



Monday 23 - Friday 27 August

Statistisk sentralbyrå Statistics Norway

#### **Financial Openness and Inequality**

Tsvetana Spasova

(University of Basel)

Stefan Avdjiev

(Bank for International Settlements)

Paper prepared for the 36th IARIW Virtual General Conference August 23-27, 2021 Session 6: Globalization and Inequality II Time: Tuesday, August 24, 2021 [16:30-18:30 CEST]

## Financial Openness and Inequality<sup>\*</sup>

Tsvetana Spasova<sup> $\dagger$ </sup> and Stefan Avdjiev <sup> $\ddagger$ </sup>

Please do not cite or circulate without author's permission

#### Abstract

Although there is a large body of literature on the link between trade openness and inequality, there is little empirical research on the relationship between external financial openness and inequality. In this paper, we attempt to fill this gap in the literature by conducting an empirical examination of the link between Ginibased inequality measures and key measures of financial openness for a sample of 47 countries between 1991 and 2013. We find that in emerging market economies the impact of external financial openness on inequality varies considerably over time. An increase in a country's external liabilities is associated with an initial rise and a subsequent fall in inequality. The main driver of this pattern appears to be foreign direct investment, whose estimated impact exhibits a very similar dynamics, in line with existing theoretical predictions. Meanwhile, the estimated impact of external financial openness on inequality in advanced economies tends to be smaller than in emerging market economies.

**JEL Codes**: F30, F40, O11

**Keywords:** Financial openness; Gini-based inequality measures; Foreign direct investments; External liabilities

#### **Highlights:**

- We find that in emerging market economies the impact of external financial openness on inequality varies considerably over time.
- An increase in a country's external liabilities is associated with an initial rise and a subsequent fall in inequality.
- The main driver of this pattern appears to be foreign direct investment, whose estimated impact exhibits a very similar dynamics.
- The estimated impact of external financial openness on inequality in advanced economies tends to be smaller than in emerging market economies.

<sup>†</sup>University of Basel, e-mail: tsvetana.spasova@gmail.com. Corresponding author.

<sup>\*</sup>The views expressed in this paper are those of the authors and do not necessarily reflect the views of the Bank for International Settlements. We would like to thank Bat-el Berger, Zuzana Filková, Pablo Garcia-Luna, Branimir Gruić, Enisse Kharroubi, Cathérine Koch, Marco Jacopo Lombardi, Giovanni Lombardo, Hyun Song Shin, Goetz von Peter, Philip Wooldridge, and seminar participants at the Bank for International Settlements.

<sup>&</sup>lt;sup>‡</sup>Bank for International Settlements (BIS), e-mail: stefan.avdjiev@bis.org.

## 1 Introduction

The increase in within country inequality that has taken place over the past few decades in large parts of the world has placed questions related to income distribution in the limelight of a public debate (Atkinson (2015)). As a consequence, studying inequality has moved to the forefront of both policy-makers' lists of priorities and researchers' agendas. In particular, there is great interest in gaining better understanding of the main drivers of inequality and the key determinants of income distributions in both, advanced economies (AEs) and emerging market economies (EMEs). The existing body of research on the topic suggests that inequality has both domestic and international determinants. Among the key domestic determinants of income inequality are education, redistributive tax-andtransfer policy, reduced role of trade unions and changing pay norms (Atkinson (2015)). Besides the domestic drivers of inequality, a very important role is also played by external factors such as globalization, international trade and financial openness.

According to most of the existing literature, the main drivers of the recent increases in inequality have been on the domestic side. It is true that there is some evidence that globalization appears to have had somewhat larger effect in the increase in income inequality in low-income countries (Cabral et al. (2016), Figini and Görg (2011), International Monetary Fund (2007) and Jaumotte et al. (2013), Denk and Cournède (2015). Nevertheless, technology appears to have been the most important factor in most countries. The returns to skilled labour, which uses technology more intensely, have increased considerably over the past couple of decades (Bank for International Settlements (2017a), Jaumotte et al. (2013), International Monetary Fund (2007), Milanović (2005)).

There is a large body literature that has examined the link between trade (real) openness and inequality. The evidence discovered by this literature has been mixed. On the one hand, trade between advanced economies and EMEs tends to increase the income of advanced economy skilled labour relative to that of advanced economy unskilled labour. On the other hand, the opposite tends to be true in the case of EMEs. That said, trade also leads to relative price falls for goods disproportionately consumed by lowerincome households, boosting their relative purchasing power (Fajgelbaum and Khandelwal (2016)). Given these offsetting effects, the net effect on inequality from trade openness is uncertain (Bank for International Settlements (2017a)).

In contrast, there is very little research on the link between financial openness and inequality. Furthermore, the empirical portion of that strand of literature is even more limited. The few existing papers have focused exclusively on Foreign Direct Investments (FDI) as a measure of external financial openness.

In this paper, we attempt to fill that gap in the existing literature by investigating the relationship between Gini-based inequality measures and various measures of external financial openness for a sample of 47 countries between 1991 and 2013.

Financial openness could increase income inequality through several channels. Similarly to trade openness, financial openness can rise inequality by boosting income from capital sources. The international mobility of goods and capital, relative to labour, can reduce the price of labour and therefore wages. This could constrain taxing capital and may lead to higher taxes on labour income (Autor et al. (2017)), leading to an increase of inequality as lower-income individuals rely mainly on labour income. Furthermore, financial openness, and in particular FDI, could increase capital intensity and the returns to skill, which could be beneficial to higher-income individuals. Last but not least, if the domestic institutions are not strong enough to prevent groups with special interests from acquiring the gains, financial openness could lead to an increase in income inequality (Claessens and Perotti (2007)).

On the other hand, there are also several channels through which financial openness could decrease inequality. First, an inflow of FDI into a given country is likely to spur a broad increase in productivity in that country, as the newly imported technologies spread through the recipient economy in a diffusion-like process. As that happens, the share of the population employed in the high-skilled industries is likely to increase (Aghion and Howitt (1998)), which would ultimately lead to a decline in inequality. Second, greater external financial openness can increase the ability of low-income individuals to borrow, which can, in turn, increase their income-generating opportunities (Beck et al. (2007), Ben Naceur and Zhang (2016), Beck et al. (2010)). Finally, greater inflows of capital into a given country would, all else the same, tend to increase the value of that country's currency, which could in turn improve the creditworthiness of currency-mismatched borrowers in that country. Since low-income individuals tend to own fewer foreign-currency assets, and are therefore more likely to have currency mismatched balance sheets, a capital flowinduced domestic currency appreciation is likely to reduce inequality.

Our results reveal that the impact of increasing external financial openness on inequality in EMEs varies considerably over time. More concretely, an increase in a country's external liabilities is associated with an initial rise in inequality (which lasts for a couple of years), followed by a subsequent decline (which starts about five years after the initial increase in external liabilities). The main driver of the above aggregate pattern is Foreign Direct Investment (FDI), which also tends to initially increase and subsequently decrease inequality in the recipient country. Meanwhile, an increase in an EME's other investment (OI) liabilities, which mostly consist of cross-border bank loans, tends to be associated with a fall in inequality, albeit after a lag of a couple of years. In contrast to EMEs, the estimated impact of external financial openness on inequality in AEs is statistically significant only for the initial time lags for most external liability components.

The rest of this paper is organised as follows. In the next section, we review the existing literature. In Section 3, we describe the data that we use for our empirical analysis. We discuss the possible channels through which external financial openness can impact inequality in Section 4. We present our empirical methodology in Section 5. In Section 6, we go over our main results and discuss the intuition behind them. We conclude in Section 7.

## 2 Literature review

There is a large body of theoretical literature on the relationship between financial development and inequality. In their review of this literature, Demirgüç-Kunt and Levine (2009) conclude that theory provides contradicting predictions about the relationship between finance and inequality.

On the one hand, there are several papers according to which finance would lead to declines in income inequality, especially for individuals with lower income. For example, Aghion and Bolton (1997) show that more finance may make it easier for poorer individuals to finance profitable projects, which would lead to a reduction in income inequality. Becker and Tomes (1979) and Greenwood and Jovanovic (1990) find that financial development is beneficial for the disadvantaged groups.

On the other hand, some theoretical models predict that more finance could widen income inequality. Matsuyama (2004) concludes that in a world in which economic agents' ability to borrow abroad is constrained by their domestic collateral, financial globalization might make some countries richer only at the expense of making the rest of the world poorer. Greenwood and Jovanovic (1990) find that improving financial services could widen inequality. An extensive overview with summary of papers on the relationship between financial development and inequality are provided by de Haan and Sturm (2017). They analyse the relationship between finance and inequality and find that all finance variables they consider, increase income inequality. Furthermore, the quality of political institutions conditions the impact of financial liberalization on income inequality, in contrast to the quality of economic institutions.

The existing empirical literature on the relationship between financial openness and inequality has focused exclusively on Foreign Direct Investments (FDI) as a measure of external financial openness. As in the case of the theoretical literature, the empirical results are not uniform. Figini and Görg (2011) find that for developing countries wage inequality increases with FDI, but this effect diminishes with further increases in FDI. For developed countries, wage inequality decreases with FDI, and there is no robust evidence to show that the effect is nonlinear. Jaumotte et al. (2013) find that, while trade globalization is associated with a reduction in inequality, FDI increases income inequality. Their estimates also suggest that technological progress has had a greater impact on inequality than globalization. Cabral et al. (2016) conclude that the level of equity has a large impact on top income shares, suggesting that the channel through which globalization affects income concentration is through FDI/equity flows. Gimet and Lagoarde-Segot (2011) discover a significant causality running from financial sector development to income distribution. In addition, they find that the banking sector seems to exert a stronger impact on inequality, and that the relationship depends more on the characteristics of the financial sector than on its size. Milanović (2005) concludes that foreign direct investment has no effect on income distribution.

In a broader sense, our paper is also related to the literature on the link between trade openness and inequality. The net effect of trade openness on inequality is uncertain in most existing studies (Bank for International Settlements (2017a)). While many papers find that trade openness has reduced inequality (Jaumotte et al. (2013), International Monetary Fund (2007)), and probably lowered unemployment (Görg (2011)), this contrasts with the review of country studies by Goldberg and Pavcnik (2007). These contradicting conclusions may reflect that other factors influence the relationship between trade and inequality. For example, Milanović (2005) finds that trade openness reduces the income share of the poor in low-income countries, but raises it at higher country income levels.

#### 3 Data

We conduct our benchmark empirical exercise using a panel of annual data for 47 countries (30 Advanced Economies (AEs) and 18 Emerging Market Economies (EMEs)) between 1991 and 2013<sup>1</sup>.

We measure inequality in each country using the Gini coefficient from Solt (2016)'s Standardized World Income Inequality Database (SWIID) Version 5.1. This is the one of the most comprehensive datasets available on income inequality. It incorporates data from several sources, including the World Bank's PovcalNet and the United Nations Uni-

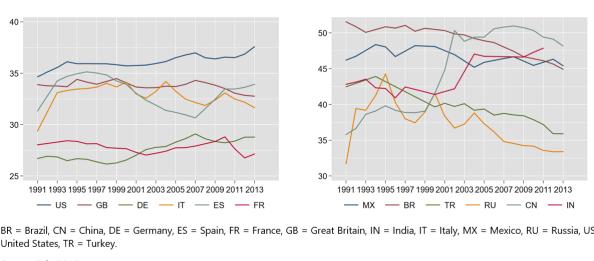
<sup>&</sup>lt;sup>1</sup>The list with all the included countries is available in Appendix B. Additional tables

versity's World Income Inequality Database, and employs the Luxembourg Income Study data as the standard. We use the net Gini index, which is an estimate of the Gini index in equivalized (square root scale) household disposable post-tax and post-transfer income.

Figure 1 shows the Gini coefficients over the observed time period for selected countries. On average, inequality has increased over the past two decades for the observed AEs and decreased for the EMEs except for India and China. That said, there is considerable heterogeneity in the evolution of inequality across countries.

#### Figure 1: Gini coefficients in selected countries

Emerging market economies



BR = Brazil, CN = China, DE = Germany, ES = Spain, FR = France, GB = Great Britain, IN = India, IT = Italy, MX = Mexico, RU = Russia, US =

Source: Solt (2016)

Advanced economies

To measure external financial openness, we use the updated and extended External Wealth of Nations Mark II dataset constructed by Lane and Milesi-Ferretti (2017). It contains annual data for the period 1970 - 2015 and for 211 countries and territories, plus the euro area as a whole. In our benchmark empirical exercise, we use five measures for external financial openness - total external liabilities (Liabilities), foreign direct investments (FDI), portfolio equity liabilities (PE), portfolio debt liabilities (PD) and other investment liabilities (OI). We scale each of the above measures by the GDP of the respective country.

Total external liabilities are defined as all liabilities that the residents of a given country have to residents of other countries in the world. Foreign Direct Investment (FDI) is defined as cross-border investment that is associated with a resident in one economy having control or a significant degree of influence on the management of an enterprise that is resident in another economy (International Monetary Fund (2009)). It includes controlling stakes in acquired foreign firms, as well as Greenfield investments<sup>2</sup>. Portfolio equity positions measure cross-border ownership of shares of companies and mutual funds below the 10 % threshold that distinguishes portfolio from direct investment. Portfolio debt positions are defined as cross-border positions involving debt securities, other than those included in foreign direct investment or reserve assets. Other investment is a residual category that includes positions and transactions other than those included in direct

 $<sup>^{2}</sup>$ Formally, controlling stakes are defined as those amounting to at least 10 % of an entity's equity. In practice, however, most FDI holdings reflect majority control.

investment, portfolio investment, financial derivatives and employee stock options, and reserve asset. For the overwhelming majority of countries in the world, cross-border banks loans and deposits represent the largest component of other investment liabilities.

In order to prevent the external financial positions of financial centers from influencing our results, we exclude two sets of countries from our benchmark empirical analysis: (i) countries defined as offshore centers in the BIS International Banking statistics and (ii) countries for which the sum of external assets and liabilities exceeds 6000 % of GDP.

	Emergin	g marke	et econon	nies		
	mean	s.d.	min	median	max	Ν
Dependent variable						
Gini	40.057	8.822	20.020	39.790	59.759	408
Financial measures						
$LIAB^{1}$	0.724	0.448	0.186	0.635	3.810	410
$\mathrm{FDI}^1$	0.265	0.307	0.008	0.199	2.457	411
$PE^{-1}$	0.090	0.084	0.000	0.063	0.372	410
$\mathrm{PD}^1$	0.095	0.079	0.001	0.077	0.426	386
$OI^1$	0.254	0.154	0.000	0.229	1.270	386
<u>Control variables</u>						
$\mathrm{GDPpCap}^2$	0.711	0.623	0.032	0.508	3.641	410
$\operatorname{credit}^1$	0.732	0.419	0.144	0.612	1.850	397
$capital^1$	10.353	7.286	2.851	8.111	58.978	358
$trade^3$	0.677	0.419	0.138	0.560	2.204	414
unemployment	0.083	0.055	0.006	0.072	0.271	414
	Adva	anced ec	conomies			
Dependent variable						
Gini net	28.978	4.403	17.599	28.668	38.162	658
<u>Financial measures</u>						
$LIAB^{1}$	2.375	3.420	0.150	1.416	28.905	679
$\mathrm{FDI}^1$	0.711	1.966	0.002	0.326	20.912	665
$\mathrm{PE}^{1}$	0.283	0.726	-0.008	0.115	8.784	673
$\mathrm{PD}^{1}$	0.460	0.549	0.000	0.333	4.091	650
<u>Control variables</u>						
$GDPpCap^2$	2.923	1.682	0.063	2.693	10.257	657
$\operatorname{credit}^1$	1.516	0.477	0.352	1.489	3.240	483
$capital^1$	6.824	7.767	2.142	5.193	117.291	573
$trade^3$	0.862	0.482	0.159	0.728	3.245	651
unemployment	0.081	0.042	0.015	0.072	0.275	667

Table 1: Summary statistics

LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, <math>PD = Portfolio Debt liabilities, OI = Other Investmentliabilities. s.d. = standard deviation, N = Number of observations. <sup>1</sup> Scaled by GDP. <sup>2</sup> In 10,000 of US dollars. <sup>3</sup> trade = (exports + imports)/GDP.

In addition, we include several additional variables, which have been demonstrated to be major determinants of the level of inequality in individual countries, as controls on the right-hand side of our benchmark equation. First, we include the Gross Domestic Product (GDP) per capita (obtained from the International Monetary Fund (2017)) as a proxy for the level of economic development. Second, we use the ratio of private credit to GDP (Bank for International Settlements (2017b)) in a given country (Beck et al. (2007)), as a measure of the country's financial development. Third, in order to control for unemployment, we use data from the world development indicators (The World Bank (2017)). A reduction in unemployment will lead to decreases in income inequality. Lowskilled workers at the bottom of the distribution are more vulnerable to unemployment, which is found to increase income inequality (Heathcote et al. (2010)). Table 1 displays summary statistics for the variables we employed in our analysis.

## 4 Theoretical predictions and channels

External financial openness can impact inequality through a number of channels. While many of the relevant channels affect multiple components of external financial openness, each component ultimately impacts inequality through a unique set of channels. This highlights the importance of examining separately the impact of each component of external financial openness on inequality (as opposed to examining their overall joint impact) and motivates the design of our benchmark empirical exercise.

Table 2 displays a summary of the channels associated with each component of external financial openness, along with the respective directional impact associated with each channel. We go over the theoretical explanations underpinning each of those channel in the rest of this section.

	Ez	xternal Liał	oility Com	ponent
Impact on inequality	FDI	PE	PD	OI
increase decrease	· ·	CG, SIG		SIG AC, FC, FX

Table 2: Summary of channels

LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investmentliabilities. AC = Access to Credit, CG = Capital Gains, FC = Funding Conditions, FX = Foreign Exchange, SIG = Special Interest Groups, SP = SkilledPremium, TD = Technology Diffusion.

FDI flows could lead to a rise in inequality through the skilled premium (SP) channel. An inflow of FDI into a given economy is typically associated with the introduction of new production technologies in that economy. Since such technologies are likely to increase capital intensity and the returns to skill, the benefits tend to accrue to higher-income individuals, who are likely to own more capital and to be more highly skilled than the rest of the population (Aghion and Howitt (1998) and Figini and Görg (2011)).

All else the same, the skilled premium channel should be stronger for EMEs than for AEs, since the former tend to have lower initial technology levels and are, therefore, more likely to experience greater technological advances as a result of FDI inflows. Importantly, the effect of this channel should be most powerful contemporaneously and in the immediate aftermath of (i.e. the first few years after) the increase in FDI since that is when the skilled premium in compensation is likely to grow the most.

FDI flows could eventually lead to a fall in inequality through the technological diffusion (TD) channel. This channel is generated by the same mechanism as the one driving the skilled premium channel. Nevertheless, it typically takes longer to materialise and goes in the opposite direction. As the improved technology brought about by the influx of FDI spreads through the recipient economy in a diffusion-like learning process, the share of the population employed in the high-skilled industries increases (Aghion and Howitt (1998)). This ultimately results in a decline in inequality as the wage distribution starts to converge towards the new, higher-level, equilibrium. As in the case of the skilled premium channel, the impact of the technological diffusion channel should be more powerful in the case of EMEs due to their generally lower initial technology levels.

It is important to note that the combination of the above two (SP and TD) channels results in a clear set of predictions about the evolution of the impact of FDI on inequality over time. Namely, the initial impact is expected to be positive, as the skilled premium channel should operate on its own in the initial periods following the increase in FDI. As the technological diffusion channel starts to operate, several years after the initial increase in FDI, the impact on inequality is expected to gradually turn negative.

Greater external financial openness could lead to a fall in inequality through the funding conditions (FC) channel. International financial inflows into a given economy increase the availability of funding in that economy. This eases credit conditions, boosts consumption and investment, and ultimately increases employment. Since unemployed individuals are (by definition) more likely to have lower incomes, the resulting drop in unemployment would lead to a fall in inequality.

Another related, but distinct, channel through which external financial openness could decrease inequality is the access to credit (AC) channel. All else the same, the easing of funding conditions generated by an increase in external financial inflows is likely to increase the access to credit of low-income individuals. In turn, this greater access to credit can enhance their opportunities for income generation (Beck et al. (2007)). In theory, the access to credit channel could operate for any of the major external financial flows components we examine. In practice, however, it should be strongest in the case of other investment (which mainly consist of cross-border bank lending) since low-income individual are most likely to access credit through bank lending. Furthermore, this channel is likely to be more powerful in EMEs, where the share of the population with limited or no access to credit is larger than in AEs.

The foreign exchange rate (FX) channel could also lead to a negative relationship between external financial openness and inequality. External financial flows into a given country tend to lead to an appreciation of that country's exchange rate. This tends to improve the creditworthiness of borrowers with currency mismatches on their balance sheets (Bruno and Shin (2015a), Bruno and Shin (2015b) and Hofmann et al. (2016)). In turn, this improves their access to credit, which, as discussed above, expands their income-generating opportunities. Since low-income individuals tend to own virtually no foreign-currency assets, they tend to have larger currency mismatches (relative to the sizes of their balance sheets). Thus, the foreign exchange rate channel would mostly benefit them, and consequently result in a decline in inequality. As in the case of the access to credit channel, the foreign exchange rate channel is likely to be stronger for EMEs, where currency mismatches tend to be larger and exchange rate fluctuations greater in magnitude than in AEs.

External financial openness could increase inequality through the special interest group (SIG) channel. If the quality of institutions in a given country is low, special interest

groups could capture (all or most of) the financial gains stemming from international financial openness (Claessens and Perotti (2007)). In theory, this channel could operate through each of the main external financial openness components that we examine (FDI, portfolio equity, portfolio debt and other investment). Since institutional quality is generally lower in EMEs than in AEs, the special interest group channel should be more powerful in the former group of countries.

Portfolio equity flows could increase inequality through the capital gains (CG) channel. All else the same, an inflow of portfolio equity into a given country would increase the value of (both publicly-traded and privately-held) equity in that country. Equity holdings in most economies tend to be concentrated in wealthy individuals, who are already in the top part of the income distribution. As a consequence, the capital gains triggered by the influx of foreign portfolio equity flows, would end up increasing inequality.

## 5 Empirical methodology

When designing our empirical investigation on the link between financial openness and inequality, we allow for the possibility that all major components of external liabilities could influence inequality in a given country. In the context of the above theoretical background, this implies that we explore the relationship between all of external financial openness measures (described in the Data Section 3) and the income inequality.

In our benchmark empirical exercise, we use the Gini coefficient computed for income inequality rather than a measure of wage inequality. We choose to do that due to two main reasons. First, focusing on income inequality measures allows us to explore a much larger set of countries over a considerably longer period of time. Second, the correlation between income inequality and wage inequality tends to be very high both, across countries and over time (Galbraith and Kum (2005)).

Since many of the key series of interest have trends, we induce stationarity by detrending all variables used in our benchmark empirical exercise. More concretely, we apply a straightforward detrending procedure which consists of two steps. First, we regress each series on a time trend. We then obtain the residuals from those regressions, designate them as the "detrended" versions of the variables and use them in our benchmark econometric specifications.

More concretely, we estimate the following regression specification with the "detrended" variables:

$$GINI_{it} = \beta_0 + \beta_1 * FM_{it-j} + \beta_2 * \hat{X}_{it} + \mu_i + \lambda_t + \epsilon_{it}$$
(1)

where  $\hat{GINI}_{it}$  is the post-tax and post-transfer inequality index in country *i* and time t,  $\hat{FM}_{it}$  is an external financial measure as a percentage of GDP country *i* and time  $t - j, j \in [0, \dots, 8], \hat{X}_{it}$  is a vector of control variables,  $\mu_i$  is a country fixed effect,  $\lambda_t$  is a time fixed effect and  $\epsilon_{it}$  is the error term.

## 6 Key results

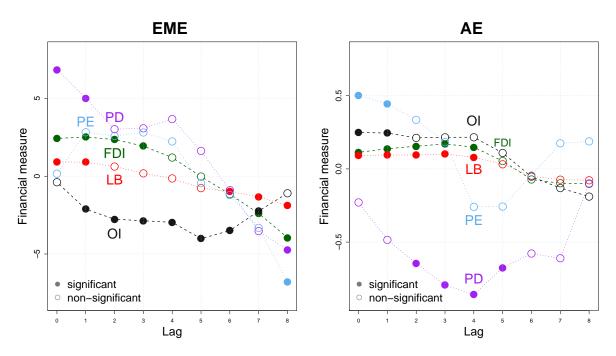
As discussed in Section 4, there are several channels through which a country's external financial openness could impact its degree of inequality.

Since not all of the above channels go in the same direction, we expect that the overall impact of financial openness on inequality would depend on the relative strengths of those channels. In turn, the relative strengths of those channels could vary over time and across countries, depending on their overall level of economic development and the phase of the business cycle they are in. Furthermore, each component of external financial openness impacts inequality through a unique set of channels. That is why, we examine the impact of each component separately, while also allowing it to vary over time.

#### 6.1 Benchmark results

In our benchmark empirical exercise, we estimate the impact of external financial openness and each of its key components on inequality by using the panel regression specification presented in equation (1). Each panel regression that we estimate includes country and time fixed effects. We estimate separate specifications for contemporaneous and lagged (for up to eight years) values of total external liabilities and each of its four main component. The main results from those regressions are summarised in Tables 3 and 4 and the key coefficients of the regressions are plotted in Figure 2<sup>3</sup>.

Figure 2: Plots of estimated coefficients from the regression results in Tables 3 (for the left-hand panel) and Table 4 (for the right-hand panel)



Our benchmark results for EMEs suggest that the impact of increasing external financial openness on inequality varies considerably over time. More concretely, an increase in a country's external liabilities (relative to its GDP) is associated with a rise in inequality in the year in which it occurs and in the subsequent year. The impact then becomes insignificant between the second and the fourth year after the increase in external liabilities. Finally, it turns negative and statistically significant from the sixth year onwards.

The regressions for the individual components of external liabilities reveal that the main driver of the above aggregate pattern in EMEs is FDI. Just as in the case of aggregate external liabilities, an increase in FDI leads to an initial increase in inequality (between

<sup>&</sup>lt;sup>3</sup>The regressions with all the results are available in Appendix A. Robustness tests.

years 0 and 3). This is followed by a subsequent reversion, which ultimately results in a statistically significant decline in inequality from the sixth year onward.

The estimated dynamics of the impact of FDI on inequality is fully in line with the theoretical predictions generated by the combination of the skilled premium channel and the technological diffusion channel, discussed in Section 4. Exactly as predicted by the theoretical model of Aghion and Howitt (1998), the skilled premium (SP) channel, begins to operate virtually immediately. The entry of foreign capital and know-how into an EME through FDI initially raises inequality by increasing the wages of skilled workers relatively to those of unskilled workers. After several years, as the technological diffusion (TD) channel gathers momentum, the wider spread of more productive technology increases the share of the population employed in the high-skilled industries. This ultimately results in a decline in inequality.

An increase in an EME's other investment liabilities, which mostly consist of crossborder bank loans, tends to be associated with a statistically significant fall in inequality, albeit with a lag of a couple of years. The decline tends to be fairly persistent, lasting for seven years. This set of results implies that, in the case of other investment liabilities, the AC, FC and FX channels, all of which should lead to a decline in inequality, dominate the SIG channel, whose effects go in the opposite direction.

The impact of the two portfolio flow categories (PE and PD) on inequality in EMEs tends to be smaller than those of FDI and other investment. The estimated coefficients on both, PE and PD, tend to be insignificant for almost all lags that we examine. The only exceptions are the contemporaneous and the first lag coefficients on PD, which are positive and statistically significant and the eighth-lag coefficient on PE, which is negative and statistically significant. These sets of results are most likely a manifestation of the fact that the channels through which external portfolio liabilities affect inequality largely offset each other.

The key coefficients for the AE sample tend to be considerably less significant than those for EMEs. The most likely explanation for that set of results is that, as discussed in Section 4, the key channels through which the various components of external financial openness impact inequality tend to be weaker in AEs than in EMEs.

That said, there are several instances in which the estimated impacts are estimated to be statistically significant even in the case of AEs. Namely, the estimated coefficients on total external liabilities are statistically significant and positive up to the fourth lag. This appears to be driven by FDI, whose estimated coefficients are positive and statistically significant also up to the fourth lag, and other investment liabilities, which has positive and statistically significant coefficients up to lag one. The former result is most likely a manifestation of the skilled premium channel, which is associated with an increase in inequality due to an increase in the gap between the wages of skilled and unskilled workers. Meanwhile, the fact that the estimated coefficients on FDI for AEs do not turn negative and significant as the number of lags grows (as in the case of EMEs) suggests that the technology diffusion channel is weaker in AEs than in EMEs (which is fully in line with the theoretical predictions outlined in Section 4).

An increase in an AE's PD, which mostly consist of portfolio debt liabilities, tends to be associated with a statistically significant fall in inequality, albeit with a lag of a couple of years. the decline lasts for four years. This set of results implies that, in the case of portfolio debt liabilities, the FC and FX channels, all of which should lead to a decline in inequality, dominate the SIG channel, whose effects go in the opposite direction.

#### 6.2 Robustness tests

In the next step of our empirical investigation, we test the robustness of our benchmark results by including additional explanatory variables as controls into our benchmark panel regression specification. More concretely, we sequentially control for trade openness (measured as the sum of exports and imports, scaled by GDP), unemployment rate, financial depth (measured as the ratio of credit to the private non-financial sector over GDP) and capital.

Variables	L0	L1	L2	L3	L4	L5	L6	L7	L8
LIAB	0.919**	0.904*	0.623	0.184	-0.141	-0.764	-0.992*	-1.317**	-1.884***
	(0.457)	(0.462)	(0.477)	(0.484)	(0.486)	(0.508)	(0.515)	(0.534)	(0.551)
GDPpCap	-1.911***	-1.851***	-1.878***	-2.048***	-2.170***	-2.322***	-2.327***	-2.053***	-1.639**
	(0.677)	(0.682)	(0.699)	(0.710)	(0.703)	(0.701)	(0.709)	(0.729)	(0.766)
Ν	402	385	368	350	332	314	296	278	260
$\mathbb{R}^2$	0.125	0.121	0.111	0.106	0.110	0.123	0.139	0.153	0.180
FDI	2.428***	2.523***	2.360***	1.938**	1.200	-0.0197	-1.156	-2.387**	-3.952***
	(0.836)	(0.841)	(0.869)	(0.906)	(0.906)	(0.980)	(1.025)	(1.083)	(1.122)
GDPpCap	-2.040***	-1.936***	-1.752**	-1.669**	-1.858***	-2.106***	-2.251***	-2.111***	-1.948**
1 1	(0.654)	(0.659)	(0.677)	(0.707)	(0.701)	(0.704)	(0.713)	(0.727)	(0.761)
Ν	403	386	369	351	333	315	297	279	261
$\mathbb{R}^2$	0.136	0.134	0.127	0.123	0.116	0.116	0.130	0.149	0.183
PE	0.163	2.826	2.593	2.813	2.236	-0.444	-1.206	-3.321	-6.799*
	(2.592)	(2.646)	(2.738)	(2.787)	(2.837)	(2.953)	(3.130)	(3.373)	(3.491)
GDPpCap	-2.258***	-2.217***	-2.143***	-2.148***	-2.130***	-2.103***	-2.148***	-2.084***	-2.003***
1 1	(0.659)	(0.662)	(0.670)	(0.670)	(0.677)	(0.689)	(0.709)	(0.737)	(0.768)
Ν	402	385	368	350	332	314	296	278	260
$\mathbb{R}^2$	0.115	0.114	0.109	0.109	0.112	0.116	0.127	0.135	0.151
PD	6.843***	5.012**	3.026	3.082	3.676	1.627	-0.894	-3.530	-4.739*
	(2.424)	(2.520)	(2.605)	(2.629)	(2.636)	(2.639)	(2.666)	(2.730)	(2.763)
GDPpCap	-1.464**	-1.768**	-2.069***	-2.140***	-2.112***	-2.188***	-2.194***	-2.058***	-1.963**
1 1	(0.690)	(0.699)	(0.708)	(0.712)	(0.712)	(0.710)	(0.718)	(0.736)	(0.763)
Ν	378	361	344	326	308	290	272	254	236
$\mathbb{R}^2$	0.132	0.128	0.122	0.121	0.126	0.135	0.152	0.165	0.173
OI	-0.384	-2.117*	-2.766**	-2.861***	-2.965***	-4.015***	-3.495***	-2.237*	-1.094
	(1.054)	(1.077)	(1.079)	(1.069)	(1.076)	(1.083)	(1.102)	(1.162)	(1.178)
GDPpCap	-2.109***	-2.422***	-2.638***	-2.650***	-2.417***	-2.235***	-1.993***	-1.790**	-1.950**
1 1	(0.686)	(0.697)	(0.701)	(0.696)	(0.693)	(0.683)	(0.698)	(0.741)	(0.782)
Ν	378	361	344	326	308	290	272	254	236
$\mathbb{R}^2$	0.112	0.128	0.137	0.138	0.143	0.179	0.186	0.172	0.164

Table 3: Benchmark panel regression results. Emerging market economies

= "\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.001"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. <math>L0 = lag 0, L1 = lag 1 etc.

Variables	LO	L1	L2	L3	L4	L5	L6	L7	L8
LIAB	0.0915**	0.0953**	0.0953**	0.101**	0.0800*	0.0316	-0.0495	-0.0734	-0.0777
	(0.0383)	(0.0377)	(0.0390)	(0.0405)	(0.0437)	(0.0486)	(0.0518)	(0.0512)	(0.0514)
GDPpCap	0.465***	0.469***	0.453***	0.433***	0.400***	0.384***	0.405***	0.426***	0.408***
	(0.140)	(0.138)	(0.138)	(0.139)	(0.142)	(0.147)	(0.150)	(0.150)	(0.151)
Ν	645	620	593	566	538	510	482	454	426
$\mathbb{R}^2$	0.133	0.124	0.135	0.147	0.147	0.142	0.145	0.152	0.149
FDI	0.113**	0.138***	0.155***	0.171***	0.148***	0.0540	-0.0723	-0.0979	-0.102
	(0.0471)	(0.0469)	(0.0489)	(0.0512)	(0.0558)	(0.0609)	(0.0635)	(0.0643)	(0.0653)
GDPpCap	0.434***	0.426***	0.405***	0.384***	0.368***	0.376**	0.409***	0.422***	0.396***
1 1	(0.140)	(0.137)	(0.136)	(0.137)	(0.141)	(0.147)	(0.150)	(0.150)	(0.151)
Ν	645	620	593	566	538	510	482	454	426
$\mathbb{R}^2$	0.134	0.127	0.141	0.155	0.154	0.142	0.146	0.153	0.149
PE	0.503**	0.444*	0.335	0.183	-0.259	-0.257	-0.0448	0.175	0.188
	(0.252)	(0.265)	(0.281)	(0.286)	(0.298)	(0.306)	(0.354)	(0.360)	(0.369)
GDPpCap	0.496***	0.469***	0.428***	0.396***	0.385***	0.391***	0.401***	0.405***	0.377**
one proop	(0.144)	(0.139)	(0.138)	(0.139)	(0.142)	(0.147)	(0.150)	(0.150)	(0.152)
Ν	645	620	593	566	538	510	482	454	426
$\mathbb{R}^2$	0.131	0.119	0.127	0.137	0.143	0.142	0.143	0.148	0.144
PD	-0.229	-0.485	-0.645**	-0.794**	-0.858**	-0.675*	-0.578	-0.610	-0.0993
	(0.283)	(0.301)	(0.307)	(0.322)	(0.339)	(0.375)	(0.380)	(0.385)	(0.421)
GDPpCap	0.383***	0.407***	0.364***	0.293**	0.246	0.289*	0.314**	0.333**	0.334**
1 1	(0.136)	(0.137)	(0.140)	(0.147)	(0.151)	(0.155)	(0.155)	(0.154)	(0.155)
Ν	623	598	570	542	514	486	458	430	402
$\mathbb{R}^2$	0.118	0.124	0.136	0.150	0.160	0.158	0.155	0.152	0.144
OI	0.249*	0.245*	0.212	0.216	0.216	0.109	-0.0527	-0.131	-0.189
	(0.141)	(0.141)	(0.143)	(0.143)	(0.143)	(0.171)	(0.189)	(0.194)	(0.199)
GDPpCap	0.425***	0.484***	0.485***	0.467***	0.415***	0.386***	0.387**	0.395**	0.367**
1 1	(0.138)	(0.142)	(0.143)	(0.142)	(0.141)	(0.145)	(0.151)	(0.155)	(0.159)
Ν	623	598	570	542	514	486	458	430	402
$\mathbb{R}^2$	0.122	0.125	0.133	0.144	0.151	0.152	0.150	0.147	0.146

Table 4: Benchmark panel regression results. Advanced economies

Standard Errors in parentheses

= "\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.001"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. L0 = lag 0, L1 = lag 1 etc.

Tables 5 and 6 contain summaries of the results (for EMEs and AEs, respectively) from the alternative specifications that we estimate. For the most part, the inclusion of the additional control variables does not alter our key results significantly. Most importantly, the estimated dynamic pattern of the impact of FDI on inequality in EMEs remains intact in virtually all alternative specifications we examine.

The only minor exception is that, once we control for financial depth and capital, the estimated coefficients on lag 7 are no longer statistically significant but the ones for lag 8 are still statistically significant. Interestingly, the counterparts to those coefficients in the case AEs, which are not significant in the benchmark estimations, do become negative and statistically significant once we control for the level of financial development. The combination of those results could be interpreted as an indication that financial development interacts with the technology diffusion channel in both EMEs and AEs.

	Trade	Unemployment	Credit	Capital
, lags 6-8 neg lags 7-8 neg	lags 0-1 pos, lags 6-8 neg lags 0-1 pos, lags 6-8 neg lags 0-3 nos lags 7-8 neg lags 0-3 nos lags 7-8 neg	lag 1 pos, lags 6-8 neg lags 0-3 nos lags 7-8 neg	lag 8 neg lags 0.4 nos lag 8 neg	lags 0-2 pos, lags 5-8 neg lags 0-3 pos lag 8 neg
1480 1-0 1108	lag 8 neg	lag 8 neg	lag 8 neg	lag 8 neg
lags 0-1 pos, lag 8 neg	lags 0-1 pos, lag 8 neg	lags 0-1 pos	lags 0-1 pos	lags 0-1 pos
lags 1-7 neg	lags 1-7 neg	lags 1-7 neg	lags 0-6 neg	lags 0-6 neg
	<mark>neg</mark> insiønificant	neg insignificant. / nos	neg DOS	mostly neg mostly neg
	insignificant	nsignificant / pos	pos	mostly neg
		neg insignificant		neg insignificant / pos

$\mathbf{AE}$	Benchmark model	Trade	${f Unemployment}$	Credit	Capital
LIAB	lags 0-4 pos	lags 0-4 pos	lags 0-4 pos	lags 0-3 pos, lags 7-8 neg lags 1-3 pos	lags 1-3 pos
FDI	lags $0-4$ pos	lags $0-4$ pos	lags 0-4 pos, lag 8 neg	lags 0-2 pos, lags 5-8 neg	lags 1-5 pos
PE	lags $0-1$ pos	lags $0-2$ pos	insignificant	insignificant	lags 7-8 pos
PD	lags $2-5 \text{ neg}$	lags 2-5 neg	lags 0-5 neg	insignificant	lags 4-8 neg
IO	lags 0-1 pos	insignificant	insignificant	lags 0-2 pos, lag 8 neg	lags 0-2 pos
GDPpCap	sod d	bos	sod	insignificant	sod
controls		insignificant	bos	insignificant	insignificant / neg

## 7 Conclusions

In this paper, we attempt to fill the gap in the empirical literature on the link between external financial openness and inequality. We do that by using an annual panel of 47 countries between 1991 and 2013 to estimate regressions of Gini - based inequality measures on various indicators of external financial openness, while controlling for country heterogeneity and time trends.

Our results for emerging economies indicate that the impact of increasing external financial openness on inequality varies considerably over time. More concretely, an increase in a country's external liabilities (relative to its GDP) is associated with a rise in inequality in the year in which it occurs and in the subsequent year. The impact then turns negative and statistically significant from the sixth year onwards. The main driver of the above aggregate pattern in EMEs is FDI, whose estimated impact exhibits a very similar dynamics.

The above set of results is fully in line with the predictions of the theoretical literature (Aghion and Howitt (1998)). More concretely, the estimated dynamics appears to be driven by the combination of the skilled premium channel (which tends to increase inequality contemporaneously) and the technological diffusion channel (which tends to decrease inequality with a lag of several years).

We also find that an increase in an EME's other investment liabilities, which primarily consist of cross-border bank loans, tends to lead to a decline in inequality. This effect takes places with a lag of one year and is fairly persistent. The estimated impacts of the two portfolio flow categories (PE and PD) tend to be much smaller.

Our results also suggest that the impact of external financial openness on inequality in AEs tends to be smaller than in EMEs. An increase in a country's external liabilities (relative to its GDP) is associated with a rise in inequality in the year in which it occurs and in the subsequent years up to the fourth year. The main driver of the above aggregate pattern in EMEs is FDI. This result is most likely a manifestation of the skilled premium channel. Meanwhile, the fact that the estimated coefficients on FDI for AEs do not turn negative and significant as the number of lags grows (as in the case of EMEs) suggests that the technology diffusion channel is weaker in AEs than in EMEs (which is fully in line with the theoretical predictions).

#### References

- Aghion, P. and Bolton, P. (1997). A theory of trickle-down growth and development. *Review of Economic Studies*, 64(2), 151–172.
- Aghion, P. and Howitt, P. (1998). *Endogenous Growth Theory*. Cambridge, MA: MIT Press.
- Atkinson, A. B. (2015). Inequality: What Can Be Done? Harvard University Press.
- Autor, D., Dorn, D., Katz, L., Patterson, C., and Van Reenen, J. (2017). Concentrating on the fall of the labor share. *American Economic Review: Papers & Proceedings*, 107(5), 180–5.
- Bank for International Settlements (2017a). 87th Annual Report, Understanding globalisation, Chapter VI, 97–124.

Bank for International Settlements (2017b). Total Credit to Non-Financial Sector Data.

- Beck, T., Demirgüç-Kunt, A., and Levine, R. (2007). Finance, inequality and the poor. Journal of Economic Growth, 12(1), 27–49.
- Beck, T., Levine, R., and Levkov, A. (2010). Big bad banks? The winners and losers from bank deregulation in the United States. *The Journal of Finance*, 65(5), 1637–67.
- Becker, G. and Tomes, N. (1979). An equilibrium theory of the distribution of income and intergenerational mobility. *Journal of Political Economy*, 87(6), 1153–89.
- Ben Naceur, S. and Zhang, R. X. (2016). Financial development, inequality and poverty: some international evidence. *IMF Working Papers*, no WP/16/32.
- Bruno, V. and Shin, H. S. (2015a). Capital flows and the risk-taking channel of monetary policy. *Journal of Monetary Economics*, 71, 119–124.
- Bruno, V. and Shin, H. S. (2015b). Cross-border banking and global liquidity. *Review of Economic Studies*, 82(2), 535–564.
- Cabral, R., García-Díaz, R., and Varella Mollick, A. (2016). Does globalization affect top income inequality? *Journal of Policy Modeling*, 38(5), 916–40.
- Claessens, S. and Perotti, E. (2007). Finance and inequality: channels and evidence. Journal of Comparative Economics, 35(4), 748–73.
- de Haan, J. and Sturm, J.-E. (2017). Finance and income inequality: A review and new evidence. *European Journal of Political Economy*.
- Demirgüç-Kunt, A. and Levine, R. (2009). Finance and inequality: Theory and evidence. Annual Review of Financial Economics, Annual Reviews, 1(1), 287–318.
- Denk, O. and Cournède, B. (2015). Finance and income inequality in OECD countries. OECD Economics Department Working Papers, no 1224.
- Fajgelbaum, P. and Khandelwal, A. (2016). Measuring the unequal gains from trade. The Quarterly Journal of Economics, 131(3), 1113–80.
- Feenstra, R. C., Inklaar, R., and Timmer, M. P. (2015). The Next Generation of the Penn World Table. American Economic Review, 105(10), 3150–3183.
- Figini, P. and Görg, H. (2011). Does foreign direct investment affect wage inequality? An empirical investigation. *The World Economy, Wiley Blackwell*, 34(9), 1455–1475.
- Galbraith, J. K. and Kum, H. (2005). Estimating the inequality of household incomes: A statistical approach to the creation of a dense and consistent global data set. *Review of Income and Wealth*, 51, 115—143.
- Gimet, C. and Lagoarde-Segot, T. (2011). A closer look at financial development and income distribution. *Journal of Banking & Finance*, 35(7), 1698–1713.
- Goldberg, P. and Pavcnik, N. (2007). Distributional effects of globalization in developing countries. *Journal of Economic Literature*, 45(1), 39–82.

- Görg, H. (2011). Globalization, offshoring and jobs. In M. Bacchetta and M. Jansen (Eds.), *Making globalization socially sustainable* (pp. 21–48). World Trade Organization.
- Greenwood, J. and Jovanovic, B. (1990). Financial development, growth, and the distribution of income. *Journal of Political Economy*, 98(5), 1076–1107.
- Heathcote, J., Perri, F., and Violante, G. (2010). Unequal we stand: an empirical analysis of economic inequality in the United States, 1967-2006. *Review of Economic Dynamics*, 13(1), 15–51.
- Hofmann, B., Shim, I., and Shin, H. S. (2016). Sovereign yields and the risk-taking channel of currency appreciation. *BIS Working Papers*, no 538, revised May 2017.
- International Monetary Fund (2007). Globalization and inequality. *World Economic Outlook*, October, Chapter 4.
- International Monetary Fund (2009). Balance of Payments and International Investment Position Manual, 6th edition.
- International Monetary Fund (2017). World Economic Outlook.
- Jaumotte, F., Lall, S., and Papageorgiou, C. (2013). Rising income inequality: Technology, or trade and financial globalization? *IMF Economic Review*, 61(2), 271–309.
- Lane, P. R. and Milesi-Ferretti, G. M. (2017). International financial integration in the aftermath of the global financial crisis. *IMF Working Paper*, no WP/17/115.
- Matsuyama, K. (2004). Financial market globalization, symmetry-breaking, and endogenous inequality of nations. *Econometrica*, 72, 853—884.
- Milanović, B. (2005). Can we discern the effect of globalization on income distribution? Evidence from household surveys. *The World Bank Economic Review*, 19(1), 21–44.
- Solt, F. (2016). The standardized world income inequality database. *Social Science Quarterly*, 97. SWIID Version 5.1(July).
- The World Bank (2017). World Development Indicators.

# Appendices

# A. Robustness tests

Table A.1:	Panel regression	results	controlling	for	${\rm trade}$	openness.	Emerging market
economies							

Variables	L0	L1	L2	L3	L4	L5	L6	L7	L8
LIAB	0.964**	0.914**	0.623	0.185	-0.135	-0.763	-1.026**	-1.376**	-1.943***
	(0.463)	(0.464)	(0.479)	(0.485)	(0.487)	(0.508)	(0.516)	(0.537)	(0.553)
GDPpCap	-2.016***	-1.907***	-1.881**	-2.005***	-2.048***	-2.179***	-2.098***	-1.790**	-1.365*
	(0.698)	(0.709)	(0.731)	(0.747)	(0.745)	(0.744)	(0.750)	(0.773)	(0.811)
trade	-0.643	-0.304	-0.0194	0.220	0.606	0.713	1.181	1.336	1.425
	(1.007)	(1.042)	(1.111)	(1.172)	(1.217)	(1.226)	(1.259)	(1.309)	(1.387)
Ν	402	385	368	350	332	314	296	278	260
$R^2$	0.126	0.121	0.111	0.106	0.111	0.124	0.142	0.157	0.184
FDI	2.427***	2.521***	2.368***	1.957**	1.219	-0.0344	-1.263	-2.624**	-4.411***
гDI	(0.837)	(0.843)	(0.872)	(0.910)	(0.907)	(0.982)	(1.031)		(1.154)
CDD-C	(0.837) -2.104***	(0.843) -1.945***	(0.872) -1.716**	(0.910) -1.603**	(0.907) -1.734**	(0.982) -1.954***	(1.031) -2.017***	(1.101) -1.798**	(1.134) -1.487*
GDPpCap									
	(0.678)	(0.691)	(0.714)	(0.750)	(0.744)	(0.744)	(0.752)	(0.773)	(0.810)
trade	-0.339	-0.0461	0.175	0.315	0.609	0.772	1.238	1.575	2.313
	(0.934)	(1.027)	(1.102)	(1.181)	(1.210)	(1.224)	(1.264)	(1.321)	(1.429)
N	403	386	369	351	333	315	297	279	261
$\mathbf{R}^2$	0.136	0.134	0.127	0.124	0.116	0.117	0.134	0.154	0.193
$\mathbf{PE}$	0.633	2.070	0.0525	0.0313	-0.136	-2.514	-2.541	-4.284	-7.727**
	(2.603)	(2.650)	(2.814)	(2.904)	(2.933)	(2.998)	(3.127)	(3.346)	(3.435)
GDPpCap	-1.999***	-1.910***	-1.776***	-1.777***	-1.788***	-1.799***	-1.829**	-1.711**	-1.545**
	(0.664)	(0.665)	(0.672)	(0.675)	(0.682)	(0.688)	(0.709)	(0.739)	(0.766)
trade	2.428***	2.810***	3.019***	2.872***	2.645***	2.744***	2.652***	2.996***	3.791***
	(0.786)	(0.825)	(0.887)	(0.920)	(0.935)	(0.935)	(0.952)	(1.083)	(1.190)
Ν	389	376	362	346	330	314	296	278	260
$\mathbb{R}^2$	0.132	0.141	0.140	0.136	0.135	0.143	0.153	0.162	0.188
PD	7.258***	5.337**	3.238	3.305	3.826	1.665	-0.933	-3.722	-5.174*
ТD	(2.513)	(2.612)	(2.647)	(2.650)	(2.653)	(2.647)	(2.677)	(2.785)	(2.853)
GDPpCap	(2.313) $-1.302^*$	(2.012) -1.643**	(2.047) -1.953***	$-1.965^{***}$	$-1.972^{***}$	$-2.116^{***}$	$-2.141^{***}$	(2.785) -1.967**	-1.789**
GDFpCap		(0.746)							
4	(0.737)		(0.751)	(0.754)	(0.754)	(0.753)	(0.759)	(0.778)	(0.813)
trade	0.666	0.538	0.545	0.853	0.701	0.363	0.287	0.499	0.906
Ът	(1.055)	(1.113)	(1.159)	(1.207)	(1.238)	(1.244)	(1.288)	(1.367)	(1.451)
N	378	361	344	326	308	290	272	254	236
$\mathbf{R}^2$	0.133	0.129	0.122	0.123	0.127	0.136	0.152	0.165	0.175
OI	-0.397	-2.279**	-3.029***	-3.075***	-3.000***	-4.011***	-3.495***	-2.240*	-1.084
	(1.159)	(1.123)	(1.115)	(1.087)	(1.079)	(1.085)	(1.107)	(1.168)	(1.193)
GDPpCap	-2.105***	-2.334***	-2.466***	-2.428***	-2.284***	-2.187***	-1.991***	-1.797**	-1.937**
	(0.701)	(0.718)	(0.725)	(0.726)	(0.731)	(0.722)	(0.739)	(0.781)	(0.816)
trade	0.0303	0.582	1.099	1.297	0.712	0.253	0.00865	-0.0379	0.0787
	(1.133)	(1.119)	(1.169)	(1.205)	(1.221)	(1.211)	(1.262)	(1.340)	(1.430)
Ν	378	361	344	326	308	290	272	254	236
$\mathbb{R}^2$	0.112	0.129	0.139	0.142	0.144	0.179	0.186	0.172	0.164
Standard En	in none		0.100			0.1.0	0.100	5.1.2	0.101

Standard Errors in parentheses

="\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.001"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. L0 = lag 0, L1 = lag 1 etc.

Variables	L0	L1	L2	L3	L4	L5	L6	L7	L8
LIAB	0.077*	0.085**	0.088**	0.097**	0.079*	0.036	-0.046	-0.071	-0.075
	(0.041)	(0.039)	(0.040)	(0.041)	(0.044)	(0.049)	(0.052)	(0.051)	(0.052)
GDPpCap	0.488***	0.500***	0.486***	$0.465^{***}$	0.442***	0.427***	$0.435^{***}$	0.446***	0.427***
	(0.144)	(0.142)	(0.142)	(0.143)	(0.147)	(0.152)	(0.155)	(0.156)	(0.156)
trade	0.690	0.625	0.658	0.594	0.745	0.737	0.517	0.353	0.364
	(0.675)	(0.655)	(0.636)	(0.639)	(0.655)	(0.676)	(0.703)	(0.719)	(0.747)
Ν	637	615	591	566	538	510	482	454	426
$\mathbb{R}^2$	0.135	0.126	0.136	0.148	0.149	0.144	0.146	0.153	0.149
EDI	0.005*	0 100**	0 1 47***	0.100***	0 1 40***	0.061	0.007	0.004	0.007
FDI	$0.095^{*}$	$0.126^{**}$	$0.147^{***}$	$0.168^{***}$	$0.148^{***}$	0.061	-0.067	-0.094	-0.097
GDPpCap	(0.050) $0.460^{***}$	(0.049) $0.456^{***}$	(0.050) $0.438^{***}$	(0.051) $0.418^{***}$	(0.056) $0.413^{***}$	(0.061) $0.420^{***}$	(0.064) $0.437^{***}$	(0.065) $0.440^{***}$	(0.066) $0.412^{***}$
GDFpCap		(0.430) (0.142)							
trade	(0.144) 0.666	(0.142) 0.546	$(0.141) \\ 0.603$	(0.142) 0.604	$(0.146) \\ 0.780$	$(0.152) \\ 0.771$	$(0.155) \\ 0.487$	$(0.155) \\ 0.308$	(0.156) 0.316
or acte	(0.680)	(0.654)	(0.632)	(0.634)	(0.653)	(0.678)	(0.487) $(0.705)$	(0.722)	(0.751)
Ν	(0.080) 637	(0.054) 615	(0.052) 591	(0.054) 566	(0.053)	(0.078) 510	(0.703) 482	(0.722) 454	(0.751) 426
$R^2$	0.135	0.129	0.142	0.156	0.156	0.145	0.147	0.153	0.150
10	0.100	0.120	0.112	0.100	0.100	0.110	0.111	0.100	0.100
PE	0.470*	0.439*	0.334	0.175	-0.266	-0.264	-0.0657	0.146	0.149
	(0.252)	(0.265)	(0.281)	(0.286)	(0.298)	(0.306)	(0.355)	(0.364)	(0.375)
GDPpCap	0.537***	0.523***	0.478***	0.438***	0.431***	0.433***	0.435***	0.429***	0.402**
	(0.147)	(0.144)	(0.142)	(0.144)	(0.147)	(0.152)	(0.156)	(0.157)	(0.158)
trade	1.008	0.986	0.903	0.736	0.791	0.714	0.578	0.397	0.436
	(0.640)	(0.632)	(0.628)	(0.639)	(0.657)	(0.675)	(0.703)	(0.726)	(0.755)
Ν	637	615	591	566	538	510	482	454	426
$R^2$	0.135	0.123	0.131	0.139	0.145	0.144	0.145	0.149	0.145
חח	0.996	0.470	-0.640**	-0.797**	-0.861**	0 670*	0 501	0,600	0.0701
PD	-0.226 (0.282)	-0.479 (0.301)		(0.322)	(0.340)	$-0.679^{*}$	-0.581 (0.381)	-0.609 (0.385)	-0.0791 (0.421)
GDPpCap	(0.282) $0.450^{***}$	(0.301) $0.461^{***}$	(0.308) $0.397^{***}$	(0.322) $0.311^{**}$	(0.340) 0.256	$(0.376) \\ 0.296^*$	(0.301) $0.329^{**}$	$(0.362^{**})$	(0.421) $0.372^{**}$
өрт роар	(0.141)	(0.1401)	(0.145)	(0.151)	(0.156)	(0.159)	(0.161)	(0.159)	(0.161)
trade	$1.149^*$	0.939	0.566	0.331	0.178	0.130	0.277	0.557	0.752
trade	(0.646)	(0.645)	(0.648)	(0.659)	(0.683)	(0.706)	(0.746)	(0.783)	(0.811)
Ν	(0.010) 622	(0.019)	(0.010)	(0.055)	(0.000)	(0.100) 486	(0.110)	(0.103)	402
$R^2$	0.123	0.127	0.138	0.151	0.160	0.158	0.155	0.153	0.146
OI	0.207	0.203	0.190	0.213	0.189	0.108	-0.0537	-0.121	-0.169
	(0.144)	(0.147)	(0.149)	(0.147)	(0.151)	(0.172)	(0.190)	(0.195)	(0.201)
GDPpCap	0.474***	$0.514^{***}$	$0.499^{***}$	$0.470^{***}$	$0.414^{***}$	$0.387^{**}$	$0.401^{**}$	$0.421^{***}$	0.397**
	(0.142)	(0.144)	(0.146)	(0.145)	(0.146)	(0.151)	(0.157)	(0.159)	(0.163)
trade	0.964	0.706	0.354	0.0641	-0.0233	0.0183	0.253	0.526	0.674
	(0.658)	(0.670)	(0.676)	(0.681)	(0.696)	(0.709)	(0.748)	(0.787)	(0.815)
	· /	· · · ·		· /	· · · ·	· · · ·			· · · ·
$rac{N}{R^2}$	$622 \\ 0.125$	598 0.126	$570 \\ 0.133$	542 0.144	514 0.151	$486 \\ 0.152$	458 0.151	430 0.148	402 0.147

Table A.2: Panel regression results controlling for trade openness. Advanced economies

="\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.001"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. L0 = lag 0, L1 = lag 1 etc.

Variables	L0	L1	L2	L3	L4	L5	L6	L7	L
LIAB	0.706	0.785*	0.571	0.183	-0.125	-0.737	-0.924*	-1.255**	-1.865***
	(0.479)	(0.472)	(0.480)	(0.484)	(0.488)	(0.517)	(0.530)	(0.555)	(0.576)
GDPpCap	-1.729**	-1.671**	-1.693**	-1.863**	-2.086***	-2.253***	-2.196***	-1.961**	-1.619**
1 1	(0.687)	(0.698)	(0.720)	(0.738)	(0.739)	(0.743)	(0.749)	(0.762)	(0.788)
unemp	6.370	5.272	4.814	4.261	1.803	1.496	3.013	2.442	0.675
P	(4.324)	(4.425)	(4.559)	(4.689)	(4.862)	(5.240)	(5.508)	(5.763)	(5.969)
Ν	402	385	368	350	332	314	296	278	260
$\mathbb{R}^2$	0.130	0.125	0.114	0.109	0.110	0.123	0.140	0.154	0.18
FDI	2.122**	2.331***	2.255**	1.880**	1.206	0.0473	-1.003	-2.233*	-4.019***
1 D1	(0.868)	(0.864)	(0.882)	(0.909)	(0.907)	(0.991)	(1.054)	(1.140)	(1.207
GDPpCap	-1.841***	-1.775***	-1.627**	-1.524**	-1.768**	-1.991***	-2.097***	-2.010***	-1.979**
орт реар	(0.672)	(0.680)	(0.699)	(0.732)	(0.733)	(0.742)	(0.753)	(0.764)	(0.787)
unemp	5.458	4.216	3.284	3.641	2.083	(0.142) 2.569	3.483	2.563	-0.951
unemp	(4.231)	(4.387)	(4.561)	(4.733)	(4.826)	(5.190)	(5.475)	(5.815)	(6.178
Ν	(4.231) 403	(4.387)	(4.501)	(4.735) 351	(4.820)	(3.190) 315	(0.475) 297	(3.813) 279	26
$R^2$	0.140	0.136	0.129	0.125	0.116	0.116	0.132	0.150	0.18
n	0.140	0.130	0.129	0.120	0.110	0.110	0.132	0.130	0.10
PE	0.272	2.878	2.563	2.766	2.174	-0.615	-1.529	-3.620	-6.974*
	(2.581)	(2.640)	(2.737)	(2.788)	(2.847)	(2.972)	(3.146)	(3.381)	(3.492)
GDPpCap	-1.914***	$-1.924^{***}$	-1.914***	$-1.966^{***}$	$-2.058^{***}$	$-1.983^{***}$	-1.926***	-1.834**	-1.743*
	(0.678)	(0.686)	(0.697)	(0.701)	(0.710)	(0.722)	(0.741)	(0.769)	(0.799)
unemp	8.306**	6.886	5.332	4.178	1.681	2.942	5.540	6.347	6.75
	(4.135)	(4.335)	(4.539)	(4.683)	(4.848)	(5.201)	(5.409)	(5.623)	(5.799)
Ν	402	385	368	350	332	314	296	278	26
$\mathbb{R}^2$	0.125	0.121	0.113	0.111	0.112	0.117	0.131	0.140	0.15
PD	6.008**	4.520*	2.816	3.186	4.012	2.043	-0.428	-3.141	-4.11
	(2.527)	(2.563)	(2.613)	(2.632)	(2.672)	(2.693)	(2.727)	(2.801)	(2.849)
GDPpCap	-1.328*	-1.610**	-1.891***	-1.955***	-1.935**	-1.997***	-1.989***	-1.910**	-1.793**
	(0.700)	(0.715)	(0.729)	(0.739)	(0.746)	(0.751)	(0.760)	(0.773)	(0.787)
unemp	5.051	4.746	4.805	4.503	3.990	4.307	4.750	3.783	5.44
P	(4.345)	(4.531)	(4.695)	(4.812)	(5.048)	(5.473)	(5.729)	(5.942)	(6.044
Ν	378	361	344	326	308	290	272	254	23
$\mathbb{R}^2$	0.136	0.131	0.125	0.124	0.128	0.137	0.155	0.166	0.17
OI	-0.487	-2.115*	-2.690**	-2.798***	-2.934***	-3.982***	-3.472***	-2.279*	-1.17
01	(1.051)	(1.076)	(1.082)	(1.074)	(1.085)	(1.089)	(1.104)	(1.163)	(1.177
GDPpCap	(1.031) $-1.793^{**}$	$-2.172^{***}$	-2.455***	$-2.518^{***}$	-2.366***	$-2.167^{***}$	-1.818**	(1.103) $-1.571^{**}$	-1.663*
от роар	(0.702)	(0.719)	(0.728)	(0.725)	(0.724)	(0.715)	(0.730)	(0.774)	(0.810
unomn	(0.702) 8.088*	(0.719) 6.198	(0.728) 4.328	(0.725) 3.152	(0.724) 1.246	(0.713) 1.730	(0.730) 4.486	(0.774) 5.650	(0.810
unemp									
N	(4.205)	(4.449)	(4.656)	(4.783)	(4.971)	(5.258)	(5.493)	(5.769)	(5.886)
N $R^2$	378	361	344	326	308	290	272	254	23
к-	0.122	0.133	0.139	0.140	0.144	0.179	0.189	0.176	0.17

Table A.3: Panel regression results controlling for unemployment rate. Emerging market economies

="\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.01"

 $\begin{array}{ccc} P > 0.001 \\ \hline \\ Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. L0 = lag 0, L1 = lag 1 etc. \end{array}$ 

L8	L7	L6	L5	L4	L3	L2	L1	L0	Variables
-0.076	-0.067	-0.044	0.027	0.078*	0.106***	0.104***	0.097***	0.089**	LIAB
(0.050)	(0.049)	(0.050)	(0.047)	(0.042)	(0.040)	(0.038)	(0.037)	(0.038)	
).606***	0.616***	0.601***	0.599***	0.616***	0.640***	0.660***	0.648***	0.641***	GDPpCap
(0.151)	(0.149)	(0.149)	(0.145)	(0.141)	(0.138)	(0.138)	(0.139)	(0.142)	1 1
12.67***	12.95***	13.30***	14.23***	13.92***	12.95***	12.70***	11.05***	10.82***	unemp
(2.439)	(2.361)	(2.346)	(2.269)	(2.198)	(2.104)	(2.083)	(2.088)	(2.117)	1
426	454	482	510	538	566	593	620	645	Ν
0.205	0.211	0.204	0.209	0.212	0.205	0.190	0.165	0.170	$\mathbf{R}^2$
-0.114*	-0.102	-0.0722	0.0414	0.143***	0.180***	0.174***	0.149***	0.120***	FDI
(0.063)	(0.062)	(0.061)	(0.059)	(0.054)	(0.049)	(0.047)	(0.046)	(0.046)	
).598***	0.616***	0.606***	0.593***	0.585***	0.590***	0.611***	0.608***	0.615***	GDPpCap
(0.151)	(0.149)	(0.149)	(0.146)	(0.140)	(0.137)	(0.136)	(0.138)	(0.141)	
12.85***	13.07***	13.34***	14.20***	13.87***	13.04***	12.97***	11.32***	11.06***	unemp
(2.438)	(2.358)	(2.343)	(2.269)	(2.190)	(2.092)	(2.075)	(2.084)	(2.115)	
426	454	482	510	538	566	593	620	645	Ν
0.207	0.212	0.205	0.209	0.218	0.214	0.199	0.171	0.172	$\mathbf{R}^2$
0.274	0.273	0.0636	-0.125	-0.0844	0.322	0.378	0.392	0.372	PE
(0.357)	(0.348)	(0.343)	(0.295)	(0.288)	(0.277)	(0.272)	(0.260)	(0.249)	
).573***	0.596***	0.595***	0.602***	0.600***	$0.608^{***}$	0.631***	0.641***	0.651***	GDPpCap
(0.152)	(0.149)	(0.149)	(0.146)	(0.141)	(0.139)	(0.138)	(0.140)	(0.144)	
12.77***	13.12***	$13.36^{***}$	14.18***	13.89***	13.04***	$12.56^{***}$	10.89***	10.56***	unemp
(2.447)	(2.367)	(2.351)	(2.275)	(2.216)	(2.123)	(2.093)	(2.098)	(2.135)	
426	454	482	510	538	566	593	620	645	Ν
0.202	0.208	0.203	0.209	0.207	0.196	0.182	0.159	0.166	$R^2$
0.0770	-0.486	-0.575	-0.790**	-1.046***	-0.934***	-0.742**	-0.548*	-0.244	PD
(0.410)	(0.373)	(0.366)	(0.360)	(0.326)	(0.311)	(0.299)	(0.292)	(0.275)	
0.575***	0.564***	$0.528^{***}$	0.487***	$0.433^{***}$	$0.481^{***}$	0.546***	0.601***	$0.565^{***}$	GDPpCap
(0.159)	(0.155)	(0.154)	(0.151)	(0.148)	(0.144)	(0.140)	(0.137)	(0.136)	
12.66***	13.09***	$13.68^{***}$	14.29***	$14.52^{***}$	13.51***	$12.30^{***}$	12.60***	11.88***	unemp
(2.692)	(2.513)	(2.377)	(2.273)	(2.205)	(2.171)	(2.166)	(2.129)	(2.103)	
402	430	458	486	514	542	570	598	623	Ν
0.194	0.208	0.218	0.227	0.231	0.213	0.187	0.177	0.165	$\mathbb{R}^2$
-0.150	-0.0432	0.0661	0.198	0.198	0.207	0.189	0.188	0.198	OI
(0.194)	(0.189)	(0.184)	(0.143)	(0.143)	(0.138)	(0.139)	(0.138)	(0.138)	
).597***	$0.606^{***}$	$0.590^{***}$	$0.598^{***}$	0.626***	0.667***	$0.664^{***}$	0.660***	$0.594^{***}$	GDPpCap
(0.162)	(0.155)	(0.149)	(0.143)	(0.140)	(0.141)	(0.143)	(0.141)	(0.138)	
12.53***	$13.25^{***}$	13.79***	14.21***	13.93***	13.00***	11.91***	$12.25^{***}$	11.66***	unemp
(2.682)	(2.523)	(2.399)	(2.285)	(2.217)	(2.180)	(2.173)	(2.136)	(2.105)	
402	430	458	486	514	542	570	598	623	N
0.195	0.205	0.214	0.221	0.218	0.202	0.180	0.174	0.167	$\mathbf{R}^2$

Table A.4: Panel regression results controlling for unemployment rate. Advanced economies

="\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.001"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. L0 = lag 0, L1 = lag 1 etc.

Variables	L0	L1	L2	L3	L4	L5	L6	L7	L
LIAB	0.483	0.575	0.516	0.409	0.287	-0.291	-0.514	-0.909	-1.573***
	(0.511)	(0.472)	(0.474)	(0.484)	(0.502)	(0.539)	(0.561)	(0.576)	(0.563)
GDPpCap	-1.854***	-1.690**	-1.563**	-1.568**	-1.661**	-1.921***	-1.999***	-1.809**	-1.406*
1 1	(0.682)	(0.681)	(0.698)	(0.716)	(0.716)	(0.715)	(0.722)	(0.737)	(0.766)
credit	2.068**	2.649***	2.959***	2.977***	2.787***	2.370**	2.145**	2.144*	2.731*
	(0.869)	(0.840)	(0.857)	(0.884)	(0.936)	(0.975)	(1.031)	(1.165)	(1.216
Ν	389	376	362	346	330	314	296	278	26
$R^2$	0.134	0.144	0.143	0.138	0.136	0.142	0.153	0.165	0.19
FDI	1.860**	2.064**	2.161**	2.255**	1.967**	0.994	-0.158	-1.593	-3.433**
1 D1	(0.875)	(0.852)	(0.863)	(0.898)	(0.921)	(1.024)	(1.095)	(1.148)	(1.138
GDPpCap	-1.882***	-1.707**	-1.453**	-1.195*	-1.316*	-1.631**	-1.870**	-1.845**	-1.695*
ODI peap	(0.662)	(0.662)	(0.677)	(0.710)	(0.710)	(0.711)	(0.723)	(0.735)	(0.763
credit	1.938**	2.522***	2.872***	3.144***	3.098***	2.886***	$2.476^{**}$	(0.100) $2.279^{**}$	2.652*
creati	(0.813)	(0.827)	(0.851)	(0.882)	(0.918)	(0.959)	(1.015)	(1.145)	(1.213
Ν	(0.813) 389	(0.827) 376	(0.851)	(0.002) 347	(0.918) 331	(0.959) 315	(1.013) 297	(1.143) 279	26
$R^2$	0.143	0.155	0.157	0.159	0.148	0.144	0.150	0.163	0.20
11	0.145	0.155	0.157	0.159	0.140	0.144	0.150	0.105	0.20
PE	0.633	2.070	0.0525	0.0313	-0.136	-2.514	-2.541	-4.284	-7.727*
	(2.603)	(2.650)	(2.814)	(2.904)	(2.933)	(2.998)	(3.127)	(3.346)	(3.435)
GDPpCap	-1.999***	-1.910***	-1.776***	-1.777***	-1.788***	-1.799***	-1.829**	-1.711**	-1.545*
1 1	(0.664)	(0.665)	(0.672)	(0.675)	(0.682)	(0.688)	(0.709)	(0.739)	(0.766)
credit	2.428***	2.810***	3.019***	2.872***	2.645***	2.744***	2.652***	2.996***	3.791**
	(0.786)	(0.825)	(0.887)	(0.920)	(0.935)	(0.935)	(0.952)	(1.083)	(1.190
Ν	389	376	362	346	330	314	296	278	26
R-squared	0.132	0.141	0.140	0.136	0.135	0.143	0.153	0.162	0.18
	F 071**	4.040*	9.170	9.610	2.005	1 579	0.000	0.040	9.00
PD	5.871**	$4.642^{*}$	3.170	3.610	3.995	1.573	-0.606	-2.848	-3.88
app a	(2.466)	(2.511)	(2.585)	(2.608)	(2.609)	(2.608)	(2.638)	(2.716)	(2.750
GDPpCap	-1.326*	-1.533**	-1.740**	-1.774**	-1.760**	-1.876***	-1.850**	-1.723**	-1.666*
	(0.691)	(0.697)	(0.709)	(0.715)	(0.714)	(0.712)	(0.723)	(0.741)	(0.763
credit	2.387***	2.772***	2.850***	2.701***	2.632***	2.500***	2.551**	2.771**	3.131*
	(0.825)	(0.859)	(0.888)	(0.909)	(0.930)	(0.948)	(0.996)	(1.161)	(1.266)
N	370	356	341	324	307	290	272	254	23
$\mathbb{R}^2$	0.151	0.153	0.150	0.147	0.151	0.158	0.175	0.186	0.19
OI	-2.208*	-2.601**	-2.622**	-2.345**	-2.354**	-3.445***	-2.894**	-1.556	-0.50
	(1.161)	(1.076)	(1.075)	(1.094)	(1.123)	(1.136)	(1.152)	(1.200)	(1.186)
GDPpCap	-2.058***	-2.197***	-2.311***	-2.322***	-2.136***	-2.041***	-1.790**	-1.568**	-1.703*
	(0.679)	(0.691)	(0.703)	(0.707)	(0.703)	(0.691)	(0.705)	(0.744)	(0.778)
credit	3.422***	3.134***	2.715***	2.184**	1.983**	1.577	1.777*	2.470**	3.248*
	(0.909)	(0.863)	(0.883)	(0.925)	(0.969)	(0.981)	(1.031)	(1.199)	(1.287
Ν	370	356	341	324	307	290	272	254	23
$R^2$	0.146	0.159	0.163	0.155	0.157	0.187	0.196	0.188	0.19
Standard Er			0.100	0.100	0.101	0.101	0.150	0.100	0.13

Table A.5: Panel regression results controlling for financial depth. Emerging market economies

="\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.01"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. L0 = lag 0, L1 = lag 1 etc.

L	L7	L6	L5	L4	L3	L2	L1	L0	Variables
-0.307	-0.287*	-0.118	-0.009	0.047	0.270*	0.384***	0.474***	0.576***	LIAB
(0.167)	(0.165)	(0.159)	(0.152)	(0.146)	(0.141)	(0.140)	(0.140)	(0.148)	
0.	0.097	0.077	0.058	0.07	0.134	0.189	$0.269^{*}$	0.310**	GDPpCap
(0.160)	(0.160)	(0.162)	(0.157)	(0.152)	(0.151)	(0.150)	(0.152)	(0.154)	1 1
0.072	-0.110	-0.173	-0.166	-0.195	-0.146	-0.328	-0.369	-0.568	credit
(0.495)	(0.494)	(0.491)	(0.479)	(0.470)	(0.455)	(0.449)	(0.442)	(0.449)	
31	334	355	376	397	418	439	460	480	Ν
0.13	0.129	0.119	0.117	0.117	0.112	0.102	0.097	0.118	$\mathbf{R}^2$
-0.751*	-1.006***	-0.857***	-0.577**	-0.217	0.233	0.594**	1.008***	1.232***	FDI
(0.307)	(0.295)	(0.284)	(0.282)	(0.283)	(0.278)	(0.276)	(0.277)	(0.279)	
0.02	0.038	0.063	0.075	0.080	0.060	0.060	0.089	0.126	GDPpCap
(0.163)	(0.159)	(0.160)	(0.156)	(0.152)	(0.149)	(0.149)	(0.148)	(0.146)	
0.31	0.005	-0.207	-0.298	-0.293	-0.132	-0.134	-0.028	-0.032	credit
(0.486)	(0.474)	(0.476)	(0.473)	(0.473)	(0.462)	(0.449)	(0.440)	(0.434)	
31	334	355	376	397	418	439	460	480	Ν
0.13	0.153	0.142	0.128	0.118	0.105	0.096	0.101	0.126	$\mathbf{R}^2$
0.40	0.431	0.232	-0.122	-0.323	0.00379	0.150	0.229	0.302	PE
(0.347)	(0.340)	(0.338)	(0.291)	(0.287)	(0.277)	(0.276)	(0.262)	(0.247)	
0.07	0.089	0.075	0.064	0.071	0.076	0.110	0.168	0.188	GDPpCap
(0.162)	(0.161)	(0.162)	(0.158)	(0.151)	(0.149)	(0.149)	(0.152)	(0.155)	
0.26	0.126	-0.0898	-0.171	-0.272	-0.194	-0.168	-0.148	-0.174	credit
(0.490)	(0.481)	(0.481)	(0.472)	(0.465)	(0.461)	(0.453)	(0.445)	(0.444)	
31	334	355	376	397	418	439	460	480	Ν
0.12	0.124	0.118	0.118	0.119	0.104	0.086	0.074	0.090	$\mathbf{R}^2$
-0.44	-0.189	0.209	0.602	0.763	0.722	0.300	-0.0881	-0.330	PD
(0.640)	(0.574)	(0.520)	(0.489)	(0.473)	(0.472)	(0.464)	(0.447)	(0.378)	
0.06	0.078	0.098	0.116	0.132	0.128	0.104	0.117	0.110	GDPpCap
(0.164)	(0.165)	(0.167)	(0.162)	(0.158)	(0.153)	(0.149)	(0.147)	(0.148)	
0.02	0.034	0.004	-0.073	-0.275	-0.413	-0.373	-0.280	-0.229	credit
(0.553)	(0.526)	(0.492)	(0.467)	(0.463)	(0.469)	(0.473)	(0.458)	(0.453)	
30	324	345	366	387	408	429	450	470	Ν
0.12	0.124	0.125	0.127	0.127	0.117	0.097	0.080	0.082	$\mathbb{R}^2$
-1.257*	-0.679	0.192	0.398	0.354	0.486	0.680*	0.613*	0.580*	OI
(0.560)	(0.508)	(0.434)	(0.398)	(0.364)	(0.355)	(0.347)	(0.327)	(0.312)	
0.15	0.125	0.071	0.056	0.064	0.109	0.152	0.172	0.141	GDPpCap
(0.166)	(0.163)	(0.162)	(0.156)	(0.151)	(0.151)	(0.150)	(0.150)	(0.148)	
-0.08	-0.067	0.009	-0.006	-0.115	-0.230	-0.420	-0.484	-0.499	credit
(0.516)	(0.501)	(0.492)	(0.475)	(0.466)	(0.455)	(0.454)	(0.453)	(0.456)	
<b>)</b> 30	324	345	<b>3</b> 66	387	408	429	450	470	Ν
0.13	0.129	0.125	0.126	0.123	0.116	0.104	0.088	0.088	$\mathbb{R}^2$

Table A.6: Panel regression results controlling for financial depth. Advanced economies

Standard Errors in parentheses ="\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.001"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. <math>L0 = lag 0, L1 = lag 1 etc.

Variables	LO	L1	L2	L3	L4	L5	L6	L7	L
LIAB	2.020***	1.451***	$1.016^{*}$	0.251	-0.392	-0.970*	-1.081**	-1.090*	-1.532**
	(0.506)	(0.487)	(0.523)	(0.531)	(0.529)	(0.529)	(0.546)	(0.585)	(0.640)
GDPpCap	-1.598**	-1.601**	-1.512*	-1.643**	-1.645**	-1.562*	-1.349	-0.977	-0.77
- r - r	(0.747)	(0.764)	(0.792)	(0.801)	(0.800)	(0.809)	(0.836)	(0.879)	(0.933
capital	-0.148***	-0.116***	-0.095***	-0.066**	-0.050	-0.048	-0.032	-0.030	-0.082
F	(0.030)	(0.029)	(0.030)	(0.030)	(0.031)	(0.032)	(0.033)	(0.035)	(0.047
Ν	354	337	320	302	284	266	248	230	21
$\mathbb{R}^2$	0.177	0.152	0.130	0.115	0.127	0.158	0.189	0.213	0.25
FDI	2.350***	2.478***	2.692***	2.053**	0.686	-0.809	-1.897	-2.186	$-2.991^{\circ}$
	(0.840)	(0.849)	(0.951)	(1.035)	(1.072)	(1.079)	(1.206)	(1.349)	(1.530)
GDPpCap	-1.862**	-1.827**	-1.489*	-1.340*	-1.410*	-1.387*	-1.253	-1.013	-1.114
- r - r	(0.745)	(0.752)	(0.773)	(0.802)	(0.800)	(0.810)	(0.836)	(0.880)	(0.956)
capital	-0.079***	-0.096***	-0.090***	-0.079***	-0.054*	-0.038	-0.017	-0.013	-0.049
F	(0.024)	(0.028)	(0.028)	(0.029)	(0.030)	(0.031)	(0.033)	(0.034)	(0.046
Ν	355	338	321	303	285	267	249	231	21
$\mathbb{R}^2$	0.151	0.151	0.145	0.137	0.128	0.146	0.182	0.209	0.23
PE	0.152	2.128	2.417	4.329	4.199	1.796	-0.265	-4.144	-8.481**
	(2.726)	(2.746)	(2.827)	(2.948)	(3.211)	(3.319)	(3.434)	(3.556)	(3.576)
GDPpCap	-2.128***	-2.065***	-1.887**	-1.781**	-1.540**	-1.296	-1.102	-0.756	-0.39
1 1	(0.753)	(0.762)	(0.773)	(0.773)	(0.781)	(0.802)	(0.835)	(0.880)	(0.937)
capital	-0.092***	-0.088***	-0.082***	-0.067**	-0.054*	-0.045	-0.022	-0.013	-0.03
1	(0.027)	(0.028)	(0.029)	(0.029)	(0.031)	(0.032)	(0.033)	(0.034)	(0.045
Ν	354	337	320	302	284	266	248	230	21
$\mathbb{R}^2$	0.135	0.129	0.121	0.122	0.131	0.146	0.174	0.204	0.25
PD	8.423***	6.097**	4.054	3.429	3.303	1.673	-0.430	-2.455	-3.29
	(2.553)	(2.606)	(2.657)	(2.654)	(2.646)	(2.645)	(2.619)	(2.682)	(2.790)
GDPpCap	-1.309*	-1.439*	-1.612 <sup>*</sup>	-1.624**	-1.438*	-1.305	-1.038	-0.693	-0.46
	(0.767)	(0.795)	(0.819)	(0.824)	(0.826)	(0.821)	(0.827)	(0.854)	(0.905)
capital	-0.100***	-0.092***	-0.081***	-0.063**	-0.054*	-0.049	-0.034	-0.024	-0.05
1	(0.028)	(0.029)	(0.030)	(0.030)	(0.031)	(0.031)	(0.032)	(0.034)	(0.044
Ν	330	313	296	278	260	242	224	206	18
$\mathbb{R}^2$	0.158	0.151	0.142	0.145	0.166	0.208	0.263	0.296	0.30
OI	4.367***	-0.565	-2.079*	-2.670**	-3.051***	-3.926***	-3.174***	-1.747	-0.94
	(1.420)	(1.188)	(1.154)	(1.116)	(1.090)	(1.063)	(1.073)	(1.132)	(1.203)
GDPpCap	-1.316*	-1.840**	-2.088**	-2.021**	-1.623**	-1.297*	-0.852	-0.447	-0.33
	(0.770)	(0.803)	(0.814)	(0.806)	(0.805)	(0.785)	(0.797)	(0.847)	(0.913
capital	-0.162***	-0.079**	-0.062**	-0.048	-0.055*	-0.066**	-0.055*	-0.038	-0.06
T	(0.037)	(0.031)	(0.030)	(0.030)	(0.030)	(0.031)	(0.032)	(0.035)	(0.048
Ν	330	313	296	278	260	(0.001)	(0.002)	206	18
$\mathbb{R}^2$	0.154	0.135	0.145	0.160	0.189	0.256	0.295	0.303	0.30

Table A.7: Panel regression results controlling for capital. Emerging market economies

="\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.001"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. L0 = lag 0, L1 = lag 1 etc.

Variables	L0	L1	L2	L3	L4	L5	L6	L7	L8
LIAB	0.056	0.076*	0.103**	0.123**	0.092	0.074	0.043	0.011	0.002
	(0.043)	(0.045)	(0.050)	(0.058)	(0.058)	(0.059)	(0.059)	(0.059)	(0.060)
GDPpCap	0.562***	0.611***	0.595***	0.557***	0.535***	0.549***	0.627***	0.722***	0.722***
	(0.160)	(0.158)	(0.157)	(0.159)	(0.165)	(0.177)	(0.184)	(0.182)	(0.182)
capital	-0.034**	-0.022	-0.019	-0.025	-0.023	-0.014	-0.007	-0.003	-0.012
	(0.014)	(0.021)	(0.022)	(0.022)	(0.023)	(0.025)	(0.028)	(0.030)	(0.032)
Ν	560	535	508	481	453	425	397	369	341
$\mathbb{R}^2$	0.153	0.142	0.168	0.196	0.212	0.223	0.248	0.285	0.310
FDI	0.065	0.112*	0.157**	0.157**	0.177**	0.139*	0.080	0.041	0.034
	(0.055)	(0.058)	(0.064)	(0.064)	(0.075)	(0.077)	(0.079)	(0.078)	(0.079)
GDPpCap	0.541***	0.574***	0.544***	0.496***	0.491***	0.524***	0.617***	0.718***	0.717***
1 1	(0.159)	(0.157)	(0.156)	(0.158)	(0.165)	(0.177)	(0.184)	(0.183)	(0.181)
capital	-0.034**	-0.023	-0.020	-0.027	-0.025	-0.016	-0.008	-0.004	-0.013
I I	(0.014)	(0.021)	(0.022)	(0.022)	(0.023)	(0.025)	(0.028)	(0.030)	(0.032)
Ν	560	535	508	481	453	425	397	369	341
$\mathbb{R}^2$	0.153	0.144	0.171	0.200	0.218	0.227	0.249	0.286	0.311
PE	0.373	0.301	0.270	0.353	-0.055	0.018	0.319	0.608*	0.639*
	(0.301)	(0.291)	(0.285)	(0.328)	(0.334)	(0.340)	(0.346)	(0.340)	(0.338)
GDPpCap	0.577***	0.605***	0.574***	0.545***	0.517***	0.548***	0.637***	0.730***	0.720***
	(0.161)	(0.158)	(0.157)	(0.160)	(0.166)	(0.177)	(0.184)	(0.181)	(0.180)
capital	-0.035**	-0.023	-0.019	-0.022	-0.022	-0.013	-0.005	-0.001	-0.010
	(0.014)	(0.021)	(0.022)	(0.022)	(0.023)	(0.025)	(0.028)	(0.030)	(0.032)
Ν	560	535	508	481	453	425	397	369	341
$\mathbb{R}^2$	0.153	0.139	0.162	0.190	0.207	0.220	0.249	0.292	0.318
PD	-0.058	-0.307	-0.399	-0.330	-0.687*	-1.021***	-1.126***	-1.206***	-0.990**
	(0.309)	(0.304)	(0.331)	(0.345)	(0.359)	(0.378)	(0.377)	(0.375)	(0.401)
GDPpCap	$0.499^{***}$	$0.571^{***}$	$0.578^{***}$	0.580***	$0.504^{***}$	$0.509^{***}$	$0.577^{***}$	$0.668^{***}$	$0.664^{***}$
	(0.154)	(0.157)	(0.162)	(0.170)	(0.177)	(0.179)	(0.180)	(0.177)	(0.179)
capital	-0.019	-0.023	-0.020	-0.013	-0.015	-0.022	-0.032	-0.058	-0.090**
	(0.023)	(0.023)	(0.024)	(0.026)	(0.028)	(0.031)	(0.034)	(0.036)	(0.039)
Ν	538	513	485	457	429	401	373	345	317
$\mathbb{R}^2$	0.142	0.157	0.186	0.219	0.261	0.298	0.325	0.355	0.376
OI	0.270*	0.316**	0.342**	0.235	0.139	0.136	0.149	0.0564	-0.00240
	(0.143)	(0.147)	(0.164)	(0.177)	(0.180)	(0.185)	(0.192)	(0.200)	(0.210)
GDPpCap	$0.564^{***}$	$0.675^{***}$	0.700***	$0.669^{***}$	$0.639^{***}$	$0.653^{***}$	0.670***	0.701***	0.663***
	(0.157)	(0.161)	(0.163)	(0.163)	(0.164)	(0.172)	(0.179)	(0.180)	(0.181)
capital	-0.019	-0.020	-0.017	-0.010	-0.008	-0.012	-0.022	-0.050	-0.086**
	(0.023)	(0.023)	(0.024)	(0.026)	(0.028)	(0.031)	(0.034)	(0.036)	(0.039)
Ν	538	513	485	457	429	401	373	345	317
$\mathbb{R}^2$	0.149	0.164	0.191	0.221	0.255	0.285	0.308	0.333	0.363

Table A.8: Panel regression results controlling for capital. Advanced economies

="\* p < 0.05 \*\* p < 0.01 \*\*\* p < 0.001"

Each specification includes country and time fixed effects. LIAB = Total external Liabilities, FDI = Foreign Direct Investment, PE = Portfolio Equity liabilities, PD = Portfolio Debt liabilities, OI = Other Investment liabilities. N = Number of observations. L0 = lag 0, L1 = lag 1 etc.

# B. Additional tables

AD	Andorra	$\mathbf{ES}$	Spain	LV	Latvia
AT	Austria	FI	Finland	MT	Malta
AU	Australia	$\mathbf{FR}$	France	NL	Netherlands
BE	Belgium	GB	United Kingdom	NO	Norway
CA	Canada	$\operatorname{GR}$	Greece	NZ	New Zealand
CH	Switzerland	IE	Ireland	$\mathbf{PT}$	Portugal
CY	Cyprus	IS	Iceland	SE	Sweden
DE	Germany	IT	Italy	SI	Slovenia
DK	Denmark	JP	Japan	SK	Slovakia
EE	Estonia	LT	Lithuania	US	United States

Table B.1: AEs (Advanced economies, 30 countries)

Table B.2: EMEs (Emerging market economies, 18 countries)

AR	Argentina	HU	Hungary	RU	Russia
	Brazil		Indonesia		Mexico
$\operatorname{CL}$	Chile	IL	Israel	MY	Malaysia
CN	China	IN	India	$\mathrm{TH}$	Thailand
CO	Colombia	$\mathbf{KR}$	South Korea	$\mathrm{TR}$	Turkey
CZ	Czech Republic	PL	Poland	ZA	South Africa

Table B.3: Data sources

Variables	Data sources
Dependent variable	
Gini net	SWIID Version 5.1, Solt (2016)
<u>Financial measures</u>	
LIAB	Lane and Milesi-Ferretti (2017)
FDI	Lane and Milesi-Ferretti (2017)
PE	Lane and Milesi-Ferretti (2017)
PD	Lane and Milesi-Ferretti (2017)
<u>Control variables</u>	
GDPpCap	World Economic Outlook, International Monetary Fund (2017)
credit	Bank for International Settlements (2017b)
trade	World Development Indicators, The World Bank (2017)
unemployment	World Development Indicators, The World Bank $(2017)$