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Globalisation versus Informality: Evidence from developing countries

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Globalisation versus Informality: Evidence from developing countries

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Abstract

A number of theoretical studies have tended to trace the nature of globalization process' impacts (mostly characterised by trade opening) on informality, while relevant empirical literature has been not well developed. The paper aims to fill this knowledge gap by shedding further light on the linkages running from globalisation to informality in developing countries. Moreover, in this study, globalisation is characterised not only by trade integration but also by other globalisation aspects, such as social globalisation, financial globalisation and so forth. To achieve the main objective, we employ the Bayesian statistical techniques, which allow one to determine, from a large set of different globalization indicators, a subset of indicators most likely to influence the size of informality. Our finding reveals that the indicators with consistently high inclusion probabilities are trade integration, trade reforms, de jure financial openness and social globalisation. On the other hand, many covariates found significant in previous empirical studies are not robust to including in informality modelling.

JEL : F16, F41, O17, J21

Keywords: Globalisation; Informality; Bayesian Model Averaging; Developing countries

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1. Introduction

Over the last decade, globalisation has become a phenomenal aspect of the world economy. In terms of trade globalisation, by 2008, the ratio of trade flows to the world GDP valued at around 60%, compared with less than 40% in the mid-1990s. Similarly, in terms of cross-border financial transactions, FDI net flows reached more than 6% of the world GDP, while this figure only attained to less than 2.5% in the mid-1990s.¹ By and large, globalisation process is believed to strongly affect the world economic growth as well as other macroeconomic aspects. Among others, this paper pays a special attention to possible causal linkages running from globalisation to informality in developing countries.

In the literature, the existing theoretical works have seemed to only focus on possible effects of "trade globalisation" aspect on informality. Concerning this issue, existing theoretical studies can be classified into two groups. Firstly, basing on the Harris-Todaro (1970) dual-economy model of rural-urban migration, Chandra and Khan (1993) and Marjit and Beladi (2007a) suggest that tariffs reduction results in a raise in both employment and wages in the informal sector when informal output is traded. By contrast, according to Beladi and Yabuuchi (2001), trade opening may decrease the size of informal employment when informal outputs are used as intermediate inputs in the formal sector. Secondly, in the light of trade models with differentiated wages, when capital is sufficiently mobile, trade integration can boost both informal employment and informal wages (Marjit and Maiti, 2005; Kar et al., 2003). In order to validate different theoretical hypotheses, empirical works have been developed and mostly available for a small group of Latin American countries. According to these studies, impacts of trade opening on informal economy strongly depend on country-specific circumstances. For instance, trade integration reduces informal activities in Mexico (Maloney, 1998), but increases informality size in Colombia and has no significant impact on Brazilian informal employment (Goldberg and Pavcnik, 2003b).

While globalisation process is manifold dimensions: economic (including trade and financial), social, political, cultural, environmental and so forth, the previous cited studies have only deepened our understanding of trade opening's impacts on informality. They have seemed to ignore impacts of other important globalisation aspects. Recently, using a new database of International Labour Organisation (ILO), Bacchetta et al. (2009) tend to clarify the multi-dimension of globalisation process as well as to investigate its impacts on informal

¹ Author's computation from WDI data

employment of developing countries. In order to capture the multifaceted nature of globalisation, the authors introduce in their estimated models a large set of globalisation indicators drawn from various international sources. The authors draw a mixed picture of globalisation's impacts on informal employment in developing countries. On one hand, they suggest that more open economies may have a lower incidence of informal employment. On the other hand, trade reforms seem to associate with higher informal employment. Similarly, larger FDI inflows may lead to an increase in informal employment. Complementing to the work of Bacchetta et al. (2009), Fugazza and Finess (2010) use three different data sets to assess the sign of such a complex relationship. This work also draws a mixed picture and no clear-cut conclusion regarding the connection between globalisation and informal employment. In a cointegration framework, more openness to trade is associated with greater informal employment and output for most countries, while lower trade restrictions appear to generate lower informal employment and output in most cases. However, the system-GMM estimation generates contrasting results that fewer trade restrictions associate with more informal output but less informal employment. Up to now, the works developed by Bacchetta et al. (2009) and Fugazza and Finess (2010) can be seen as the pioneer ones that endeavour to trace possible impacts of all globalisation dimensions on informal employment of developing countries. These works have focused on regression models involving a large and specific set of covariates collected from various data sources and regrouping information on different aspects of globalisation. Nevertheless, this empirical strategy seems to ignore uncertainty regarding the model specification itself, which can have dramatic consequences on inference, because the inference regarding the effects of included covariates can depend critically on the inclusion versus the exclusion of other covariates. Consequently, the potential uncertainty problem may lead to little agreement on the relationship between globalisation and informality between these two recent empirical studies.

So that, in our present paper, we aim to revisit potential affects of different globalisation dimensions on informality in developing countries by employing another empirical technique that is the Bayesian Model Averaging (BMA) technique. Employing the BMA technique allows us to confront uncertainty regarding the appropriate set of covariates to include in a regression model explaining informality. Moreover, differing from earlier empirical works, which aim to clarify the nature (positive or negative, significant or insignificant) of globalisation's impacts on informal employment, the main objective of this paper is to

determine, from a large set of potential covariates, a subset of globalisation indicators with high inclusion probabilities in the “true” informality model.

The reminder of this paper is organised as follows. Instead of providing a brief literature review of globalisation’s impacts on informality that has been extensively documented in the orthodox work of Bacchetta et al. (2009), *Section 2* outlines our empirical strategy that encompasses specifying the BMA technique. *Section 3* describes different datasets being used for the testing. *Section 4* reports and analyses the econometric results as well as robustness checks, and makes comparisons to related literature. Concluding remarks are in *Section 5*.

2. Empirical model specifications

As mentioned, to achieve the research objective, we employ here the BMA technique for which the statistical foundation is extensively documented in excellent surveys by Raftery (1995) and Hoeting et al. (1999). In the empirical literature, the BMA technique has been widely accepted as a way of accounting for model uncertainty, particularly in regression models for identifying the determinants of economic growth (e.g. Leamer, 1978; Levine and Renelt, 1992; Fernandez et al., 2001a, b; Sala-i-Martin et al., 2004). Moreover, BMA has been also used to evaluate an extensive set of potential determinants of other macroeconomic variables, notably FDI determinants (e.g. Eicher et al., 2011; Blonigen and Piger, 2011). Hence, in this paper, applying the BMA technique allows us to identify globalisation indicators, which truly affect informality in developing countries, from a large set of potential indicators. Specifically, we focus on the following linear regression model:

$$y = \alpha_i + X_i\beta_i + \varepsilon \quad \text{with } \varepsilon \approx N(0, \sigma^2 I) \quad (1)$$

where y is the dependent variable holding the size of informality, α_i is a constant, X_i is a set of potential independent variables explaining the dependent variable y , β_i is the coefficients, and ε is a normal independent and distributed (IDD) error term with mean zero and variance σ^2 . If there are many potential explanatory variables in a matrix X , we are interested in two questions: Which appropriate variables $X_i \in \{X\}$ should be included in the model? And how important are they? However, the fact is that the direct approach to do inference on a single linear model with all potential explanatory variables X is inefficient or even infeasible with a limited number of observations.

Now, suppose X contains K potential variables and the model uncertainty problem in (1) is mainly due to the selection of variable X_i . In this case, a particular selection of X_i defines the i^{th} model, denoted M_i . If the variable $X_i \in \{X\}$ can be freely included in the regression model, there are 2^K variable combinations and thus 2^K models to consider. The Bayesian approach tackles the uncertainty problem by estimating alternative models for all possible combinations of $\{X\}$ and constructing a weighted average over all of them. In the Bayesian words, the posterior probability that M_i is the "true" model of y is as follow:

$$p(M_i|y, X) = \frac{p(y|M_i, X)p(M_i)}{p(y|X)} = \frac{p(y|M_i, X)p(M_i)}{\sum_{j=1}^{2^K} p(y|M_j, X)p(M_j)} \quad (2)$$

where $p(M_i)$ is the prior probability that M_i is the "true" model, $p(y|X)$ denotes the marginal or integrated likelihood which is constant over all models and is thus simply a multiplicative term. Therefore, the posterior model probability (PMP) $p(M_i|y, X)$ is proportional to the marginal likelihood of the model $p(y|M_i, X)$.

Turning now to our regression model, the PMP can be used to confront the model uncertainty present in the informality regression. The PMP allows us to select the "true" model with highest posterior probability and then to make inference about other alternative models based on this "best" model. Nevertheless, if the PMP is widely dispersed across a large number of models, making inferences from a single and highest probability model can yield biased empirical results. Hence, instead of the PMP, the BMA technique calculates the averaging posterior inference across all alternative models. Precisely, for any statistic θ , the BMA posterior distribution is calculated as:

$$p(\theta|y, X) = \sum_{i=1}^{2^K} p(\theta|M_i, y, X)p(M_i|X, y) \quad (3)$$

where $p(\theta|M_i, y, X)$ represents the posterior distribution for θ conditional on M_i . The BMA posterior distribution $p(\theta|y, X)$ is not conditioned on a particular model, which is the "true" model, but is conditioned on the data. Hence, BMA is believed to confront uncertainty regarding the appropriate set of covariates included in a regression model.

In order to implement BMA, specifications for both the prior model probability $p(M_i)$ and the prior density function for the parameters $\{\alpha, \beta, \sigma\}$ of M_i are required. Suppose the prior

model probability is $p(M_i)$ uniform with respect to all alternative models, it is calculated as follow:

$$p(M_i) = \frac{1}{2^K} \quad (4)$$

The prior model probability specification (4) is a common choice in BMA approaches. It means that the prior probability of any single covariate in the “true” model is at least 50%. Unlike the specification of prior model probability $p(M_i)$, that of prior parameter densities is not an easy task.

In our analysis, we use two different automatic procedures for setting parameter priors. First, we take into consideration a Bayesian regression linear model with a specific prior structure called “Zellner's g prior”. Following this procedure, each individual model M_i is considered as a normal error structure. In order to specify the priors on the model parameters, we assume that constant α_i and error term ε have improper priors, meaning they are evenly distributed. Hence, the implementation of BMA only depends on the specification of curial prior probability of coefficients β_i . Here, we formulate the prior beliefs on coefficients into a normal distribution with a specified mean and variance. According to Zellner's g setting, the variance

structure is defined as follow: $\sigma^2 \left(\frac{1}{g} X_i' X_i \right)^{-1}$ with $\beta_i | g \approx N \left(0, \sigma^2 \left(\frac{1}{g} X_i' X_i \right)^{-1} \right)$

The previous formula means that the coefficients β_i are believed to be zero and that their variance-covariance structure is in line with that of X. The parameter g embodies how certain one is that the coefficients β_i are zero. A small g corresponding to few prior coefficient variance implies one is quite certain that the coefficients are indeed zero. By contrast, a large g means that one is very uncertain that the coefficients are zero. Hence, the specification of prior of coefficients β_i concerns the value of the parameter g. In the literature, g value is set through a popular default approach so-called the Unit Information Prior (UIP), which sets $g = N$ commonly for all models M_i . According to Kass and Wasserman (1995) and Raftery (1995), the UIP setting suggests a convenient approximation to the marginal likelihood based on the Bayesian Information Criterion (BIC). For this reason, the UIP procedure seems to be simple to implement.

In the second step, as a robustness check, we employ an alternative prior specification developed by Fernandez et al. (2001a) (henceforth, the FLS). Unlike the UIP procedure, the FLS setting should be applied if one wishes to use little subjective information in setting prior densities. In addition, Fernandez et al. (2001a) employ a uniform model prior and the birth-death Markov Chain Monte Carlo (MCMC) sampler.² Differing from the UIP approach with $g = N$, in the FLS procedure g prior is set to $g = \max(N; K^2)$.³

3. Data issues

Our empirical study is based on an unbalanced annual panel data covering various individual indicators, such as the size of informality, globalisation indicators and other macroeconomic indicators, for a selected sample of developing countries over the period 1990-2006 (see ANNEX 1 for a detailed description of the data and their sources).

Informality data

The data sets covering the size of informality in developing countries come from two principal sources. The first data set used in this paper is available in the ILO's Key Indicators of the Labour Market (KILM) database. As defined in the ILO report (1993) on "*Decent Work and the Informal Economy*", informal employment of a country comprises:

- own-account workers and employers who have their own informal sector enterprises;
- contributing family workers, irrespective of whether they work in formal or informal sector enterprises;
- employees who have informal jobs, whether employed by formal sector enterprises, informal sector enterprises, or as paid domestic workers by households;
- members of informal producers' cooperatives;
- persons engaged in the own-account production of goods for own final use by their household (e.g. subsistence farming, do-it-yourself construction of own dwellings).

The ILO informal employment database allows us to construct an unbalanced panel for 49 countries (17 Latin American countries; 11 Asian countries; 15 African countries; and 6 East and Central European countries) over the period 1990-2006. Here the size of informality is measured by the ratio of informal employment to total employment.

² "Birth-death MCMC sampler is the standard model sampler used in most BMA routines. One of the K potential covariates is randomly chosen; if the chosen covariate forms already part of the current model M_i , then the candidate model M_j will have the same set of covariates as M_i but for the chosen variable. If the chosen covariate is not contained in M_i , then the candidate model will contain all the variables from M_i plus the chosen covariate" (Zeugner, 2011).

³ Remind that N is number of observations and K is number of covariates.

The second data set used in this paper employs the shadow economy estimates of informal activity developed by Schneider (2005, 2007), which are derived from a combination of the Currency Demand Approach and the dynamic multiple-indicators multiple causes (DYMIMIC). Schneider (2005) provides a snapshot of informality for 110 countries in 1990/91, 1994/95 and 1999/2000, and then Schneider (2007) provides annual observations for the same countries over the period 2000-2004. However, in order to consist with the country sample in the first data set, we only use the shadow economy estimates by Schneider for 49 developing countries of interest. In this second data set, the size of informality is measured as the share of informal sector's GDP in official GDP.

Globalisation indicators

As mentioned, globalisation is manifold dimensions. In this paper, we pay our attention to two main dimensions of globalisation process, notably economic and non-economic dimensions. In order to capture the non-economic dimensions of globalisation, we use a number of different globalisation indicators developed by the Zurich-based Konjunkturforschungsstelle (KOF) (Dreher et al., 2008), including three social globalization indicators and one political globalization indicator (see Annex 2). Likewise, we classify economic dimension of globalization process into three sub-dimensions:

- *First*, to measure general degree of economic globalization of a given country we use the "Actual economic flows" indicator also provided by Dreher et al. (2008). This indicator presents a weighted average of trade, foreign direct investment, portfolio investment and income payments to foreign nationals.
- *Second*, the financial dimension of globalization is measured through two alternative indicators being considered as "de facto" one or "de jure" one. The "de facto" indicator is calculated as the ratio of foreign assets and liabilities to GDP.⁴ According to Baltagi et al. (2009), the advantage of this indicator is to provide a useful summary of the financial integration progress of a country. We also deploy the second "de facto" indicator, which is considered as the share of FDI inflows in GDP. Whereas the "de jure" indicator is the Chinn and Ito (2006) index of capital account openness (KAOPEN)⁵ that is widely used in previous cross-country studies.
- *Third*, to capture trade globalisation dimension we use a broad set of indicators. The first one is the trade openness indicator. Among others, the most well-known trade

⁴ This indicator is initially constructed by Lane and Milesi-Ferretti (2006).

⁵ Available at: <http://www.ssc.wisc.edu/~mchinn/research.html>.

openness indicator is the Sachs and Warner (1995) index.⁶ Although this index serves as a proxy for a wide range of policy and institutional differences and not only of trade policy (Rodriguez and Rodrik, 1999), it can only suggest that a country is either open or closed. Also, this index is difficultly constructed due to the unavailability of many data components. Furthermore, the statistical correlation between the SW index and other variables of interest is not always obvious and difficult to set an econometric model and to interpret the empirical results. For these reasons, we deploy a standard trade openness indicator measuring the sum of exports and imports to GDP. The second one is a set of *de jure* trade openness indicators including: *i)* Trade restrictions; *ii)* Most-favoured nation (MFN) rate; *iii)* Revenue from trade taxes. The last one comprises the diversification and concentration index of exports and imports.

Data setting

Remind that the main objective of this paper is not to identify the determinants of informal employment in developing countries, but is to clarify the multi-dimensions of globalisation process and their possible impacts on informality of developing countries by employing the BMA technique. Thus, the inclusion of other explanatory variables beside a large number of globalisation indicators is not strongly required. However, in order to compare our BMA results to those of previous studies (Bacchetta et al., 2009; Fugazza and Fiess, 2010), together with a data set only including globalisation indicators, we build another dataset that includes both globalisation indicators and other macroeconomic indicators collected from various international sources. The choice of these additional explanatory variables is strongly based on Bacchetta et al. (2009) and Fugazza and Fiess (2010) (see further Annex 2). Besides, to gain a better understanding of impacts of trade integration on informal employment, we construct another alternative dataset, in which we use two separate trade openness indicators, notably exports to GDP ratio and imports to GDP ration, instead of the ratio of total exports and imports to GDP. By and large, our empirical study is based on eight alternative data sets summarised in Table 1.

⁶ The SW index, which is constructed by Sachs and Warner (1995), is a dummy variable for openness based on five individual dummies for specific trade-related policies. Relying on this index, a country is classified as closed if it displays at least one of the following characteristics: Average tariff rates of 40 percent or more; Non-tariff barriers covering 40 percent or more of trade; A black market exchange rate that is depreciated by 20 percent or more relative to the official exchange rate, on average, during the 1970s or 1980s; A state monopoly on major exports; A socialist economic system.

Table 1: Data setting's outline

Datasets	Dependent variable	Independent variables
DS 1	(1)	(3)
DS 2	(1)	(4)
DS 3	(2)	(3)
DS 4	(2)	(4)
DS 5	(1)	(3) + (5)
DS 6	(1)	(4) + (5)
DS 7	(2)	(3) + (5)
DS 8	(2)	(4) + (5)

Notes: **(1)** The size of informal sector is measured by the ratio of informal employment to total employment; **(2)** The size of informal sector is measured as the ratio of informal GDP to "official" GDP; **(3)** A set of globalisation indicators, in which trade openness is measured by the share of total imports and exports to GDP; **(4)** A set of globalisation indicators, in which trade openness is measured by exports/GDP ratio and imports/GDP ratio; **(5)** A set of other explanatory variables;

4. Empirical analysis

This section reports and discusses the BMA results. It also reports the results of an alternative econometrical technique that check the sensitivity of the results to different estimation methods and different data settings. Finally, we compare our main findings with those of related literature.

Estimation results

The main results are presented in Tables 2-5. Tables 2 and 3 contain the BMA estimates of restrictive regressions that only include various globalisation indicators in the matrix X_i . Tables 4 and 5 report the BMA estimates of a larger regression comprising both globalisation indicators and other macroeconomic indicators. In addition, *Tables a* present the results using the data set with the sum of exports and imports to GDP as a *de facto* trade openness indicator, while *Tables b* report the results using the dataset with two separate ratios, exports/GDP and imports/GDP. In all reporting tables, the importance of the covariates in explaining the dependent variables is given in the first column PIP, which represents posterior inclusion probabilities calculated through the UIP procedure. The second column displays the coefficients averaged over all models, including the models wherein the variable was not contained (implying that the coefficient is zero in this case). The last column presents the robustness test from the FLS alternative prior specification.

Results for the ILO datasets

We begin with the empirical results using the ILO datasets (reported in Tables 2 and 4), in which the size of informal sector is measure as the ratio of informal employment to total

employment. Going straight to question of interest, we note that among 14 different globalization indicators, only 3 variables have inclusion probabilities above 50% in the levels specification, meaning a fairly parsimonious globalization dimensions are sufficient to explain the size of informal sector in developing countries. Globalization indicators have high inclusion probabilities are trade openness, KAOPEN index and personal contacts level. On the other hand, our empirical results show that broad categories of globalization dimensions have received little statistical support, particularly those related to the MFN rate, actual economic flows and foreign assets and liabilities. From these results, many comments may arise.

Firstly, trade openness measured as the share of total trade in GDP plays a significant role in explaining informal employment. We see that with 99.7%, virtually all of posterior informality model include trade openness indicator. Moreover, its unconditional coefficient is positive, meaning that more openness to trade generates larger size of informal employment. However, this coefficient is quite low (0.13). This suggests that trade openness plays a significant but not much important role in increasing the size of informal employment. In the case of separately using exports/GDP ratio and imports/GDP ratio as two *de facto* trade openness indicators, we find that the inclusion probability of imports is high (97.9%) while that of exports is fairly small (only 7.9%). This result indicates different impacts of imports and exports on informal employment: imports growth seems to be correlated with more informality while exports growth is not sufficient to explain informal employment rates. Unlike a high inclusion probability of *de facto* trade openness, all other trade indicators have such a small PIP value. It means that trade reforms in developing countries plays an insignificant role in explaining the evolution of their informal labour markets. In other words, even though trade reforms require an economic adjustment process and labour reallocation across sectors in developing countries, such reforms may only have impacts (positive or negative) on formal sector.

Secondly, we also obtain mixed results regarding the impacts of financial globalization on informal employment. With a high inclusion probability (58.7% and 59.5%), KAOPEN index, a *de jure* financial openness indicator, confirms its significant role in explaining the size of informal employment. Precisely, less restrictions on cross-border financial transactions lead to higher informal employment rates. Conversely, with small PIP values, both inward FDI flows and foreign assets seem to have no impact on informal employment. This result could be interpreted as evidence that most foreign investment inflows take place in the formal sector.

However, if we take into consideration the structuralism hypothesis of the informal economy serving the formal sector, foreign investment could still increase informality rates through an indirect channel: *foreign investment inflows encourage the development of formal sector, which in turn lead to an enlargement of informal sector to respond to larger demand of formal sector*. This hypothesis will be considered in our further research.

Looking now at non-economic dimensions of globalization, our empirical results also offer a mixed picture. On one hand, the covariate “personal contacts” has a most important PIP value (100%) and a negative coefficient. Accordingly, personal contacts level has helped to reduce informality in our sample developing countries. On the other hand, with small PIP values, other social and political globalization indicators do not seem to matter much. Consequently, their coefficients are quite low.

We now turn our attention to Tables 4a and 4b, in which the BMA estimates are reported for a larger dataset including both globalization covariates and other ones. Despite the different magnitudes of each PIP and PM value, the qualitative nature of our results regarding all globalisation covariates, by and large, remains unaltered in this dataset. Furthermore, we also find another set of 6 variables out of the 13 additional potential covariates that have high inclusion probabilities (above 50%). These 6 variables can be classified into two sub-categories: the size of economy (GDP and population indicators) and regulation (government spending, rule of law and government accountability). In terms of economic size, a high PIP value and a negative coefficient of GDP per capita indicates that an increase in economic development level can reduce informal employment in developing countries. In addition, population-related indicators, which have been used to control for the growth of labour force, have also a remarkable impact on informal employment. Regarding the regulatory environment, our empirical results nevertheless provide a more nuanced picture of the connection between government regulation and informal employment. With a high PIP and a negative coefficient, governance efficiency can be shown to contribute to a decline in informal rates. By contrast, regarding to the rule of law, we obtain such a “curious” result. A negative coefficient of this covariate implies that high informality rates seem to be an undesired outcome of the rule of law in our sample developing countries. Likewise, labour market regulation measured as decentralised wage bargaining systems and minimum wage levels has no effect on informal employment. It embodies that the adjustments in labour

market regulation have been not enough efficient to allow developing countries to increase formalisation rates and to decrease the size of informal sector.

Results for Schneider's shadow economy datasets

Now we look at the empirical results (reported in Tables 3 and 5) for Schneider's shadow economy datasets, in which the share of informal GDP in "official" GDP is used as a measure of the average size of shadow economy. Above all, using Schneider datasets our results provide only partial support to previous results using the ILO datasets, which stipulates that economic integration, here characterized by the KAOPEN and trade openness covariates, plays an important role in explaining the growth of shadow economy. In detail, posterior inclusion probabilities of KAOPEN index, trade openness indicator and imports ratio have a high value of 97.7%, 99.8% and 100% respectively. This is also true if economic integration is replaced by a more general concept "Actual economic flows" that represents a weighted average of trade, FDI, portfolio investment and income payments to foreign nationals. Moreover, the negative value of exports coefficient (-0.28 in Table 3b) indicates that exports growth can be shown to contribute to high formalization rates that in turn reduce shadow activities. Together with these similar results, we find a broad set of different results as using the shadow economy dataset.

Firstly, we obtain PIP value above 50% for two trade reforms indicators, notably trade restrictions and MFN rate. In addition, the coefficients of these covariates are negative, meaning that trade reforms seem to be correlated with less informality. At the same time, the relative concentration and diversification of merchandise exports have high PIP values. Moreover, these two covariates have a comparatively large coefficient and seem to be most important. However, they have quite different impacts on shadow economy. While trade concentration may reduce shadow activities, trade diversification may lead to the deepening of informal sector. Regarding other globalization indicators, the empirical results confirm the role of cultural proximity in reducing informality in developing countries.

Examining now the results relating to a larger shadow economy dataset in Tables 5a and 5b, we first note that the effects of economy size (characterized by GDP per capita and population-related indicators) appear to be either quantitatively or qualitatively similar to those obtained for the ILO dataset. Specifically, higher level of GDP per capita can lead to less informality while the population growth has been considered as a main source of informality in developing countries. Tables 5a and 5b also list other explanatory variables that

have high inclusion probabilities, including top marginal tax rates, transfers and subsidies, and corruption. Among others, high marginal tax rates for top earners seem to foster shadow economy. This result is in line with those presented in the literature assuming that high marginal tax rates would set incentives for high-skilled employees to become informal. Also, transfers and subsidies tend to contribute to the enlargement of shadow economy. By contrast, a negative coefficient of corruption index embodies that the growth of shadow economy probably results from a high level of corruption.

By and large, regarding a group of covariates including *de facto* trade openness, *de jure* financial openness, economy size-related indicators and government effectiveness, our empirical analysis gives similar findings from both ILO and Schneider datasets. Relating to other globalization indicators and a set of additional macroeconomic variables, our findings are different across these two datasets. Table 6 provides a comparison of main findings.

Our different empirical results are probably due to the differences in data setting. First, the ILO dataset is an unbalance panel data, which covers information on informal employment of 55 developing countries over the period 1990-2006. By contrast, the Schneider dataset is a balance panel data, which provides information on economic growth of shadow economy during three shorter periods 1990/91, 1994/95 and 1999-2004. Second, each dataset is based on a given definition of concept "informality" as well as a given method of estimating informality rates. The ILO database defines "*the informal sector as production units operated by single individuals or households that are not constituted as separate legal entities independent of their owners and in which capital accumulation and productivity are low*", while Schneider (2005) defines "*the shadow economy includes all market-based legal production of goods and services that are deliberately concealed from public authorities for the following reasons: (1) to avoid payment of income, value added or other taxes, (2) to avoid payment of social security contributions, (3) to avoid having to meet certain legal labor market standards, such as minimum wages, maximum working hours, safety standards, etc., and (4) to avoid complying with certain administrative procedures, such as completing statistical questionnaires or other administrative forms*". Likewise, to estimate the size of informality, the ILO dataset employs the direct methods⁷ while Schneider (2005, 2007) applies

⁷ The direct methods are microeconomic in nature and use either voluntary survey data or the results from tax audits to construct estimates of total economic activity and its official and unofficial (or measured and unmeasured) components.

an indirect and modeling approach, which is a comparison of the DYMIMIC model and currency-demand model.

Table 6: Main findings

Similar results				
Variables	(1)		(2)	
Trade openness	Yes		Positive	
KAOPEN index	Yes		Positive	
GDP per capita (level)	Yes		Negative	
Population growth	Yes		Positive	
Working-age population	Yes		Positive	
Government accountability	Yes		Negative	

Different results				
Variables	ILO dataset		Schneider dataset	
	(1)	(2)	(1)	(2)
Trade diversification	No	-		
Trade concentration	No	-		
Exports	No	-	Yes	Negative
Imports	Yes	Positive	Yes	Positive
Trade restrictions	No	-	Yes	Negative
Most-favoured nation (MFN) rate	No	-	Yes	Negative
Actual economic flows	No	-	Yes	Positive
Personal contacts	Yes	Negative	No	-
Cultural proximity	No	-	Yes	Negative
Top marginal tax rates	No	-	Yes	Positive
Transfers and subsidies	No	-	Yes	Positive
Rule of law	Yes	Positive	No	-
Corruption	No	-	Yes	Negative

Variables with low inclusion probabilities in both datasets: Most-favoured nation (MFN) rate; Trade restrictions; Revenue from trade taxes; FDI; Foreign assets and liabilities; Information flows; Political globalization; GDP growth; Government spending; Price controls; Minimum wages; and Wage bargaining.

Notes: (1) High inclusion probabilities; (2) Nature of connection with informality

Even though we find a broad set of different results from two datasets, these results are not contrary each other. Importantly, the results from Schneider's dataset seem to complement to those obtained from the ILO dataset, and vice versa.

Robustness checks

As revealed, the BMA estimates suggested by Fernandez et al. (2001a) are carried out to examine the sensitivity of our empirical results to alternative estimation strategies and datasets. Here we only report the PIP values estimated by the FLS method in the last column of Tables 2-5. Even though the magnitudes of each estimated PIP are different compared to those obtained from the IUP approach, the qualitative nature of our results remains unaltered

across different datasets. It supports that our empirical results are robust to a range of alternative measures, datasets and estimation approaches.

Comparisons with earlier studies

With the BMA results in hand, we now turn to address another question of how these results compare to those of previous studies. Reminder that empirically investigating the relationship between globalisation and informality for a large sample of developing countries has not received the attention. Up to now, there have been only two recent empirical works that tend to develop this issue, one by Bacchetta et al. (2009) and another by Fugazza and Fliess (2010). Our results, particularly those related to the ILO dataset can be compared with those of Bacchetta et al. (2009) who also use the same dataset. On the other hand, our results, which are obtained from Schneider's dataset, can be compared with those suggested by Fugazza and Fliess (2010). We list our main findings as well as those of two cited previous studies in Table 7.

The work developed by Bacchetta et al. (2009) aims to clarify such a complex interaction between globalisation and informal economy in developing countries. To do so, their empirical analysis is based on the ILO database providing information on informal employment for 31 developing countries. In addition, the multifaceted nature of globalisation process is measured not only by trade integration indicators but also by financial, social and political globalisation indicators. The authors also include in their estimation models a large number of other macroeconomic variables. In general, our present study uses the same indicators favoured in Bacchetta et al. (2009). However, our findings provide only partial support to those of Bacchetta et al. (2009). Specifically, our findings are consistent with those suggested by Bacchetta et al. (2009) on role of various variables (e.g. personal contacts, GDP per capita, working-age population, and government effectiveness) in explaining the size of informal employment in developing countries. Regarding trade openness, according to Bacchetta et al., this globalisation dimension seems to be correlated with less informality. In sharp contrast, we find that trade openness, particularly imports growth, can increase informality rates in developing countries. Here, our finding seems to be consistent with much of theoretical models supporting a positive connection between trade opening and informality. Also, differing from Bacchetta et al., who exclude *de jure* financial indicator in their regression estimations, our empirical results indicate that this indicator (KAOPEN index) having a high PIP and a positive coefficient can be shown to contribute to an increase in

informality rates. In other words, capital account openness or global capital mobility has become a determinant factor of the deepening of informal activities in developing countries. This finding seems to complement to the theoretical hypothesis of Marjit and Acharyya (2003), who argue that the impact of globalisation on informality depends on the degree of capital mobility between the formal and the informal economy. Looking at other covariates (e.g. trade reforms, political globalisation, government policies and regulation), Bacchettat et al. (2009) find that all these variables enter significantly in all estimations and consequently have positive or negative impact on informal employment. Yet, employing the BMA technique, we obtain a fairly small PIP value for each of these variables. It means that they should not be considered as appropriate covariates to include in a regression model explaining informal employment in developing countries.

In another related empirical work, Fugazza and Fliess (2010) only tend to assess the sign of the relationship between trade liberalisation and informality. In their study, three different datasets are used: ILO micro-founded data; ILO survey-based data; and Schneider estimates. Here, we only refer to the findings of Fugazza and Fliess (2010) that are obtained from two latter datasets. Also, it should be worth noting that in Fugazza and Fliess (2010) all datasets include information on informal sector for both developed and developing country. Additionally, with respect to their research objective, the authors only include a narrow set of trade globalisation indicators in their regression estimation, notably *de facto* trade openness indicator (trade/GDP) and trade restrictions. Above all, employing the generalised method of moments (GMM) of Arellano and Bond (1991) as well as the system-GMM of Blundell and Bond (1998), Fugazza and Fliess fail to identify the nature of the relationship between *de facto* trade liberalisation and informality in all estimations. Besides, using the same Schneider dataset, while our empirical results signify that informal output falls with more restricted trade, Fugazza and Fliess find contrasting evidence that more trade restrictions is always associated with a larger share of informal output in total GDP. Nevertheless, relating to variable “actual economic flows”, our empirical study offers a similar finding with that of Fugazza and Fliess. Specifically, higher economic flows generate higher share of informal output, while this variable has no significant impact on informal employment.

To this end, the differences in empirical findings between our study and the two previous works may be explained by the fact that the model specifications are not identical, the estimation procedures are not the same and the data setting and frequencies used for

estimation are different in each study. Furthermore, there is also little agreement on the nature of such a complex relationship between globalisation and informality in earlier theoretical studies. Therefore, little empirical consensus on this relationship has been ineluctable. Notwithstanding these important differences, it is always useful to carry out such a comparison that allows us to clarify the extent of our current contribution in the context of related literature.

5. Concluding remarks

This paper aims to identify, from a large set of potential covariates, a subset of globalisation indicators, which truly and significantly influence informality in developing countries. With respect to this aim, we take the Bayesian Model Averaging procedure that allows us to confront uncertainty regarding the potential set of covariates included in a given regression model. Also, this procedure allows us to find out the globalisation indicators with high inclusion probabilities in the “true” informality model. In this paper, we implement two alternative BMA procedures, one based on the “unit information prior” approach and another one developed by Fernandez et al. (2001a), for two different datasets. The first one measures the size of informality as the share of informal employment in total employment, which is collected from the ILO database. The second one deploys the estimates of Schneider (2005, 2007) on the share of informal GDP to official GDP. Both these datasets contain a broad set of 14 globalisation indicators and a group of 13 additional explanatory variables.

The results presented in this paper are robust to a range of alternative measures, datasets and estimation methods. They, by and large, provide some important systematic investigation of the appropriate determinants of informality in developing countries. *First*, our empirical results indicate that the globalisation covariates with consistently high inclusion probabilities are: *de facto* trade openness; *de jure* financial openness; trade diversification and concentration; economic flows; social globalisation (except information flows); trade restrictions and the MFN rate. Accordingly, these variables should be included in empirical work or for modelling informality. *Second*, this paper does not support the inclusion of a set of variables favoured in Bacchetta et al. (2009) (including FDI, foreign assets and liabilities, trade taxes and information flows) as well as other macroeconomic variables (e.g. government spending; price controls; minimum wages, and wage bargaining). Specifically, we find that the BMA estimates of these variables are not robust to considering them as appropriate covariates in informality model. *Third*, this paper embodies that the size of informal sector in

developing countries depends not only on some specific aspects of globalisation process but also on other macroeconomic aspects, such as the size of economy (GDP per capita), population-related factors, government policies and regulation. *Lastly*, our empirical findings also suggest that the nature (positive or negative, significant or insignificant) of the relationship between globalisation process and informality strongly depends on informality measures. This last finding may raise an important question about the compatibility between different informality datasets, particularly between the ILO dataset and Schneider's dataset used in this current contribution.

To this end, the empirical evidence presented in this paper offer a mixed picture of globalisation's role in explaining informal activities in developing countries. In this case, formal economic modelling should be called for to deepen our understandings of the real impacts of each globalisation dimension on informality, which are important implications for the design of labour markets' regulation in developing countries.

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Table 2.a: BMA Results for DS1 (ILO database)

Variables	Zellner's g (UIP)		FLS
	PIP	PM	PIP
Trade diversification	14.1	1.47	12.2
Trade concentration	9.7	0.68	4.8
Trade openness	99.7	0.13	99.7
Trade restrictions	34.2	-0.06	45.0
Most-favoured nation (MFN) rate	6.4	0.25	3.5
Revenue from trade taxes	20.0	-0.20	19.6
Foreign direct investment	14.1	0.06	13.7
Foreign assets and liabilities	6.8	0.26	9.6
KAOPEN	58.7	0.96	63.2
Actual economic flows	6.3	-0.00	7.2
Personal contacts	100	-0.94	100
Information flows	13.4	-0.20	11.9
Cultural proximity	7.1	0.00	2.9
Political Globalization	10.5	-0.01	6.4

Notes: Posterior Inclusion Probabilities (PIP); Posterior Median (PM)

Table 2.b: BMA Results for DS2 (ILO database)

Variables	Zellner's g (UIP)		FLS
	PIP	PM	PIP
Trade diversification	10.4	0.89	13.1
Trade concentration	6.4	0.41	15.4
Exports	7.9	0.00	11.2
Imports	97.9	0.27	99.2
Trade restrictions	41.1	-0.07	30.8
Most-favoured nation (MFN) rate	7.7	0.00	8.9
Revenue from trade taxes	16.1	-0.16	18.4
Foreign direct investment	10.0	0.05	8.5
Foreign assets and liabilities	10.0	4.7	14.0
KAOPEN	59.5	0.99	57.9
Actual economic flows	5.8	0.00	7.2
Personal contacts	100	-0.95	100
Information flows	13.1	-0.00	9.3
Cultural proximity	5.9	0.00	8.6
Political Globalization	4.2	-0.00	8.8

Notes: Posterior Inclusion Probabilities (PIP); Posterior Median PM

Table 3.a: BMA Results for DS3 (Shadow economy database)

Variables	Zellner's g (UIP)		FLS
	PIP	PM	PIP
Trade diversification	93.62	22.41	95.6
Trade concentration	26.2	-2.58	31.8
Trade openness	99.8	0.02	44.3
Trade restrictions	59.5	-0.10	63.4
Most-favoured nation (MFN) rate	66.8	-0.13	71.5
Revenue from trade taxes	9.7	4.8	8.3
Foreign direct investment	14.5	0.06	14.4
Foreign assets and liabilities	6.0	-0.02	4.0
KAOPEN	97.69	1.93	99.2
Actual economic flows	77.0	0.11	71.5
Personal contacts	11.5	-0.01	7.7
Information flows	7.6	0.00	6.3
Cultural proximity	41.7	-0.04	38.7
Political Globalization	7.3	-0.00	7.0

Notes: Posterior Inclusion Probabilities (PIP); Posterior Median (PM)

Table 3.b: BMA Results for DS4 (Shadow economy database)

Variables	Zellner's g (UIP)		FLS
	PIP	PM	PIP
Trade diversification	65.5	14.15	72.4
Trade concentration	0.37	-4.47	47.1
Exports	93.4	-0.28	93.5
Imports	100	0.35	99.9
Trade restrictions	51.0	-0.08	56.4
Most-favoured nation (MFN) rate	50.6	-0.08	50.5
Revenue from trade taxes	4.9	0.01	13.5
Foreign direct investment	8.4	0.02	7.9
Foreign assets and liabilities	3.4	0.01	6.0
KAOPEN	97.7	1.79	95.9
Actual economic flows	94.7	0.17	94.6
Personal contacts	19.9	-0.02	19.0
Information flows	15.9	0.01	13.1
Cultural proximity	70.5	-0.09	71.8
Political Globalization	7.2	0.00	5.8

Notes: Posterior Inclusion Probabilities (PIP); Posterior Median (PM)

Table 4.a: BMA Results for DS5 (ILO database)

Variables	Zellner's g (UIP)		FLS
	PIP	PM	PIP
Trade diversification	15.6	1.95	2.9
Trade concentration	22.9	-2.32	8.6
Trade openness	66.2	5.31	51.6
Trade restrictions	19.7	-2.67	11.6
Most-favoured nation (MFN) rate	9.1	-1.58	6.6
Revenue from trade taxes	6.2	-3.29	4.1
Foreign direct investment	4.1	1.41	2.7
Foreign assets and liabilities	19.7	1.62	12.9
KAOPEN	68.0	1.16	56.9
Actual economic flows	24.4	3.79	38.3
Personal contacts	100	-5.70	100
Information flows	9.4	-8.43	6.9
Cultural proximity	7.5	1.99	9.3
Political Globalization	11.4	7.41	8.7
GDP per capita (level)	91.5	-1.29	74.6
GDP growth	1.8	5.53	4.8
Population growth	74.2	2.52	69.9
Working-age population	100	5.60	100
Government spending	9.3	2.77	4.6
Top marginal tax rates	4.1	9.38	5.5
Transfers and subsidies	19.4	2.34	7.1
Rule of law	100	1.12	100
Corruption	8.7	-1.45	4.2
Government accountability	100	-1.63	100
Price controls	18.5	1.28	10.3
Minimum wages	13.6	-7.13	13.0
Wage bargaining	10.7	-4.71	5.4

Notes: Posterior Inclusion Probabilities (PIP); Posterior Median (PM)

Table 4.b: BMA Results for DS6

Variables	Zellner's g (UIP)		FLS
	PIP	PM	PIP
Trade diversification	10.4	3.02	6.8
Trade concentration	15.3	-1.58	10.1
Exports	12.1	7.10	12.0
Imports	58.9	5.75	54.7
Trade restrictions	21.6	-3.14	10.0
Most-favoured nation (MFN) rate	14.8	-2.38	3.8
Revenue from trade taxes	6.2	-2.29	5.9
Foreign direct investment	9.6	3.33	5.4
Foreign assets and liabilities	24.2	2.13	19.5
KAOPEN	65.7	1.09	68.1
Actual economic flows	42.0	6.69	34.2
Personal contacts	100	-5.65	100
Information flows	10.5	-7.73	6.2
Cultural proximity	14.2	-6.46	1.3
Political Globalization	10.9	5.19	9.8
GDP per capita (level)	84.8	-1.14	81.3
GDP growth	4.7	1.88	1.0
Population growth	76.9	2.51	72.2
Working-age population	100	5.81	100
Government spending	8.0	1.85	4.1
Top marginal tax rates	13.7	1.43	0.4
Transfers and subsidies	9.6	8.72	5.5
Rule of law	100	1.12	100
Corruption	5.6	-1.20	4.6
Government accountability	100	-1.64	100
Price controls	21.8	1.27	3.9
Minimum wages	14.5	-7.77	8.1
Wage bargaining	1.5	-6.14	1.0

Notes: Posterior Inclusion Probabilities (PIP); Posterior Median (PM)

Table 5.a: BMA Results for DS7 (Shadow economy database)

Variables	Zellner's g (UIP)		FLS
	PIP	PM	PIP
Trade diversification	71.5	1.59	7.1
Trade concentration	72.3	-8.57	58.5
Trade openness	70.0	-1.74	32.8
Trade restrictions	6.9	6.84	0.0
Most-favoured nation (MFN) rate	29.9	-2.66	10.8
Revenue from trade taxes	2.2	6.28	0.2
Foreign direct investment	85.1	6.24	75.3
Foreign assets and liabilities	7.1	-1.21	0.1
KAOPEN	10.4	6.70	15.4
Actual economic flows	100	2.34	100
Personal contacts	4.1	-2.03	3.4
Information flows	4.9	2.84	5.1
Cultural proximity	74.7	-9.31	70.3
Political Globalization	11.9	7.02	8.5
GDP per capita (level)	100	-1.76	100
GDP growth	7.0	-6.85	4.0
Population growth	98.0	2.88	99.8
Working-age population	100	4.09	100
Government spending	3.3	9.97	2.6
Top marginal tax rates	81.8	5.91	75.9
Transfers and subsidies	54.4	5.04	48.5
Rule of law	15.3	-4.63	15.2
Corruption	50.1	-2.45	61.2
Government accountability	62.2	-3.78	52.7
Price controls	28.5	1.59	35.4
Minimum wages	8.6	2.01	5.1
Wage bargaining	13.3	1.01	12.1

Notes: Posterior Inclusion Probabilities (PIP); Posterior Median (PM)

Table 5.a: BMA Results for DS8 (Shadow economy database)

Variables	Zellner's g (UIP)		FLS
	PIP	PM	PIP
Trade diversification	76.0	3.60	23.7
Trade concentration	77.9	-9.43	63.6
Exports	71.0	-5.46	40.9
Imports	75.2	2.02	20.1
Trade restrictions	8.5	1.41	1.3
Most-favoured nation (MFN) rate	21.7	-1.95	9.3
Revenue from trade taxes	5.8	5.82	3.3
Foreign direct investment	79.0	5.68	61.3
Foreign assets and liabilities	11.2	-1.70	1.9
KAOPEN	5.8	3.63	8.4
Actual economic flows	100	2.41	100
Personal contacts	7.5	-1.43	4.9
Information flows	4.9	6.27	2.9
Cultural proximity	77.7	-9.43	60.1
Political Globalization	8.2	4.03	5.1
GDP per capita (level)	100	-1.73	100
GDP growth	8.8	-1.06	3.3
Population growth	99.9	2.74	97.0
Working-age population	100	4.06	100
Government spending	5.7	-3.15	9.2
Top marginal tax rates	79.8	5.70	68.4
Transfers and subsidies	54.1	4.74	45.1
Rule of law	17.4	-4.93	8.3
Corruption	74.1	-3.43	72.2
Government accountability	48.6	-2.55	38.3
Price controls	27.7	1.44	23.0
Minimum wages	17.1	2.36	5.4
Wage bargaining	16.5	1.32	22.9

Notes: Posterior Inclusion Probabilities (PIP); Posterior Median (PM)

Table 7: Empirical results' comparison

Explanatory Variables	ILO dataset			Schneider dataset	
	Present study	Bacchetta et al. (2009)	Fugazza & Fiess (2010)	Present study	Fugazza & Fiess (2010)
<u>Globalisation covariates</u>					
Trade diversification	S	x	x	h/(+)	x
Trade concentration	S	y/(+)	x	h/(-)	x
Trade openness	h/(+)	y/(-)	y/(non)	h/(+)	y/(non)
Exports	S	x	x	h/(+)	x
Imports	h/(+)	x	x	h/(-)	x
Trade restrictions	S	y/(-)	y/(-)	h/(-)	y/(+)
Most-favoured nation rate	S	y/(-)	x	h/(-)	x
Revenue from trade taxes	S	y/(+)	x	s	x
Foreign direct investment	S	y/(+)	x	s	x
Foreign assets & liabilities	S	x	x	s	x
KAOPEN	h/(+)	x	x	h/(+)	x
Actual economic flows	S	y/(-)	y/(non)	h/(+)	y/(+)
Personal contacts	h/(-)	y/(-)	x	h/(-)	x
Information flows	S	y/(-)	x	s	x
Cultural proximity	S	x	x	h/(-)	x
Political Globalization	S	x	x	s	x
<u>Other covariables</u>					
GDP per capita (level)	h/(+)	y/(-)	y/(-)	h/(+)	y/(-)
GDP growth	S	y/(-)	x	s	x
Population growth	h/(+)	y/(-)	x	h/(+)	x
Working-age population	h/(+)	y/(+)	x	h/(+)	x
Government spending	S	y/(-)	x	s	x
Top marginal tax rates	S	y/(-)	x	h/(+)	x
Transfers and subsidies	S	y/(-)	x	h/(+)	x
Rule of law	h/(+)	y/(-)	x	s	x
Corruption	S	y/(+)	y/(-)	h/(-)	y/(-)
Government accountability	h/(-)	y/(-)	x	h/(-)	x
Price controls	S	y/(+)	x	s	x
Minimum wages	S	y/(-)	x	s	x
Wage bargaining	S	y/(+)	x	s	x

Notes: **h**: high inclusion probabilities; **s**: small inclusion probabilities; **y**: variable is included in the regression model; **x**: variable is not included in the regression model; **(+)**: positive relationship; **(-)**: negative relationship; **(non)**: insignificant relationship.

ANNEX 1: Country sample

Country	Time period	Source	Definition	Coverage
LATIN AMERICA (17)				
Argentina	1996-2006	Roca et al. (2006)	Harmonized non-registered employment	Urban
Bolivia	1990-2005	Key Indicators of the Labour Market database ILO Regional Database for Latin America and the Caribbean	All own-account workers (excluding professionals and technicians) and unpaid family workers, and employers and employees working in establishments with less than 10 persons engaged, depending on the available information.	National
Barbados	1998	Key Indicators of the Labour Market database ILO Regional Database for Latin America and the Caribbean	Informal sector: Unincorporated or unregistered enterprises, without a complete set of accounts, and with less than 6 persons engaged.	National
Brazil	1990, 1992-2006	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	National
Chile	1990-2006	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	National
Colombia	1992-2004	Bustamante (2006)	Workers in small businesses with less than 10 employees, family workers without remuneration, domestic employees, self-employed workers except independent professionals.	Seven main cities
Costa Rica	1990-2006	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	National
Dominican Republic	2000-2004	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	All own-account workers (excluding professionals and technicians) and unpaid family workers, and employers and employees working in establishments with less than 5 or 10 persons engaged, depending on the available information.	National
Ecuador	1994-2006	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	Urban
Honduras	1990-2006	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	National
Mexico	1991-2006	Rani (2008) ; Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	National

Nicaragua	2000	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	All own-account workers (excluding professionals), employers, employees and unpaid family workers working in enterprises with less than 5 persons engaged. Public sector excluded.	8 main cities
Paraguay	1995-2006	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	National
Panama	1991-2006	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), Contributing family workers (unpaid) and domestic workers.	National
Peru	1985-1997	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Persons working in private enterprises with 4 or less persons engaged in all branches of economic activity other than manufacturing, and with 9 or less persons engaged in manufacturing.	Urban
Uruguay	1996-2005	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	National
Venezuela	1994-2006	Rani (2008) calculations based on data processed by the ILO's Information System and Labour Analysis (SIAL) in Panama city	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers.	National

ASIA (11)

Bangladesh	1993	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Production of goods or services which is undertaken by one or more household members in their place of residence, without any commercial name-plate, for the purpose of generating income.	National
China	1990, 2003	Ghose et al. (2008)	Informal employment is taken as the difference between formal employment and total employment. Formal employment includes employment in state-owned enterprises, in collectively owned enterprises and in private large-scale enterprises.	National
India	1993-4, 1999-2000, 2003-2004	Rani (2008)	Informal sector includes enterprises with less than 5 workers, entrepreneurs, own-account workers or self-employed (excluding professional, managerial and technical workers), contributing family workers (unpaid) and domestic workers	National
Indonesia	1990-2003	Sakernas as cited by (Sugiyarto et al., 2006)	Sakernas National Labour Force Survey Indonesia classifies workers employment status into employers, employees, self-employed and unpaid family workers. In Indonesia informality is defined as self employed and unpaid family workers.	Urban
Iran	1996	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Production of goods or services which is undertaken by one or more household members in their place of residence, without any commercial name-plate, for the purpose of generating income.	National
Kazakhstan	1994-1995	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: Persons engaged in economic activities undertaken without official statistical registration.	National

Kyrgyzstan	1994	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: Self-employed persons not registered with the tax authorities and the persons working for them.	National
Philippines	1995	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector: Unincorporated private enterprises with less than 10 persons engaged in the (main) activity/establishment and without a complete set of accounts. Domestic workers are included if they consider themselves as self-employed business operators. Enterprises not employing any employee are excluded if they engage in contract work for other enterprises on the basis of contracts made as individuals remunerated at time rates or receiving specific employee benefits.	National
Pakistan	1992, 1997, 2000	Key Indicators of the Labour Market database	Households in unincorporated enterprises owned by own-account workers; households in unincorporated enterprises owned by employers with less than 10 persons.	Urban
Sri Lanka	1990, 1994, 2000-2003	Ghose et al. (2008)	Employment in the formal segment is taken as the sum of wage employment in the public sector and wage employment in those private sector establishments that employ at least 10 wage workers. Informal employment is taken as the difference between formal employment and total employment.	National
Thailand	1990, 1992, 1994, 2001, 2002	Rani (2008) ; Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal employment is taken as the difference between formal employment and total employment. Formal: enterprises with more than 5 workers and those in public administration.	National

AFRICA (15)

Benin	1992, 1999	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector: Included are all economic units which are mobile or which have a semi-fixed location, plus those economic units with a fixed location which have one or more of the following characteristics: lack of formal accounts, non-inclusion in the register of commerce, or non-registration with the OBSS (Office Béninois de Sécurité Sociale).	National
Botswana	1994, 1996, 2001	Rani (2008)	Informal employment is taken as the difference between formal employment and total employment. Formal enterprises include those with more than five workers and those in public administration.	National
Cameroun	1993, 2005	Enquête sur l'emploi informel au Cameroun (EESI), Institut National de la Statistique	Informality is measured on the basis of unregistered production units and those that do not keep books formally.	National
Côte d'Ivoire	1996	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Employers, own-account workers, unpaid family workers, apprentices and unskilled labourers working in enterprises owned by themselves or by a member of their household or family.	Urban
Egypt	1988, 1998, 2003, 2006	El Mahdi and Rashed (2007) ; Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Private non-agricultural wage-workers as a share of non-agricultural wage workers.	National
Kenya	1992-1994, 1997, 1999	Standardized Survey Bulletin based on surveys conducted by African National Statistical Offices. World Bank African Region ; Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: own account workers, unpaid family workers who work for at least 7 hours per day, and employers and employees in small establishments (less than 5 workers).	National
Madagascar	1995	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector: Production units without statistical number and/or without formal written accounts.	National

Mali	1989, 1996, 2004	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: Own-account workers excluding professionals; employers with less than 10 employees excluding professionals; employees (including apprentices) and unpaid family workers working in unincorporated enterprises.	National
Mauritius	1992, 1997	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Small or micro-enterprises: Establishments with less than 10 persons engaged and units operating without fixed location.	National
Malawi	1998	Standardized Survey Bulletin based on surveys conducted by African National Statistical Offices. World Bank African Region	Informal sector employment: own account workers, unpaid family workers who work for at least 7 hours per day, and employers and employees in small establishments (less than 5 workers)	National
South Africa	1995-2006	Essop and Yu (2008)	October Household Survey, conducted annually by SSA from 1994 to 1999, consisting of those businesses that are unregistered and do not have a value added tax (VAT) number. A cautionary note is that the 1995 OHS did not ask respondents whether their employers were registered, thus undercounting the informal sector in this year. From 1999 on, self-declaration was given preference over VAT registration as the defining characteristic, i.e. respondents were specifically asked if they considered themselves part of the informal economy.	National
Tunisia	1997	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Small or micro-enterprises: Enterprises with less than 10 persons permanently engaged in the enterprise. Activities without fixed location are not covered.	National
Uganda	1993	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Persons working in household enterprises or in establishments with less than 5 persons engaged.	National
Zambia	1998	Standardized survey bulletin based on surveys conducted by African National Statistical Offices. World Bank African Region	Informal sector employment includes own-account workers, unpaid family workers who work for at least seven hours per day, and employers and employees in small establishments (less than five workers)	National
Zimbabwe	1990, 2002	Rani (2008)	Informal employment is taken as the difference between formal employment and total employment. Formal enterprises are those with more than 5 workers and those in public administration.	National

CENTRAL & EST EUROPE (6)

Georgia	1998	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: Self-employed persons (employers and own-account workers) working in unregistered or seasonal enterprises as their main or secondary activity, excluding agriculture in rural	National
Latvia	1996, 1999	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: Persons working in active enterprises which are not included in the business register and not covered by statistical surveys.	National
Lithuania	1998-2000	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector: Small private non-registered enterprises, which have employees working without legal contracts, do not pay contributions to social security funds, and sell miscellaneous goods and services in the market.	National
Moldova	2004	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: Persons working in active enterprises which are not included in the business register and not covered by statistical surveys.	National
Slovakia	1994-1999	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: Individual entrepreneurs (small tradesmen) not registered in the Business Register, and the persons employed by them.	National

Turkey	2000	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector: All non-agricultural economic units which are unincorporated (establishments whose legal position is individual ownership or simple partnership), pay lump sum tax or no tax at all and work with less than 10 persons engaged. The figures cover only own-account workers and employers.	National
Ukraine	1997, 2003, 2004	Key Indicators of the Labour Market database. ILO Regional Database for Latin America and the Caribbean	Informal sector employment: Employees hired on the basis of a gentlemen's agreement and employed by individuals or private unincorporated enterprises with less than 6 persons engaged; employers and own-account workers (self-employed persons) operating unregistered enterprises; unpaid family workers working for individuals or private unincorporated enterprises with less than 6 persons engaged; and persons exclusively engaged in individual subsidiary agriculture who sell all or part of their produce.	National

ANNEX 2: Globalisation indicators

	Variables	Definition	Source
Size of informality	ILO employment	Informal employment to total employment ratio (%)	ILO's Key Indicators of the Labour Market (KILM)
	Shadow economy estimates	Informal sector's GDP to official GDP ratio	Schneider (2005, 2007)
Trade globalization indicators	Trade diversification	Diversification index of exports and imports of countries and country groups	UNCTAD, 2011
	Trade concentration	Concentration index of exports and imports of countries and country groups	UNCTAD, 2011
	Exports	Share of exports in GDP	WDI, 2011
	Imports	Share of imports in GDP	WDI, 2011
	Trade openness	Sum of exports and imports as a share of GDP	WDI, 2011
	Trade restrictions	Index (0-100) representing a weighted average of hidden import barriers, mean tariff rate, taxes on international trade and capital account restrictions. The indicator moves from most to least restrictive.	Dreher et al. (2008)
	Most-favoured nation (MFN) rate	Applied MFN average duty	UNTAD, 2011
Revenue from trade taxes	Index based on amount of taxes on international trade as a share of exports and imports. The formula used to calculate the ratings for this sub-component was: $(V_{max} - V_i) / (V_{max} - V_{min})$ multiplied by 10. V_i represents the revenue derived from taxes on international trade as a share of the trade sector. The values for V_{min} and V_{max} were set at zero and 15%, respectively	Fraser Institute, Economic Freedom of the World (2008)	

Financial globalization indicators	Foreign direct investment	Inflows of foreign direct investment as a share of GDP	WDI, 2011
	Foreign assets and liabilities	Foreign assets and liabilities as a share of GDP	WDI, 2011
	De jure financial openness (KAOPEN index)	The Chinn and Ito (2006) index of capital account openness is constructed from four binary dummy variables codifying restriction on cross-border financial transactions that are reported in the IMF's Annual Report on Exchange Arrangements and Exchange Restrictions	http://www.ssc.wisc.edu/~mchinn/research.html
	Actual economic flows	Index (0-100) representing a weighted average of trade, foreign direct investment, portfolio investment and income payments to foreign nationals	Dreher et al. (2008)
Social globalisation indicators	Personal contacts	Index (0-100) of a weighted average of the indicators on "outgoing telephone traffic", "transfers (as a percentage of GDP)", "international tourism", "foreign population (as a percentage of total population)" and "international letters (per capita)"	Dreher et al. (2008)
	Information flows	Index (0-100) of a weighted average of the indicators on "internet hosts (per 1000 people)", "internet users (per 1000 people)", "cable television (per 1000 people)", "trade in newspapers (percent of GDP)" and "radios (per 1000 people)"	Dreher et al. (2008)
	Cultural proximity	Index (0-100) of a weighted average of the indicators on «number of McDonald's restaurants (per capita) », « number of Ikea (per capita) », and « trade in books (percent of GDP)	Dreher et al. (2008)
	Political Globalization	Index (0-100) of a weighted average of the indicators on « Embassies in Country », « Membership in International Organizations », « Participation in U.N. Security Council Missions », and « International Treaties »	Dreher et al. (2008)
Other explanatory variables			
	GDP per capita (level)	Gross Domestic Product per capita in constant US\$ 2000 prices	WDI, 2011
	GDP growth	Annual real GDP growth rate	WDI, 2011
	Population growth	Annual rate of change in population size	WDI, 2011
	Working-age population	Working-age population (15-64 years) in percentage of total population	WDI, 2011
	Price controls	Index based on the extent to which price setting in sectors is subject to controls/marketing boards	Fraser Institute, Economic Freedom of the World (2008)

Government spending	Index based on general government consumption spending as a percentage of total consumption. The rating for this component is equal to: $(V_{max} - V_i) / (V_{max} - V_{min})$ multiplied by 10. The V_i is the country's actual government consumption as a proportion of total consumption, while the V_{max} and V_{min} were set at 40 and 6, respectively	Fraser Institute, Economic Freedom of the World (2008)
Top marginal tax rates	Index based on marginal income tax rates at the highest income bracket. Countries with higher marginal tax rates that take effect at lower income thresholds received higher ratings	Fraser Institute, Economic Freedom of the World (2008)
Transfers and subsidies	Index based on general government transfers and subsidies as a share of GDP. The rating for this index is equal to: $(V_{max} - V_i) / (V_{max} - V_{min})$ multiplied by 10. The V_i is the country's ratio of transfers and subsidies to GDP, while the V_{max} and V_{min} values are set at 37.2 and 0.5, respectively	Fraser Institute, Economic Freedom of the World (2008)
Rule of law	Index (-2.5 to 2.5) capturing perceptions of the extent to which agents have confidence in and abide by the rules of society, and in particular the quality of contract enforcement, property rights, the police, and the courts, as well as the likelihood of crime and violence	Kaufmann et al. (2009)
Corruption	Index (-2.5 to 2.5) capturing perceptions of the extent to which public power is exercised for private gain, including both petty and grand forms of corruption, as well as state capture by private interests	Kaufmann et al. (2009)
Government accountability	Index (-2.5 to 2.5) capturing perceptions of the extent to which a country's citizens are able to participate in selecting their government, as well as freedom of expression, freedom of association, and a free media	Kaufmann et al. (2009)
Minimum wages	Index based on the ratio of mandated minimum wages to average value added per worker. Countries with higher mandated minimum wages relative to average value added per worker are given lower ratings. The formula used to calculate the zero-to-10 ratings for this subcomponent was: $(V_{max} - V_i) / (V_{max} - V_{min})$ multiplied by 10. V_i represents the ratio between minimum wage and average value added per worker. The values for V_{max} and V_{min} were set at 79% (1.5 standard deviations above average) and 0%, respectively	Fraser Institute, Economic Freedom of the World (2008)
Centralization of wage bargaining	Index based on responses to the Global Competitiveness Report's question: "Wages in your country are set by a centralized bargaining process (= 1) or up to each individual company (= 7)"	World Economic Forum, Global Competitiveness Report and Fraser Institute, Economic Freedom of the World (2008)
