provides important basic descriptives about the size and distribution of labor shares in developing countries that can serve as benchmark. Empirical insights into the statics and the dynamics of the labor share can also later be used to perform a robustness check of national account data. Summary statistics of the 51 country-specific SAMs display that the labor share ranges between 0.24 and 0.71, with a mean and a median of 0.46. It is almost normally distributed as shown in Figure 1 which provides its probability density function obtained from Epachenikov kernel density estimates.

Figure 1: Probability density function of the labor share (from SAMs)

Source: Author’s illustration [based on IFPRI and UN DESA data]

Figure 2: Labor share and GDP of South Africa, 1993-2005

Source: Author’s illustration [based on IFPRI, UN DESA and World Bank WDI data]

By taking the example of South Africa, the data pool also allows to analyze the dynamics of the labor income share in the course of a country's economic development. South Africa is selected as case study mainly because it is the only country for which at least three SAMs covering a time span of more than ten years are available but also because it is a good example of a fast growing emerging country. Figure 2 shows how South Africa's real GDP as well as labor's share in GDP developed from 1993 until 2005. While real GDP increased by almost 50 per cent, the labor income share decreased by 7 percentage points (from 56.3. to 49.8 per cent). This change in the labor share is quite remarkable and an evidence of non-constant factor shares. The downward trend in the labor share of South Africa might just be an exception, but one cannot rule out the possibility that there is a general link between the economic development of a country and its factor income shares.

**5.2 Three Case Studies: Indonesia, Zambia and Bolivia**

Second, I take a closer look at individual SAMs that include an analysis of self-employed and their factor income. This is to better understand the essence of self-employment in the developing world and to integrate lessons learned into the process of computing the labor share. Hereby, the information content of micro- and macro-level data can be exploited in a meaningful manner. Studies on Zambia, Indonesia and Bolivia, which represent the three major developing regions Africa, Asia and Latin America, are used for this.

Thiele and Piazolo (2002) construct a SAM for Bolivia for 1997 which decomposes self-employed income into mixed income and employers' profits and identifies the sectors in which self-employment prevails. Thurlow et al. (2004), who present a 2001 SAM for Zambia, provide insights into the factor intensities, skill contents and productivities of each sector. Their study especially sheds light on the agricultural sector as it differentiates between small- and large-scale farming and itemizes land capital as production factor. Finally, Yusuf (2006) constructs a SAM for Indonesia for 2003 which calculates the labor and non-labor share of self-employment. Self-employed labor income is imputed on the basis of the average hourly wage in the corresponding employment sector for a similar type of labor (according to skills, activity and urban-rural location), multiplied with the number of working hours recorded by labor force surveys.[[6]](#footnote-6) Only self-employed income that exceeds the imputed labor income is counted as non-labor income.

Studying these SAMs reveals that self-employment in developing countries is more prevalent in rural than in urban areas. Self-employed mostly work in the agricultural sector and are smallholder farmers. Outside agriculture, self-employment usually takes the form of own-account enterprises that primarily engage in wholesale, retail or hospitality. A smaller share conducts light manufacturing (food manufacturing), especially in rural areas where processing of agricultural products is common. Employers are mostly active in modern agriculture and the service sector, mainly transport. On the other hand, self-employment hardly appears in the sectors of mining, manufacturing and textile.

It is furthermore observed that the sectors in which self-employment prevails are the labor-intensive sectors. For example, in the case of Indonesia, the labor share in the hospitality sector is 0.8, in the agricultural sector 0.65 and in the retail sector 0.58 (Yusuf 2006). At the same time, mining has with 0.2 the lowest labor share and also the one in the manufacturing and textile industry is only about 0.4. The SAM of Zambia takes a closer look at the agricultural sector and shows that smallholder agriculture is highly labor intensive: It displays a labor income share of 0.73, followed by a capital share of 0.2 and a land share of 0.07 (Thurlow et al. 2004). An exception are urban self-employed and employers who have a relatively high capital share (ibid.).

It also becomes apparent that in a given sector, self-employed pursue a more labor-intensive strategy than larger firms. This is also true for the agricultural sector. Thurlow et al. (2004) distinguish between small, medium and large-scale farms and show that the labor share of agricultural production decreases and the capital share increases with the farm size (while the land share does not show a specific pattern). The reason is often seen in the availability of family labor which gives smallholders easy access to manpower (World Bank 2013).

Furthermore, there is some evidence that the self-employed have much lower skills and less education compared to the rest of the workforce, suggesting that they are less productive per unit of labor than the employed staff. This seems to be true for both self-employed inside and outside agriculture. Most of labor in smallholder farming is uneducated and the skill content of labor is higher for larger farms. Similarly, urban own-account enterprises are less productive than their larger counterparts. This corresponds to their low educational attainment but is also due to their limited access to (financial) capital (World Bank 2013).

Finally, the SAMs give also some indication on how the labor income of the self-employed compares to employees' wages. It is found that smallholders and urban self-employed are worst off suggesting a relatively low labor income (besides a low income from capital). This also corresponds to the finding from Indonesia that the agricultural sector absorbs a large share of the workforce but contributes comparatively little to national GDP. An exception are urban employers who have a quite high labor as well as capital income; they are small in number though. It can be concluded that self-employed labor-income is at most equal but, more likely, lower than that of employees.

Of course, these are just individual cases but findings are consistent with the general literature on self-employment in developing countries (for example, Fields 2014, Fox and Sohnesen 2012, Mead and Liedholm 1998 or World Bank 2013).

# Measuring the Labor Share

**6.1 Steps of construction**

This section describes how the labor share data set is constructed. As illustrated above, there are mainly three alternatives to adjust the naive labor share for self-employment. The adjustment is a delicate exercise since it requires making assumptions about self-employed income as well as about its labor income share. To substantiate my assumptions, I rely on the qualitative and quantitative information about self-employment gained from SAMs.

Gollin's (2002) first adjustment treats all self-employed income as labor income. Although SAMs have shown that typical self-employed activities are associated with high labor shares, this approach is likely to overestimate the labor share, given that the capital share in agriculture, retail or hospitality usually accounts for one quarter to one third.

Gollin's (2002) second adjustment assumes self-employed income to contain the same mix of capital and labor income as the rest of the economy. This, by contrast, rather understates the labor share of developing countries since self-employed have been shown to systematically differ from employees in their factor shares and relatively more rely on labor.

Unfortunately, data on self-employed income is only reported by less than one third of developing countries on a regular and comprehensive basis (mainly countries in Latin America, Caribbean, Eastern Europe and Central Asia). Applying Gollin's (2002) first or second adjustment thus makes the sample size shrink. Yet, the remaining observations can provide some guidance on the level of adjusted labor shares: They range between 0.26 and 0.87 with a mean of 0.57 (first adjustment) and 0.21 and 0.73 with a mean of 0.46 (second adjustment).

Gollin's (2002) third alternative imputes the average wage sum of employees to the incorporate sector relying on the share of self-employed in the workforce. This method might overestimate the labor share. As shown in SAMs, self-employed, although more labor-intensive, are usually less productive and mostly worst off than employees. By imputing not the total but just a share of employees' wage bill, it can, however, be a meaningful starting point. As data on the self-employment share is insufficient as well, it is required to chose a proxy that can serve as indirect measure beforehand.

I select the share of employment in agriculture as proxy variable, assuming that most of the self-employed in poor countries are smallholders and most of the farm labor force is self-employed. The case studies of SAMs revealed that there is a clear overlap of self-employment and agricultural employment in developing countries. Of course, this proxy is more appropriate in some regions than in others, depending on the sectoral structure of an economy, and also disregards self-employed activities other than in agriculture. But the correlation of 0.80 between self-employment and the agricultural employment share shows that it serves as a good proxy. Aside from the high congruence, another strength of this proxy variable is the high availability and good quality of data. Data on the agricultural employment share is provided for almost all developing countries by either WB WDI or UN's Food and Agricultural Organization (FAOSTAT). Figure 3 illustrates the development of the agriculture employment share by region over time. Similar to what we know about self-employment in the developing world, it shows how the importance of agriculture varies across regions and how it declines in the course of time and with the economic development of a country.

Figure 3: Regional agricultural employment shares, 1990-2012

Source: Author’s illustration [based on FAOStat and World Bank WDI data]

After having chosen a proxy for self-employment, the labor income of the self-employed can be imputed in a next step. The full imputation seems to be suitable for many countries: In Eastern Europe, Central and East Asia, Middle East, North Africa, Latin America and the Caribbean, this adjustment yields labor shares that range between 0.17 and 0.82 and average at 0.5 or below, which is very close to the labor shares obtained from SAMs. Furthermore, where reported, it ranges between Gollin's (2002) first and second adjustment. I therefore hold on to this adjustment in these countries. At the same time, however, it yields implausibly high values for other countries (for example, 2.08 in China, 2.4 in Bhutan or 3.18 in the case of Burkina Faso). There may be three reasons for this: First, countries might already correct for labor income of the self-employed in their reported 'compensation of employees' such that any further amendment would mean a double adjustment. Second, the database simply may contain errors or agricultural employment may be inappropriate as proxy, both with the consequence that the adjusted labor share data becomes unreliable as well. Third, and most relevantly, the assumption behind this adjustment might not hold for all countries. Indeed, the assumption of same wages is always poor when there are systematic differences between the two sectors of employment.

Bhutan is certainly a case where it has already been corrected for the labor income of the self-employed in light of the fact that even the naive labor share amounts to 0.91. This seems also to be the case for a few other countries. So no further modification is added in countries where the naive labor share is already reasonably high (greater than 0.21) and an imputation of wages would overshoot (greater than 0.91).[[7]](#footnote-7)

Table 1: The constructed labor share and its components

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
|  | **Range** | **Mean** | **No. of Obs.** | **Form of Adjustment** |
| **Labor Shares from SAMs** | **0.21-0.71** | **0.46** | **51** | **Imputation of labor income based on average sectoral wages** |
| **Composed labor share** | **0.06-0.91** | **0.47** | **1396** | **Composition of the below adjustments** |
| Naive labor share | 0.24-0.91 | 0.53 | 79 (5.6 %) | No further adjustment |
| Gollin's 1st adjustment | 0.26-0.64 | 0.49 | 54 (3.9 %) | Adding all mixed income |
| Gollin's 2nd adjustment | 0.21-0.73 | 0.46 | 291 (20.8 %) | Adding a share of mixed income (same capital-labor ratio as in the rest of the economy) |
| Gollin's 3rd adjustment (based on agricultural employment share) | 0.06-0.87 | 0.46 | 716 (51.3 %) | Imputation of labor income based on total average wages |
| 0.11-0.77 | 0.47 | 256 (18.3 %) | Imputation of labor income based on 2/3 of average wages (in low and lower middle income countries in South Asia and Sub-Saharan Africa) |

Source: Author’s calculation [based on IFPRI, UN DESA, UN SNA, FAOStat and WB WDI data]

In other cases, the underlying data seems to be flawed as there are implausible jumps in 'employee compensation' to GDP ratio. A special case are post-Soviet states which all show a considerable labor share plunge in the early 1990s. Behind this fall is not only the heavy economic transformation but also stagnant statistics: Suddenly, a previously non-existing shadow economy sprang up in the former Soviet republics that the national statistical offices were not able to capture (Kaufman and Kaliberda 1996; Johnson, Kaufman and Shleifer 1997). Many formerly official workers began to work as self-employed in the informal economy and no longer appeared in official statistics. To correct for the increasing shadow economy and the related drop in the labor share, I leave the naive labor share in the years before the plunge so that incorrect upward adjustments are avoided.

The data of most countries with implausibly high labor shares, however, gives reason to conclude that the full imputation of average wages is not appropriate: Very high adjusted labor shares (above 0.91 or even above 1) go hand in hand with very low naive labor shares (below 0.21), suggesting that the actual labor share lies somewhere in between. The most extreme example is Burkina Faso which has a mean naive labor share of 0.21 that wage imputation lifts up to 3.18. An adjustment assuming relatively lower average wages for the self-employed seems to be more suitable in these countries. The cases concerned are the poorest of the poor, basically countries in Sub-Saharan Africa and South Asia, where the economy and formal labor markets are least developed and most people engage in low-productive subsistence farming (FAO 2012; World Bank 2013). For the low and lower middle income countries from these two regions, I therefore only impute a share of employees' wages. Following Bentolila and Saint-Paul (2003), the self-employed are assumed to earn on average two-thirds of employee income. Of course, this assumption is of arbitrary kind but the resulting adjusted labor shares appear reliable. They range between 0.11 and 0.77 in Sub-Saharan Africa and 0.42 and 0.75 in South Asia and, where available, move between the first and second adjustment or below.

After completing these steps, Gollin's (2002) first adjustment functions as upper and his second adjustment as lower bound in countries which report 'mixed income' and in case the so far adjusted labor share exceeds either of these limits.

The final labor share data set is thus based on Gollin's third assumption, using the agricultural employment share as proxy and imputing the full average wage of employees to the self-employed in the emerging regions and a fraction of two-thirds in the less developed regions of South Asia and Sub-Saharan Africa, correcting for obvious measurement errors, leaving the naive share where already accounted for self-employed income and framing the whole with Gollin's first and second adjustment whenever available.

**6.2 Validation of Data**

Table 1 summarizes the constructed labor share and its components. The final data set covers 100 low and middle income countries from 1990 until 2011. It is an unbalanced panel and has in total 1396 observations. The labor share ranges from 0.06 to 0.91 with a mean of 0.47, which is much lower than the 'two-thirds' standard in economic literature. Due to the systematic differences between self-employed and employees in developing countries, Gollin's (2002) third correction runs risk to overestimate the labor share of developing countries. The summary statistics suggest that this is not the case though. The results are also very close to the findings from SAMs, giving further support to the way of construction.

Figure 4: Labor share of South Africa (in %), 1990-2011

Source: Author’s illustration [based on UN SNA, FAOStat, World Bank WDI data, IFPRI and UN DESA]

When looking again at the case study of South Africa (see Figure 4), we can see that Gollin's (2002) third adjustment based on the agricultural employment share is an appropriate way of proceeding. My constructed labor share is similar to the one obtained from the country's SAMs, not only in terms of level but also trend.[[8]](#footnote-8) Leaving the naive share as it is, on the other hand, would understate the labor share.

We can also learn from looking at the temporal development of the constructed labor share in comparison to its components (see Figure 5). As the labor share data set is mainly based on Gollin's third adjustment, both show a similar pattern. The composed labor share is also just about the same level as Gollin's second adjustment, indicating that the imputation of labor income from employee compensation results in an average capital-labor mix of self-employment like in the rest of the economy. The composed labor share is centered between the naive labor share (no self-employed income) and Gollin's first adjustment (all self-employed income) which further suggests that the labor share of self-employment is on average close to one half. What also immediately becomes visible is the non-constant, negative performance of the labor share over time, which is investigated more closely in the next section.

Figure 5: Different labor share adjustments (in %), 1990-2011

Source: Author’s illustration [based on UN SNA, FAOStat and World Bank WDI data]

# Properties of the Data Set

## 7.1 Descriptives on the Labor Share

After having prepared the labor share data, it can be used to investigate global trends in the development of the labor share over time as well as differences across developing regions. A first important finding is that labor's relative income in developing countries has declined over time (see Figure 5), suggesting that labor income has been lagging behind overall productivity increases. My results thus provide strong evidence against the hypothesis of constant labor shares in the long-run and are consistent with the findings of previous studies.

The labor share is found to be stable in the early 1990s but starts declining with the end of the Cold War. In the post-1993 era, it on average falls by about 10 percentage points (from 53.7 to 43.6 per cent). The labor share recovers slightly in the late 2000s in the course of the Global Financial Crisis of 2007-8 but continues falling afterwards. This temporarily reversed trend mainly goes back to the countercyclical movement of the labor share, meaning that capital owners usually loose more than wage earners during crises (ILO 2013).

It should be emphasized that the fall in the labor share is independent of the form of measurement. It is noticeable that even the naive share, which only captures wage employment, is decreasing significantly over time. It is a well-known fact that the labor share has fallen in high-income economies over the last two decades. This is mainly explained with capital-augmenting technological progress and the specialization into capital-intensive commodities in the course of globalization − an argument based on the factor-proportion models by Heckscher, Ohlin, Stolper and Samuelson (Heckscher 1919; Ohlin 1933; Samuelson 1948; Stolper 1941). To the extent that labor is abundant in developing countries, one would rather expect the labor share in developing countries to rise with economic growth and international integration. This should be especially the case for the naive share, considering that development is associated with the expansion of the (assumingly labor intensive) corporate sector. The reasons for the decreasing labor share might be different in various contexts. The literature mainly discusses the effects of increasing automation, worldwide competition and unemployment as well as the decreasing bargaining power of labor.

As shown in Figure 6, this downward trend of the labor share has been present in most developing regions. East Asia and the Pacific is the region which experienced the fastest decrease (on average 14 percentage points since 1990), closely followed by Eastern Europe and Central Asia (11 percentage points) and Latin America and the Caribbean (10 percentage points). Only in the second half of the 2000s, the downturn of labor shares came to a halt. A considerable decline also occurred in Sub-Saharan Africa, where labor shares fell by 6 percentage points between 1990 and 2011. Exceptions to the downward trend are only South Asia (Sri Lanka, Bhutan and India), the Middle East and North Africa whose labor shares fluctuated but more or less remained on a constant level.

Figure 6: Labor shares by region, 1990-2011

Source: Author’s illustration [based on UN SNA, FAOStat and World Bank WDI data]

In terms of levels, South Asia exhibits the highest labor shares (0.7 on average), which is, however, mainly due to Bhutan data. With labor shares ranging between 0.4 and 0.6, East Asia, Eastern Europe and Central Asia, Latin America and the Caribbean are mid-table. Sub-Saharan Africa has slightly lower labor shares moving around 0.4, while labor shares in the Middle East and North Africa remain below 0.4 which is not surprising in view of the fact that most of the oil-producing countries are located in this region.

Figure 7 displays labor shares for different income groups according to the country classification by World Bank (2011). The level of a country's labor share seems to be independent of a country's stage of development and also the negative trend occurs in all income groups. However, it is more pronounced in low income countries, followed by lower middle income and finally upper middle income countries.

Figure 7: Labor Shares by Income Classification, 1990-2011

Source: Author’s illustration [based on UN SNA, FAOStat and World Bank WDI data]

7.2 Unit Roots

Descriptives on labor share data already provide some evidence against the long-prevailing hypothesis of constant factor shares. Nevertheless, many theoretical models are still based on the Cobb-Douglas production technology or similar constructs that treat the labor share as a persistent variable. For future applications, it is therefore important to be on notice of the possible presence of unit roots in the labor share data. If the labor share is used as variable in regression analyses − for example, as dependent variable to explore the reasons behind its negative time trend − and it follows a unit root process together with another variable in the model, the problem of spurious regression may arise (Granger and Newbold 1974). This renders inference of conventional methods invalid and therefore would require special estimation techniques, such as a cointegration method or estimation in first differences (Greene 2003; Enders 2010). I therefore test for the presence of unit roots, using a Fisher test statistic as presented in Maddala and Wu (1999):

where π*i* is the p-value of any unit root test for each cross section *i*.[[9]](#footnote-9) Within this framework, the augmented Dickey-Fuller (1979) test is performed which can be considered as the most common method, testing the null hypothesis of unit root against the alternative of stationarity. The Fisher test statistic is preferred over other test statistics, such as Im-Pesaran-Shin, as it is not an asymptotic but exact test which does not require a balanced panel. The results, however, should still be treated with caution, given the low power of unit root tests in finite samples like here (Blander and Dhaene 2012).

Table 2 shows the test results. The test comes in three versions (without constant and trend, with trend and with constant) and is run on the first, second and third lag which are appropriate lag lengths for annual data. The null of unit root is rejected for all versions of the test at the one per cent level of statistical significance, thus providing strong evidence against the persistence of the labor share in developing countries.[[10]](#footnote-10)

Table 2: Augmented Dickey-Fuller Tests

|  |  |  |  |
| --- | --- | --- | --- |
| **Lags** | **No Trend** | **With Trend** | **With Constant** |
| 1 | χ2(174)=341.4\*\*\* | χ2(174)=409.1\*\*\* | χ2(164)=478.6\*\*\* |
| 2 | χ2(170)=253.7\*\*\* | χ2(170)=282.4\*\*\* | χ2(152)=398.5\*\*\* |
| 3 | χ2(162)=319.8\*\*\* | χ2(162)=253.4\*\*\* | χ2(142)=363.3\*\*\* |

degrees of freedom in parantheses, \*\*\* p<0.01

Source: Author’s illustration

# Conclusion

This paper reveals that measuring the labor share of developing countries is neither direct nor straightforward. There clearly is a quality-coverage trade-off regarding its computation, meaning that the more global the coverage, the greater the prevalence of poor quality data. This may induce to measure the labor share only at the national or at most regional level. However, giving up on its cross-country measurement cannot be the consequence. After all, research on the labor share simply is too important to be hampered by a poor data situation. Although different developing regions can hardly be measured with the same yardstick, global data sets are required to analyze broad trends in labor shares. Up to date, the development of the labor share in low and lower middle income countries remains unexplored, which is a clear gap in economic literature. Labor share data can, for example, be used to better research the distributional effects of labor market policies, the trend of factor and personal income inequalities in the process of development or the distribution of potential gains from globalization. Particularly in developing countries, knowledge about the labor share can be used to develop poverty reduction strategies. For example, the finding that productivity gains in developing countries do not translate into broad wage increases suggests that the opportunity of increasing the living standard of the poor is missed. In fact, the labor share does not only influence income inequality within a country but also has significant implications for aggregate demand and thus growth.[[11]](#footnote-11) In this function, it can serve as a decision-making parameter in favor or against a certain development path, such as export-led growth.

This paper contributes to the literature by providing the necessary data to address these globally relevant issues also for the developing world. SAMs, a micro-funded representation of the economy, provide input and feedback for the construction of labor share data. Yet, future research on the labor share depends crucially on more stalwart and robust data. Counter-checking national accounts with micro-economic data can only be a second best option. It is hence recommended that national statistics offices increase their effort in gathering data on the (informal) self-employment sector. For this, as Jerven (2012) and others demand, more funding and qualified personnel directed towards reliable and regular data collection will be necessary. Until high quality data is available, it is inevitable to conduct robustness checks on the national account data, with SAMs being just one possibility here.

Notwithstanding these restraints, the data set can provide first reliable and valuable insights into the capital-labor ratio in the developing world over the last decades. It is found that the average level of the labor share is with 0.47 much less than the established "two-thirds" in economic literature. Similarly, in contrast to the long-lasting belief of constant factor shares, it becomes evident that there is a significant downward trend of the labor share since the early 1990s. In this way, my findings confirm Piketty's (2014) prominent hypothesis of wealth accumulation faster than economic growth also for most of the developing regions: Income generation more and more shifts from labor to capital, implying nothing else than a shrinking labor share. An increase in human capital seems not to be able to reverse this trend. Future research should focus on the potential reasons behind this development. The downturn takes place during times of increased globalization and evolving financialization, suggesting a link between these phenomena. It is therefore proposed to investigate these relationships and explore whether there is not only a correlation but also causation in place.

**References**

Atkinson, A.B., and F. Bourguignon (eds.) (2015). *Handbook of Income Distribution*. North Holland: Elsevier.

Arpaia, A., E. Pérez, and K. Pichelmann (2009). 'Understanding labour income share dynamics in Europe'. MPRA Paper, No. 15649. Munich: University of Munich.

Bentolila, S., and G. Saint-Paul (2003). 'Explaining Movements in the Labor Share'. *Contributions to Macroeconomics*, Vol. 3(1).

Bernanke, B.S., and R.S. Gürkaynak (2001). ' Is Growth Exogenous? Taking Mankiw, Romer, and Weil Seriously'. NBER Macroeconomics Annual 2001, Vol. 16. Cambridge: National Bureau of Economic Research.

Blanchard, O.J. (1997). 'The medium run'. Brookings Papers on Economic Activity, Vol. 2. Washington DC: Brookings.

Blander, R.D., and G. Dhaene (2012). 'Unit Root Tests for Panel Data with AR(1) Errors and Small T'. *The Econometrics Journal*, Vol. 15(1): 101-24.

Cobb, C.W., and P.H. Douglas (1928). 'A Theory of Production'. *American Economic Review*, Vol. 18 (Supplement): 139–65.

Daudey and García-Peñalosa (2010). 'Labour Market Institutions and the Personal Distribution of Income in the OECD'. *Economica*, Vol 77(307): 413-50.

Decreuse, B., and P. Mareek (2013). 'FDI and the labor share in developing countries: A theory and some evidence'. Thema Working Paper, No. 2013-20. Cergy-Pontoise: University of Cergy- Pontoise.

Dickey, D.A., and W.A. Fuller (1979). 'Distribution of the Estimators for Autoregressive Time Series With a Unit Root'. *Journal of American Statistical Association*, Vol. 74(366): 427-31.

Diwan, I. (2001). 'Debt as Sweat: Labor, Financial Crisis, and the Globalization of Capital'. Mimeo. Washington DC: World Bank.

Douglas, P.H. (1967). 'Comments on the Cobb-Douglas production function'. In M. Brown (ed), *The Theory and Empirical Analysis of Production*. New York: Columbia University Press.

Douglas, P.H. (1976). 'The Cobb-Douglas Production Function Once Again: Its History, Its Testing, and Some New Empirical Values'. *Journal of Political Economy*, Vol. 84(5): 903-16.

Ellis, L., and K. Smith (2007). 'The global upward trend in the profit share'. BIS Working Papers, No. 231. Basel: Bank for International Settlements.

Enders, W. (2010). *Applied Econometric Time Series*. New Jersey: John Wiley & Sons.

European Commission (2007). 'Employment in Europe'. Brussels: European Commission.

European Commission (2012). 'Annual Macro-Economic Database'. http://ec.europa.eu/economy \_finance/db\_indicators/ameco/index\_en.htm. Access on 19th May 2012.

Fields, G. (2014). 'Self-employment and poverty in developing countries'. *IZA World of Labor*, Vol. 60.

Food and Agriculture Organization (2012). 'Statistical Yearbook'. Rome: Food and Agricultural Organization.

Fox, L., and T.P. Sohnesen (2012). 'Household Enterprises in Sub-Saharan Africa − Why They Matter for Growth, Jobs, and Livelihoods'. Policy Research Working Paper. Washington DC: World Bank.

Gollin, D. (2002). 'Getting Income Shares Right'. *Journal of Political Economy*, Vol. 110(2): 458-74.

Guerriero, M., and K. Sen (2012). ' What Determines the Share of Labour in National Income? A Cross-Country Analysis'. IZA Discussion Papers, No. 6643. Bonn: Institute for the Study of Labor.

Guscina, A. (2006). 'Effects of Globalization on Labor's Share in National Income'. Working Paper, Vol. 294. Washington DC: International Monetary Fund.

Granger, C.W.J., and P. Newbold (1974). 'Spurious Regressions in Econometrics'. *Journal of Econometrics*, Vol. 2: 111-20.

Greene, W.H. (2003). *Econometric Analysis*. Upper Saddle River: Prentice Hall.

Hardoon, D. (2015). 'Wealth: Having it all and wanting more'. Oxfam Issue Briefing. Oxford: Oxfam.

Harrison, A. (2005). 'Has Globalization Eroded Labor's Share? Some Cross-Country Evidence'. MPRA Paper. Munich: University of Munich.

Heckscher, E. (1919). 'The Effects of Foreign Trade on the Distribution of Income'. *Ekonomisk Tidskrift*, Vol. 21: 497-512.

Hutchinson, J., and D. Persyn (2012). 'Globalisation, Concentration and Footlose Firms: In Search of the Main Cause of the Declining Labor Share'. *Review of World Economics*, Vol. 148(1): 17-43.

Inklaar, R., and M. Timmer (2013). 'Capital, labor and TFP in PWT 8.0'. Groningen Growth and Development Center. Groningen: University of Groningen.

International Labor Organization (2013): 'Global Wage Report'. Geneva: International Labor Organization.

International Labor Organization (2014a). 'Current Guidelines'. http://www.ilo.org/global/statistics- and-databases/statistics-overview-and-topics/status-in-employment/current-guidelines/lang-- en/index.htm. Access on 11th August 2014.

International Labor Organization (2014b), 'Global Employment Trends 2014'. Geneva: International Labor Organization.

International Labor Organization (2014c), 'World of Work Report 2014'. Geneva: International Labor Organization.

International Monetary Fund (2007). 'World Economic Outlook'. Washington DC: International Monetary Fund.

Ivanic, M. (2004). 'Reconciliation of the GTAP and household survey data'. GTAP Research Memorandum, Vol. 5. West Lafayette: Center for Global Trade Analysis.

Jaumotte, F., and I. Tytell (2007). 'How Has The Globalization of Labor Affected the Labor Income Share in Advanced Countries?'. Working Paper, Vol. 298. Washington DC: International Monetary Fund.

Jayadev, A. (2007). 'Capital Account Openness and the Labor Share of Income'. *Cambridge Journal of Economics*, Vol. 31(3): 423-43.

Jerven, M. (2012). 'Comparability of GDP Estimates in Sub-Saharan Africa: The Effect of Revisions in Sources and Methods Since Structural Adjustments'. *Review of Income and Wealth*, Vol. 59(S1): S16-36.

Johnson, S., D. Kaufman, and A. Shleifer (1997). 'The Unofficial Economy in Transition'. Brookings Papers on Economic Activity, Vol 2. Washington DC: Brookings.

Kaldor, N. (1957). ' A Model of Economic Growth'. *The Economic Journal*, Vol. 67(268): 591-642.

Kaufman, D., and A. Kaliberda (1996). 'Integrating the Unofficial Economy into the Dynamics of Post-Socialist Economies'. Policy Research Working Paper, No. 1691. Washington DC: World Bank.

Krueger, A.B. (1998). 'Measuring Labor's Share'. *The American Economic Review*, Vol. 89(2): 45-51.

Maddala, G.S., and S. Wu (1999). 'A Comparative Study of Unit Root Tests with Panel Data and a New Simple Test'. *Oxford Bulletin of Economics and Statistics*, Vol. 61: 631-52.

Marterbauer, M., and E. Walterskirchen (2003). 'Bestimmungsgründe der Lohnquote und der realen Lohnstückkosten'. WIFO-Monatsberichte, Vol. 2. Vienna: Österreichische Institut für Wirtschaftsforschung.

Mead, D. C., and C. Liedholm (1998). 'The Dynamics of Micro and Small Enterprises in Developing Countries'. *World Development*, Vol. 26(1): 61-74.

Niño-Zarazúa, M., L. Roope and F. Tarp (2014). 'Global interpersonal inequality: Trends and measurement'. Working Paper, 2014/004. Helsinki: UNU-WIDER.

OECD (2004). 'OECD Employment Outlook'. Paris: OECD Publishing.

OECD (2012). 'OECD Employment Outlook'. Paris: OECD Publishing.

Ohlin, B. (ed.) (1933). *Interregional and International Trade*. Cambridge: Havard University Press.

Ortega, D., and F. Rodriguez (2001). 'Openness and Factor Shares'. Mimeo. College Park: University of Maryland.

Piketty, T. (ed.) (2014). *Capital in the Twenty-First Century*. Cambridge: Harvard University Press.

Pyatt, G. and J. Round (1985). 'Social Accounting Matrices: A Basis for Planning'. World Bank Symposium. Washington DC: World Bank.

Ray, D. (1998). *Development Economics*. Princeton: Princeton University Press.

Richardson, J.D., and E.B. Khripounova (1998). 'U.S. Labor Market Power and Linkages to International Trade: Identifying Suspects and Measures'. Mimeo. Washington DC: U.S. Department of Labor.

Robinson, S., A. Cattaneo, and M. El-Said (2001). 'Updating and Estimating a Social Accounting Matrix Using Cross Entropy Methods'. *Economic Systems Research*, Vol. 13(1): 47-64.

Rodriguez, F., and A. Jayadev (2010). 'The Declining Labor Share of Income'. Human Development Reports Research Paper, Vol. 36. New York: United Nations Development Programme.

Samuelson, P.A. (1948). 'International Trade and the Equalisation of Factor Prices'. *The Economic Journal*, Vol. 58(230): 163-84.

Slaughter, M. (2001). 'International Trade and Labor Demand Elasticities'. *Journal of International Economics*, Vol. 54(1): 27-56.

Stolper, W., and P.A. Samuelson (1941). 'Protection and Real Wages'. *Review of Economic Studies*, Vol. 9(1): 58-73.

Thiele, R., and D. Piazolo (2002). 'Constructing a Social Accounting Matrix with a Distributional Focus − The Case of Bolivia'. Kiel Working Paper, Vol. 1094. Kiel: Kiel Institute of the World Economy.

Thurlow, J., D. Evans, and S. Robinson (2004). 'A 2001 Social Accounting Matrix for Zambia'. Washington DC: International Food Policy Research Institute.

United Nations (2011). 'System of National Accounts'. http://data.un.org/Explorer.aspx?d=SNA. Access on 20th March 2011.

United Nations (2012). 'Report of the Friends of the Chair on the barriers to the implementation of the System of National Accounts 1993'. New York: Economic and Social Council Statistical Commission.

United Nations Development Programme (2013). 'Humanity Divided'. New York: Bureau for Development Policy.

World Bank (2011). 'A Short History'. http://data.worldbank.org/about/country-classifications/a- short-history. Access on 23rd March 2011.

World Bank (2013). 'World Development Report 2013'. Washington DC: World Bank.

Yusuf, A.A. (2006). 'Constructing Indonesian Social Accounting Matrix for Distributional Analysis in the CGE Modelling Framework'. MPRA Paper, No. 1730. Munich: University of Munich.

**Appendix**

*Countries Included*

Algeria, Argentina, Armenia, Azerbaijan, Bahrain, Belarus, Benin, Bhutan, Bolivia, Bosnia and Herzegovina, Botswana, Brazil, Bulgaria, Burkina Faso, Burundi, Cameroon, Central African Republic, Chad, Chile, China, Colombia, Costa Rica, Cote d'Ivoire, Croatia, Cuba, Czech Republic, Democratic Republic of Congo, Djibouti, Dominican Republic, Ecuador, Egypt, Estonia, Fiji, Gabon, Georgia, Greece, Guatemala, Guinea, Honduras, Hungary, India, Iran, Iraq, Jamaica, Jordan, Kazakhstan, Kenya, Kyrgyz Republic, Lao People's Democratic Republic, Latvia, Lebanon, Lesotho, Lithuania, Macedonia, Malta, Mauritania, Mauritius, Mexico, Moldova, Mongolia, Morocco, Mozambique, Namibia, Netherlands Antilles, Nicaragua, Niger, Nigeria, Oman, Panama, Papua New Guinea, Paraguay, Peru, Philippines, Poland, Portugal, Republic of Korea, Romania, Russia, Saudi Arabia, Senegal, Serbia, Seychelles, Sierra Leone, Slovak Republic, Slovenia, South Africa, South Sudan, Sri Lanka, Sudan, Suriname, Tajikistan, Tanzania, Thailand, Trinidad and Tobago, Tunisia, Turkey, Ukraine, Uruguay, Venezuela, Zimbabwe

1. According to the International Labor Organisation (ILO), self-employed comprise employers, own-account workers, members of producer's cooperatives and contributing family workers as self-employed (ILO 2014a). [↑](#footnote-ref-1)
2. Available in the World Top Income Database. [↑](#footnote-ref-2)
3. That is (1) Gollin's first adjustment assuming all of agricultural value added to be labor income, (2) his second adjustment assuming UN mixed income to partly consist of labor and capital income, (3) his third adjustment using data on the number of self-employed from ILO's LABORSTA, or (4) the naive share. [↑](#footnote-ref-3)
4. For example, an important guideline is that the lowest labor share from either (1) or (3) is choosen, given the high risk of overestimation. [↑](#footnote-ref-4)
5. For example, it averages 0.52 and 0.46 in Inklaar and Timmer (2013) and Harrison (2005) respectively. [↑](#footnote-ref-5)
6. This strategy is, among others, also applied by Ivanic (2004) for the Global Trade Analysis Project (GTAP) database. [↑](#footnote-ref-6)
7. The marking values stem from the most extreme labor shares observed in SAMs, the naive share and after Gollin's (2002) first and second adjustment, see Table 1. [↑](#footnote-ref-7)
8. Unfortunately, data on mixed income is not available for South Africa. [↑](#footnote-ref-8)
9. The test is χ2-distributed with 2N degrees of freedom. [↑](#footnote-ref-9)
10. The degrees of freedom equal the number of countries in the data times two. This number, however, varies across the different tests because data gaps prevent to perform the test on all countries. [↑](#footnote-ref-10)
11. Labor income is assumed to have a higher consumption share than capital income. [↑](#footnote-ref-11)