ISSN 1833-4474



The Making of Pro-poor Growth

Fabrizio Carmignani School of Economics, The University of Queensland, QLD 4072, Brisbane, Australia f.carmignani@uq.edu.au.

> THIS DRAFT September 23, 2009

Abstract

A system of three endogenous equations is used to estimate the determinants of poverty eduction. The system incorporates: (i) the direct effect of growth and income inequality on poverty, (ii) the feedback effect of poverty on inequality and growth, and (iii) different channels through which economic policies can affect poverty reduction. Results indicate the existence of a virtuous and a vicious equilibrium. The policy mix is then critical in deciding to which of the two equilibria a country converges.

JEL classification: C30, E60, I32, O11

Keywords: poverty, growth, inequality, system of equations, macroeconomic policy.

The making of pro-poor growth

Fabrizio Carmignani School of Economics, The University of Queensland, QLD 4072, Brisbane, Australia f.carmignani@uq.edu.au.

September 23, 2009

Abstract

A system of three endogenous equations is used to estimate the determinants of poverty reduction. The system incorporates: (i) the direct effect of growth and income inequality on poverty, (ii) the feedback effect of poverty on inequality and growth, and (iii) different channels through which economic policies can affect poverty reduction. Results indicate the existence of a virtuous and a vicious equilibrium. The policy mix is then critical in deciding to which of the two equilibria a country converges.

JEL classification: C30, E60, I32, O11

Keywords: poverty, growth, inequality, system of equations, macroeconomic policy.

1 Introduction

On average, economic growth reduces poverty, and there is a voluminous body of empirical research to support this statement. However, the dispersion around the average is large: the elasticity of poverty with respect to economic growth varies considerably across countries and over time. The challenge for development economists and policymakers is to understand to what extent economic policies contribute to making growth more or less pro-poor.¹ This paper takes up the challenge from a macroeconometric perspective: using a relatively large sample of countries, a system of three endogenous equations is estimated in order to disentangle the different channels through which economic policy variables affect poverty dynamics. The specific policy variables that are investigated are meant to capture some of the key dimensions of the economic policy space of

¹For the purpose of this paper, saying that growth is more or less pro-poor is equivalent to saying that poverty is more or less responsive to economic growth. Responsiveness is in turn associated with a notion of elasticity. Letting Δy and Δp denote the percentage change in per-capita income and poverty headcount respectively, the marginal effect of Δy on Δp gives a measure of the responsiveness of poverty to growth.

developing countries: openness to international trade, macroeconomic stability, depth of financial intermediation, size of government, and education.

There is a fast growing literature on the relationship between growth and poverty. Evidence of a significant poverty-reducing effect of economic growth goes back to Fields (1989) and Squire (1993), who estimate income elasticities of poverty reduction well in excess of two. Similar findings have been reported by Bruno et al. (1998) and Adams (2004). In a seminal paper, Dollar and Kraay (2002) show that the elasticity of mean income of the poor with respect to average mean income is not significantly different from one, meaning that growth is unambiguously good for the poor. Kraay (2006) provides additional evidence that most of the cross-country variation in poverty changes is due to the variation in the rate of average incomes growth. Ravaillon (1997, 2001 and 2004), Bourguignon (2003), Epualard (2003), Kawkani et al. (2004), Mosley et al. (2004), Kalwij and Verschoor (2007) stress that the growth elasticity of poverty significantly differ across countries and that these differences are largely explained by the interaction between growth and income inequality. In a recent contribution, Loyaza and Raddatz (2009) take a complementary perspective and show that the sectoral composition of growth affects its capacity to reduce poverty.

This paper extends the existing literature by exploring the role of economic policies in shaping the growth-inequality-poverty relationship. The theoretical prior of the paper is that economic policies can affect poverty in three separate ways: (i) by determining the rate of economic growth, (ii) by determining the dynamics of income inequality, and (iii) by determining the responsiveness of poverty to growth for any given rate of growth and change in inequality. These three channels are jointly estimated by using a system of three endogenous equations. The paper is therefore related to the work of Lundberg and Squire (2003) and Huang et al. (2009) on the simultaneous evolution of growth and inequality, although the addition of a poverty equation represents a significant innovation relative to previous studies. The use of a system of equations also makes the paper quite different from the analysis of Chhibber and Nayyar (2008), who go about exploring the impact of economic variables on the growth elasticity of poverty within a single equation framework.

The key findings of this paper can be summarized as follows. First, growth and redistribution reduce poverty. However, the growth elasticity of poverty is not affected by changes in income inequality. Second, there is no trade-off between growth and redistribution. In fact, the two processes seem to reinforce each other. Third, there is strong evidence of a feedback effect of poverty on both growth and inequality: an increase in the poverty headcount causes both a slowdown in growth and an increase in inequality. Fourth, economic policies determine poverty dynamics through multiple channels. Financial development and bigger government size contribute to poverty reduction via a positive effect on growth. This contribution is however offset, at least to some extent, by a contemporaneous inequality-increasing effect. Furthermore, financial development also appears to reduce the responsiveness of poverty to growth. Macroeconomic stabilization (here represented by a lower rate of inflation) promotes poverty reduction in two ways: by fostering growth and by making poverty more pro-poor. The role of trade openness and education appears to be limited to their effect on the responsiveness of poverty to growth. More specifically, an increase in trade openness makes poverty less responsive to growth while the opposite is true for education. Taking stock of these results, one can conclude that countries tend to move towards one of two possible equilibria. The positive (virtuous) equilibrium is characterised by faster growth, redistribution, and rapid poverty reduction. The negative (vicious) equilibrium involves slow (or even negative) growth, sharpening inequalities, and resilient poverty. The policy mix is then critical in determining to which of the two equilibria the country will converge.

The rest of the paper is organized as follows. Section 2 presents the data and some preliminary stylized facts. Section 3 introduces the econometric model and the estimation methodology. Results are reported in Section 4. Section 5 discusses the role of growth volatility in explaining why some economic policies make growth less pro-poor. Section 6 draws some policy conclusions. The description of the variables and the full list of countries/periods included in the dataset are provided in the Appendix.

2 Data and stylized facts

The source of poverty data is the PovcalNet Poverty Analysis Tool of the World Bank (www.http://iresearch.worldbank.org/PovcalNet/jsp/index.jsp). Let the sequence (t_n) denote the years when the data on a poverty indicator p are available for a generic country i. The dataset used in this paper consists of observations taken over spells determined from the sequence (t_n) . Thus, for country i, the first spell starts at t_1 and ends at t_2 ; the second spell starts at t_2 and ends at t_3 ; and so forth until all data points are exhausted. The only additional restriction is that the minimum duration of a spell must be five years. In this way it is possible to filter out short-term dynamics and focus on long-term effects. In total, after the exclusion of some outliers, 151 spells, covering 83 countries, are identified. All the variables, with the exception of a set of country-fixed effects (see below), are expressed as annualized percentage changes over each spell. In addition, the dataset includes the initial values of all variables at the beginning of each spell plus their lagged annualized percentage change over the five years prior to the beginning of the spell. Initial and lagged values turn out to be useful instruments in the estimation process.

The poverty indicator used as a reference is the poverty headcount at 1 dollar per-day $(ph_\$1)$. Income is measured by per-capita GDP at constant prices and its annualized percentage change is of course the rate of economic growth (y_pc) . Inequality is proxied by the *Gini coefficient* (gini), so that a negative percentage change in gini corresponds to redistribution. The policy variables are as follows: (i) imports plus exports in percent of GDP to measure openness to international trade (trade), (ii) domestic credit to the private sector in percent of GDP to proxy for domestic financial depth (credit), (iii) the consumer price index (cpi) to capture macroeconomic stability, (iv) government expenditure in percent of GDP (gov) to measure the size of the government, and (v) the average number of years of schooling in the population (tyr) to proxy for the accumulation of human capital. Finally, the country-fixed effects are: distance from the equator (lat_abst) , French legal origins $(legor_fr)$, and ethnic fragmentation (ethnic). As discussed below, these country fixed effects will be used in some equations to account for the role of institutional quality.

Table 1 reports some key summary statistics for $ph_\$1$, y_pc , and the ratio of $ph_\$1$ to y_pc . Given that the variables are expressed as annualized percentage changes, their ratio can be effectively interpreted as the growth elasticity of poverty. Two sets of summary statistics are provided: one for the full sample of 151 spells and one for a restricted sample of 71 spells that are associated with a negative change in the Gini coefficient (that is gini < 0).

INSERT TABLE 1 ABOUT HERE

The average change in poverty is negative, but significantly more so when the sample is restricted to spells with decreasing inequality. The difference in the average value of the growth elasticity of poverty between the full and the restricted sample is however small. Therefore, redistribution appears to alleviat poverty directly rather than by conditioning the responsiveness of ph \$1 to growth. The rate of economic growth is on average positive and slightly higher in the restricted sample, thus suggesting that redistribution and growth are not necessarily two mutually exclusive processes. The standard deviations of both $ph \$ \$1 and $ph \$ \$1/y pc are quite large. This wide dispersion around the averages is what motivates the study of the determinants of cross-country variation in the reposnsiveness of poverty to growth. Another interesting characteristic of ph \$1 and ph $\frac{1}{y}$ pc is that they occasionally display very large values (as the maximum and minimum statistics indicate). These large values tend to occur in association with initially low values of the poverty headcount. In this sense, there seems to be a mild positive correlation between the initial level of the poverty indicator and the absolute value of the elasticity. Finally, while the growth elasticity of poverty is negative on average, in 49 spells out of 151 it is actually positive. These are all spells where inequality and per-capita income change in the same direction, but the change in inequality is so large that it more than offsets the effect of growth.

To uncover some empirical regularities in the data, table 2 reports a few basic correlations. The coefficients shown in the table are obtained from simple OLS regressions and hence they are not necessarily indicative of the direction of causality.

INSERT TABLE 2 ABOUT HERE

The coefficients in columns I and II indicate that there is a positive association between poverty reduction and growth and between poverty reduction and redistribution. However, the lack of significance of the coefficients in columns III and IV suggests that: (i) the relationship between growth and poverty reduction is linear (i.e. the absolute value of the growth elasticity of poverty does not increase as the rate of growth increases) and (ii) the responsiveness of poverty to growth is not necessarily affected by the interaction between growth and redistribution. The coefficients in the last two columns indicate that there is no trade-off between growth and redistribution. In fact, both the inequality elasticity of growth and the growth elasticity of inequality are negative. This is certainly good news given that both growth and redistribution seem to contribute to poverty reduction.

3 Econometric model

Mathematically, a change in poverty is the combination of a change in average per-capita income and a change in the distribution of incomes around the average. Following Bourguignon $(2003)^2$, let z denote the poverty line drawn at 1 dollar per-day and F(y) the cumulative distribution function of per-capita income in a given country. The poverty headcount at time t is therefore defined as $p_t = F_t(z)$. The change in the poverty headcount between t and t + n is then given by $\Delta p = F_{t+1}(z) - F_t(z)$. Let then $G(\cdot)$ denote the distribution of income after normalizing for average income y_m . Any change in $F(\cdot)$ is the algebirc sum of (i) a change in $G(\cdot)$, holding y_m constant and (ii) a change in y_m that leaves $G(\cdot)$ unchanged. The change in $G(\cdot)$ is the "distribution component" of Δp and the change in y_m is the "growth component".

The above decomposition suggests that economic policies affect poverty via their contribution to changes in y_m (i.e. their contribution to economic growth) and $G(\cdot)$ (i.e. their contribution to redistribution). In addition, economic policies might shape poverty dynamics by making poverty more or less responsive to a given change in y_m for any given change in $G(\cdot)$. The theoretical rationale underlying this additional effect draws on the work of Vinod et al. (2000) on the quality of growth. The quality content of growth depends on the change in the value of assets like human and social capital and on the stability the growth process. For any given change in inequality, the decrease in poverty associated with a given rate of growth is likely to be higher the better the quality of this growth. Therefore, to the extent that they are able to promote better (or worse) quality of growth, economic policies will strengthen (or weaken) the responsiveness of poverty to growth.³

All in all, the effects of policies on poverty are transmitted via: (i) economic growth, (ii) changes in inequality, and (iii) changes in the responsiveness of poverty to growth. Accordingly, the econometric model consists of three endogenous equations:

$$y_pc_s = \alpha_0 + \alpha_1 gini_s + \alpha_2 ph_\$1_s + A_3V_s + A_4W_s + \varepsilon_s \tag{1}$$

 $^{^{2}}$ See also Datt and Ravaillon (1992), Kakwani (1993), and Kraay (2006)

 $^{^{3}}$ Section 5 shows that most of the effect of policies on the responsiveness of poverty to growth is explained by their effect on the volatility of the growth process.

$$gini_s = \beta_0 + \beta_1 y_p c_s + \beta_2 ph_\$ 1_s + B_3 V_s + B_4 Z_s + v_s \tag{2}$$

$$ph_{\$1_{s}} = \gamma_{0} + \gamma_{1}y_{pc_{s}} + \gamma_{2}gini_{s} + \gamma_{3}gini_{s} * y_{pc_{s}} + C_{3}(y_{pc_{s}} * V_{s}) + \omega_{s}$$
(3)

where y pc, gini, ph \$1 are the annualized percentage changes in per-capita income, Gini index, and poverty headcount respectively, ε, v, ω are the error terms, and s indicates the generic spell in the dataset. V is a vector that includes the economic policy variables tyr, credit, trade, cpi, and qov (all expressed in annualized percentage changes). W is a vector of controls that includes the values of per-capita income and Gini coefficient at the beginning of each spell. Initial per-capita income $(i \ y \ pc)$ is meant to account for conditional convergence effects while the initial value of the Gini index (*i Gini*) captures the effect of given levels of inequality on subsequent growth dynamics. Z is a vector that includes the initial value of the Gini index $(i \ gini)$ and the country fixed effects legor fr, lat bast, and ethnic. The logic underlying the inclusion of these country fixed effects draws on Carmignani (2009): they represent structural determinants of institutional quality that are important in shaping the distribution of income over time. Finally, the parameters to be estimated are $\alpha_0, \alpha_1, \alpha_2, \beta_0, \beta_1, \beta_2, \gamma_0, \gamma_1, \gamma_2, \gamma_3$ plus the vector of coefficients $A_3, A_4, B_3, B_4, \text{and } C_3.$

Two main features of the above specification are worth a mention. First, the system formalizes the three transmission channels of the effect of policies on poverty dynamics: A_3 captures the effect via growth, B_3 the effect via changes in inequality, and C_3 the effect via the responsiveness of poverty to growth. Note that in equation (3), $y_pc_s * V_s$ is in fact a vector of interactive terms between growth and the policy variables, so that the coefficients C_3 measure the contribution of policy variables to the partial derivative of ph \$1 w.r.t. y_pc . Because γ_1 is going to be negative, a positive coefficient on the interactive term means that higher values of the policy variable reduce the responsiveness of poverty to growth. A similar interpretation holds for the coefficient on the interactive term $gini_s * y_pc_s$: if $\gamma_3 > 0$, the an increase in inequality lowers the responsiveness of poverty to growth. The second important feature of the system is that it allows for endogenous relationships between growth, changes in poverty, and changes in inequality as each of these three variables appears on the r.h.s. of the regression of the other two variables. In other words, the system incorporates the endogenous evolution of growth and changes in inequality (slope coefficients α_1 and β_1), the direct effects of growth and changes in inequality on poverty (slope coefficients γ_1 and γ_2), and the feedback effect of poverty dynamics on growth and changes in inequality (slope coefficients α_2 and β_2). The reason why this feedback effect might be important is that the poor are most likely to be socially excluded and economically marginalized from the rest of the society. As a consequence, an increase in poverty is expected to result in larger income inequalities and a smaller growth potential.

The system is estimated using a GMM system estimator (Wooldridge, 2003). The system estimator accounts for cross-correlations among the error terms of the three equations. At the same time, the system estimator is not immune from problems. In particular, if one of the three equations were misspecified, then the estimates of all the equations would be negatively affected. It is therefore wise to use equation-by-equation estimates to test the robustness of system estimates. The GMM system estimates will be then accompanied by a set of 2SLS estimates of each equation taken separately from the others. These equation-by-equation estimates are also useful to generate first stage diagnostics on the validity of the instruments chosen for the policy variables (see below).

A complication arising in the estimation of the system of equations (1), (2),and (3) is that the changes in the economic policy variables are likely to be endogenous to the dependent variables. For this reason, the economic policy variables (and the interactive terms in equation (3)) have to be instrumented. To this purpose, the time dimension of the dataset is exploited and lagged and initial values of all the potentially endogenous variables are used as instruments. In the case of interactive terms, the lagged and initial values are obtained from the interaction of the lagged and initial values of the individual terms. Country fixed effects and initial values of per-capita income and inequality are taken to be pre-determined and hence they are instrumented by themselves. This choice of instruments is then submitted to a battery of tests. To test the exogeneity of instruments, the Sargan test of overidentifying restrictions is used. The null hypothesis that the overidentifying restrictions implied by the choice of instruments are valid is never rejected at usual confidence levels. The relevance of instruments is assessed using first stage diagnostics from the equation-byequation estimation. The standard partial R^2 is in the range of 0.5 to 0.7 and the p-value of the F-test of joint significance of the excluded instruments is always below 0.1. The Shea's partial R^2 is slightly lower than the standard partial R^2 , but still in the range of 0.4 to 0.6. As discussed in Baum et al. (2003 and 2007), these values indicate that the chosen instruments are indeed relevant. All in all, the tests seem to suggest that the instruments are valid.⁴

4 Results

4.1 System estimates

System estimates are presented in Table 3. Column I reports the estimates of the growth equation (1), column II the estimates of the inequality equation (2), and column III the estimates of the poverty equation (3).

INSERT TABLE 3 ABOUT HERE

 $^{^{4}}$ The Sargan statistics and the associated p-values are reported at the bottom of the tables in the next section. The first stage diagnostics for each endogenous regressor can instead be obtained from the author upon request.

To start with, consider the endogenous relationship between growth, changes in inequality, and changes in poverty. The negative coefficient on y_pc and the positive coefficient on gini in column III imply that a combination of growth and redistribution provides the most favorable scenario for poverty reduction. At the same time, there is no evidence of a trade-off between growth and redistribution. In fact, the negative coefficients on gini in column I and on y_pc in column II indicate that redistribution favors growth and vice-versa. Interestingly, however, the lack of significance of the coefficient on the interactive term $y_pc * gini$ in column III suggests that the change in inequality does not affect the growth elasticity of poverty. Finally, changes in poverty significantly feed back onto growth prospects and inequality dynamics. The signs of the coefficients on $ph_$ \$1 in columns I and II confirm the simple intuition that an increase in the poverty headcount reduces growth and increases inequality.

Turning to the effect of policy variables, the evidence is that multiple channels operate at the same time. For government size and financial intermediation two effects of opposite sign seem to be at work. On the one hand, both *gov* and *credit* contribute to poverty reduction by strengthening growth (column I). On the other hand, both variables also cause larger inequalities (column II). Moreover, the positive coefficient on $y_pc*credit$ in column III implies that credit deepening reduces the responsiveness of poverty to economic growth. Macroeconomic stability (here identified by negative - or marginally positive - changes in the consumer price index) reduces poverty in two complementary ways: (i) by strengthening growth and (ii) by increasing the growth elasticity of poverty. Trade openness and education do not seem to produce any significant growth or redistribution effect. However, they affect poverty via their effect on its elasticity with respect to growth. More specifically, greater openness to trade makes poverty less responsive to growth, while wider eduction increases responsiveness.

Some of the above findings deserve a more explicit interpretation. The inequality-increasing effect associated with gov indicates that broad government expenditure is not well targeted to the poor. Hence, narrower and better targeted spending programs are required to reduce inequalities in developing countries. With respect to *credit*, one would probably expect deeper credit to reduce inequalities. However, in most developing economies, the rich have preferential access to credit. Therefore, an increase in the volume of credit might well widen the distance between rich and poor. Finally, the coefficients on the interactive terms between growth and policies in equation (3) are consistent with the view that policies make poverty more or less responsive to growth by affecting the quality of growth. Education is expected to improve the quality of growth by increasing the value of human and social capital. On the contrary, trade openness, macroeconomic instability, and financial deepening are likely to reduce quality of growth by making growth itself more volatile (see (5). The estimated coefficients then confirm that education makes growth more propoor while openness, macroeconomic instability, and financial deepening make growth less pro-poor.

The last bits of evidence concern the role of the other control variables in the growth equation (column I) and inequality equation (column II). In the growth

equation, there is no evidence of conditional convergence. In fact, the positive coefficient on i y pc suggests that initially richer countries tend to grow faster. The estimated coefficient on the initial level of inequality (i gini) is instead negative, meaning that initially more unequal societies tend to experience slower growth. Note however that these societies do not necessarily fall into a poverty trap given that they can still undertake redistribution. Redistribution would then allow them to strengthen growth and reduce poverty at the same time. In the inequality equation, French legal origins and proximity to the equator tend to make the distribution less equal. The interpretation of this finding follows from la Porta et al. (1999). French legal origins and proximity to the equator tend to generate worse institutions. Bad institutions then sharpen income inequality (Carmignani, 2009). Finally, a higher initial inequality makes redistribution more difficult, possibly through some hysteresis effect. Combined with the evidence on the effect of i_gini from the growth equation, this result indicate that the progress in curbing inequalities and poverty might be initially slower in countries characterized by a high Gini coefficient. However, once initial gains are made, the process will be self-fulfilling.

4.2 Single equation estimates and robustness checks

As discussed in Section (3), it is important to check whether results are sensitive to the use of a system estimator vs. an equation-by-equation estimator. To this purpose, two stages least squares estimates of the three equations are reported in Table 4. The information in Table 4 is organized in the same way as in Table 3. The only addition is the R^2 of each equation to have an idea of the goodness of fit. It is immediately clear that the main results obtained from the system estimates are confirmed. Most of the variables retain their sign and significance. Only two notable exceptions emerge, both in the inequality equation. First, the feedback effect of changes in poverty on changes in inequality is no longer significant at usual confidence levels. Second, education now plays a perverse role as it increases inequalities.⁵ Because the sign on the interactive term y pc *tyr is still negative and significant, one can argue that wider education reduces absolute poverty while increasing relative poverty. In other words, an increase in average years of schooling makes both the rich and poor better off, but the income opportunities of the rich increase proportionally more than the income opportunities of the poor.

INSERT TABLE 3 ABOUT HERE

To test the robustness of the system estimates, the following sensitivity checks are also conducted.⁶ First, the interactive term y_pc*i_gini is added to the specification of the poverty equation. The rationale for including this interactive term follows from the mathematical decomposition of changes in

⁵This result is not new in the literature, see for instance Carmignani (2009).

 $^{^6}$ These additional estimates are not reported as they do not imply any qualitative change in the main result of the paper. Nevertheless, they are available from the author upon request.

the poverty rate, when the assumption is made that the distribution of income is log-normal (see Bourguignon, 2003). The coefficient of this new interactive term is expected to be positive, meaning that the growth elasticity of poverty is lower the more unequal income distribution initially is. It turns out that y pc * i gini is largely insignificant. All of the other coefficients of the system are qualitatively the same as those reported in Table 3. Second, the initial level of per-capita income i y pc is added to the inequality equation. Again, the estimated coefficient is largely insignificant, while nothing else in the system changes. To allow for possible non-linearities, a specification including both i y pc and $(i y pc)^2$ is also estimated, but neither of the two terms appear to be statistically significant. Third, the country fixed effect legor_fr, lat_abst, and *ethnic* are added to the specification of the growth equation. Neither of the three is statistically significant. At the same time, i_y_pc and i_gini in the same equation become insignificant. This might be a symptom of multicollinearity, as the three country fixed effects might actually determine income and inequality levels via their effect on institutional quality. When i y pc and i gini are removed from the growth equation, then both legor fr and ethnic display a statistically significant negative coefficient. Again, this result can be interpreted as evidence of the adverse effect of bad institutions on growth.⁷

Finally, to maximize the number of observations in the sample, the variable tyr and the associated interactive term y_pc*tyr are dropped from the system. As a matter of fact, tyr is not available for all 151 spells of the original dataset, and that is why the number of observations per equation is only 90. When dropping tyr it is possible to estimate the system on 120 observations per equation, for a total of 360 system observations. Relative to the estimates in table 3, estimated coefficients retain their sign, but occasional changes in their level of statistical significance are observed. The effect of government expenditure on growth and inequality is no longer significant. However, gov_cons now affects poverty by reducing its responsiveness also become statistically insignificant, so that in the end *credit* only affects poverty via a positive growth-effect. Finally, the effect of macroeconomic stability on the responsiveness of poverty becomes insignificant, while its effect on growth remains positive and significant.⁸

⁷Ideally, one would like to add a direct measure of institutional quality to the system. To construct a proxy that matches the irregular frequency and length of the spells in this dataset, annual data on institutions would be needed over the period 1980-2005 for all (or most of) the developing countries in the sample. However, this type of data is hardly available. An interesting possibility that will deserve attention in future work is to use "contract intensive money" (Clague et al. 1999) to proxy for institutional quality. Being constructed from widely available money aggregate data, contract intensive money can be computed over spells of irregular frequency and length.

⁸ The system without tyr and $tyr * y_pc$ is also re-estimated on a sub-sample that excludes transition economies to make sure that results (particularly those concerning the endogenous relationship between growth and redistribution) are not driven by the dynamics of the transformation from plan to market. No major changes relative to the estimates in Table 2 are observed.

5 Discussion: policies and the volatility of growth

The previous section offers some intriguing evidence on the role of financial intermediation. The positive coefficient on *credit* in equation (1) means that financial deepening strengthens growth, and this is indeed a rather common result in the growth literature. However, the positive coefficient on the interactive term *credit* * y_pc in equation (3) means that financial deepening also makes poverty less responsive to growth. How could this latter finding be explained? This section puts forward the following explanation: (i) more volatile growth tends to be less pro-poor and (ii) financial deepening increases growth volatility. It will turn out that the volatility channel also explains why trade openness and inflation make poverty less responsive to growth.

Intuitively, volatility hurts the poor more than the rich as the poor generally have less means to cope with the adverse consequences of sharp cyclical fluctuations. For instance, the poor might have very limited access to consumption smoothing technologies or their jobs might be less secure in a cyclical downturn. It then follows that higher volatility makes poverty less elastic with respect to growth. An empirical test of this conjecture can be simply performed by adding the interaction between growth and volatility to the standard model in column II of Table 2. Letting *vol* denote the standard deviation of the annual growth rate over each spell, then the estimated coefficients in the extended standard model are as follows (p-values in brackets):

$$ph_{\$1} = \underbrace{1.210}_{(0.46)} + \underbrace{-2.767 * y_pc}_{(0.000)} + \underbrace{2.414 * gini}_{(0.000)} + \underbrace{0.225 * (y_pc * vol)}_{(0.037)}$$
(4)

The positive coefficient on the interactive term y_pc*vol implies that higher values of vol reduce the marginal effect of y_pc on $ph_$ \$1. The inclusion of volnot interacted with growth does not change the results (and vol alone turns out to be largely insignificant). There is thus evidence that more volatile growth tends to be less pro-poor. The implication is that economic policy variables that increase volatility will also reduce the growth elasticity of poverty.

The question is then whether or not financial deepening is positively associated with growth volatility. In fact, a more developed financial system should allow for better consumption smoothing, even though this does not necessarily mean that output fluctuations will also be smoother. Moreover, if the expansion of credit occurs in waves, then boom-bust cycles would generate both higher average growth and higher growth volatility. A volatility equation is then estimated and the results are as follows:

$$vol = 2.427 + 0.023 * credit + 0.043 * cpi + 0.031 * trade \tag{5}$$

In addition to *credit*, the equation includes two other policy variables that are likely to affect volatility. The positive coefficient on *cpi* indicates that there is a positive association between macroeconomic instability and output volatility. The positive coefficient on *trade* instead suggests that more open economies are probably more exposed to external shocks. The effect of *credit* on volatility is also positive, in line with the credit-cycles argument.⁹

Taken together, equations (4) and (5) tell an interesting story: higher inflation, greater trade openness, and credit deepening make growth less pro-poor because they tend to increase the volatility of growth. Thus, the volatility channel explains the positive estimated coefficients on the interactive terms $y_pc * cpi, y_pc * credit, y_pc * trade$ reported in Tables 3 and 4. Also note that the lack of significance of gov_cons in the volatility equation (see 5) might explain why the interactive term $y_pc * gov$ is not statistically significant in the system estimates. Finally, growth volatility does not seem to be the relevant channel to explain the effect of education on the responsiveness of poverty to growth. As discussed above, this effect is in fact better understood in terms of redistribution of opportunities and value human capital.

6 Policy analysis and conclusions

The paper employs a system of three endogenous equations to study the determinants of absolute poverty dynamics. A key feature of the system is that it allows economic policy variables to affect poverty via separate effects on (i) the rate of economic growth, (ii) the change in income inequality, and (iii) the responsiveness of poverty to any given rate of growth. The system also incorporates a feedback effect of poverty dynamics on growth and inequality. The main results can be summarized as follows. First, the elasticity of poverty is negative with respect to growth and positive with respect to changes in income inequality, meaning that a combination of growth and redistribution represents the most favorable scenario for poverty reduction. Importantly, there is no evidence of a trade-off between growth and redistribution. If anything, the two processes appear to reinforce each other. However, the growth elasticity of poverty is not significantly affected by changes in inequality. Second, the feedback effect of poverty on growth and inequality is significant: an increase in poverty slows growth down and sharpens income inequalities. Third, economic policy variables affect poverty in different ways through different channels. Higher government expenditure and deeper level of financial intermediation generate two effects of opposite sign: they strengthen growth, thus reducing poverty, but they also sharpen inequality. Macroeconomic stability also determines a positive growth effect, without however causing any adverse effect on inequality. Economic policy variables have also significant effects on the extent to which any given rate of growth is more or less pro-poor. Greater openness to international trade, higher inflation, and deeper financial intermediation tend to increase the volatility of output growth and in this way the contribute to making poverty less elastic

⁹The equation is estimated by two stages least squares. The Hausman test indicates that cpi is potentially endogenous and therefore it is instrumented by its lagged and initial values. When added to the volatility equation, both gov and edu turn out to be largely insignificant. The inclusion of the level of per-capita GDP does not qualitatively change the estimated coefficients on *credit*, *trade*, and *cpi*.

with respect to growth. On the contrary, wider education increases the value of human capital in the economy, improves income opportunities, and therefore makes poverty more responsive to growth.

From a policymaking perspective, the empirical results highlight the importance of choosing the appropriate policy mix to fight poverty. The endogenous links between growth, redistribution, and poverty reduction imply that there exist two possible equilibria to which countries can converge. In the virtuous equilibrium, a country achieves a scenario of fast growth and redistribution. These two processes strengthen each other and reduce poverty. In turn, lower poverty improves growth prospects and reduces inequalities. In the vicious equilibrium, the country is unable to achieve the growth and redistribution scenario and hence poverty is not significantly reduced. Persistently high poverty in turn depresses growth and causes large inequalities, thus making the achievement of the favorable scenario even less likely. The policy mix then plays a critical role in determining which of the two equilibria the country converges to. The evidence discussed in the previous sections provides some indications on how the ideal policy mix (that is, the policy mix leading to the virtuous cycle) ought to be designed:

- Low and stable inflation should be set as a priority in the macroeconomic policy framework of developing countries. However, the framework should also be flexible enough to provide policymakers with the policy space required to take a truly counter-cyclical policy stance. In fact, a counter-cyclical macroeconomic policy would reduce the volatility of output growth, thus contributing to making poverty more elastic with respect to growth. In this respect, low inflation should be established as a mediumterm target, thus allowing actual inflation to overshoot (or undershoot) the target in the short term to respond to cyclical shocks. A counter-cyclical macroeconomic policy that reduces volatility will also maximize the contributions of domestic financial development and trade liberalization to poverty reduction.
- Government size should not be reduced as it can be an important driver of growth. However, the scope of the government must be adjusted in view of the need to promote the redistribution of incomes. Government consumption must be therefore better targeted towards the poor. In this sense, subsidies and transfers appear to be rather effective in lowering the Gini coefficient (see for instance Carmignani, 2009). Therefore, budgetary allocations to these categories of public expenditure should be increased.
- The public supply of education services should be increased as wider education makes growth more pro-poor. The government should make sure that the delivery of education translates into an effective improvement of opportunities for the poor. This in turn calls for greater focus on "quality" issues, such as the update of curricula, the formation of teachers in rural areas, and the supply of teaching and study material.

Acknowledgement 1 I would like to thank Abdour Chowdhury, Desire Avom, and Stuart McDonald for helpful discussion. I am solely responsible for any remaining error and inconsistencies.

Appendix Variables description

Unless otherwise indicated, all of the variables are expressed in *annualized* percentage changes. WDI is the World Development Indicators Database of the World Bank, 2008 issue.

- *ph_\$1*, poverty headcount: Proportion of population living on less than 1 dollar per day. Source: PovcalNet, World Bank.
- y_pc , per-capita income: real per-capita GDP at constant US dollars (base year 2000). Source: WDI.
- *gini*, Gini coefficient of inequality of income distribution. Source: Povcal-Net, World Bank.
- *gov*, government size: total government expenditure in percent of GDP. Source: WDI.
- *credit*, financial depth: domestic credit to the private sector in percent of GDP. Source: WDI.
- *trade*, international trade openness: exports plus imports in percent of GDP. Source: WDI.
- cpi, consumer price index. Source: WDI.
- *edu*, education: number of years of schooling of the average individual in the population. Source: Barro and Lee (2004) and UNESCO.
- *legor_fr*, French legal origin: dummy variable taking value 1 if country's legal system originates from the French civil code. Source: La Porta et al. (1999). This variable is expressed in levels.
- *ethnix*, ethnic fragmentation: probability that two randomly selected individuals are not from the same ethnic group. Source: La Porta et al. (1999). This variable is expressed in levels.
- *lat_abst*, distance from equator. Source: La Porta et al. (1999). This variable is expressed in levels.
- vol, growth volatility: standard deviation of y_pc over a given spell. Source: WDI
- i_y_pc , real per-capita GDP at the beginning of a given spell. Source: WDI. This variable is expressed in log-levels.
- *i_gini*,Gini coefficient at the beginning of a given spell. Source: WDI. This variable is expressed in log-levels.

List of countries and spells in the dataset

Albania 97 02 Algeria 88_95 Argentina 86_92 Argetnina 92_98 Argentina 98_03 Armenia 96_01 Azerbaijan 95 01 Bangladesh 84 89 Bangladesh 89 96 Belarus 88_93 Belarus 93_98 Bolivia 91 97 Bolivia 97 02 Botswana 86_94 Brazil 81 87 Brazil 87 92 Brazil 92 97 Brazil 97 02 Bulgaria 88_93 Bulgaria 93 98 Bulgaria 98 03 Burundi 92_98 Cameroon 96_01 Chile 87_92 Chile 92_{98} China 87_93 China 93_99 Colombia 81 88 Colombia 88_95 Colombia 95 00

Costa Rica 81 86 Costa Rica 86_{93} Costa Rica 93_98 Cote d'Ivoire 85_93 Cote d'Ivoire 93 98 Crotia 88_98 Czech Rep. 88_93 Dominc. Rep. 86 92 Dominic. Rep. 92 00 Ecuador 87_94 Egypt 91 00 El Salvador 89 95 El Salvador 95 00 Estonia 88_93 Estonia 93 98 Estonia 98 03 Ethiopia 82 95 Ethiopia 95 00 Gambia 92 98 Georgia 96 01 Ghana 88 98 Guatemala 87_98 Guyana 93_98 Honduras 86 92 Honduras 92_{98} Honduras 98_03 Hungary 87_93 Hungary 93 98 India 87_{93} India 93_{00}

Indoneisa 87 93 Indonesia 93_98 Iran 86_{94} Jamaica 88_93 Jamaica 93 99 Jordan 92_{97} Jordan 97 03 Kazakhstan 88 93 Kazakhstan 93 01 Kenua 92_97 Kyrgyz Rep. 93 98 Kyrgyz Rep. 98 03 Lao 92_97 Lao 97_02 Latvia 88_93 Latvia 98 03 Lesotho 87 93Lithuania 93_98 Lithuania 98 03 Macedonia 98 03 Madagascar 80 93 Madagascar 93_99 Malawi 98_04 Malaysia 84 89 Malaysia 89_95 Mali 89_94 Mali 94_01 Mauritania 87 03 Mauritania 93_{00} Mexico 84 89

Mexico 89 95 Mexico 95 00 Moldova 92_97 Moldova 97_02 Mongolia 95 02 Morocco 85_91 Morocco 91 99 Nepal 85 96 Nepal 96 04 Nicaragua 93_98 Nigeria 86 93 Nigeria 93 03 Pakistan 87 93 Pakistan 93_99 Panama 89_95 Panama 95 00 Paraguay 90 95 Paraguay 95 02 Peru 86_94 Peru 94 00 Philippines 85 91 Philippines 91_97 Poland 89_96 Poland 96 02 Romania 89_94 Romania 94_00 Russia 94_00 Rwanda 85 00 Senegal 91 01 Slovak Rep. 88 96

Slovenia 87 93 Slovenia 93_98 South Africa 93_00 Sri Lanka 85_90 Sri Lanka 90 96 Sri Lanka 96_02 Tanzania 91 01 Thailand 81 88 Thailand 88 96 Thailand 96_{02} Tunisia 85 90 Tunisia 90 95 Tunisia 95 00 Turkey 87_94 Turkey 94_{00} Turkmenistan 93 98 Uganda 89 96 Uganda 96 02 Ukraine 88 95 Ukraine 95 02 Uruguay 81 89 Uruguay 89_96 Uzbekistan 93_98 Venezuela 81 87 Venezuela 87_93 Venezuela 93_98 Vietnam 93_98 Yemen 92 98 Zambia 91 96 Zambia 96 03 Zimbabwe 90 95

List of references

Adams, R.H., 2004. Economic growth, inequality and poverty: Estimating the growth elasticity of poverty. World Development, 32, 1989-2014.

Barro, R.J., Lee, J.W., 2000. International Data on Educational Attainment: Updates and Implications. CID Working Paper No. 42.

Baum, C., Schaffer, M., Stillman S., 2003. Instrumental variables and GMM: Estimation and testing. Stata Journal, StataCorp LP, 3, 1-31

Baum, C., Schaffer, M., Stillman S., 2007. Enhanced routines for instrumental variables/generalized method of moments estimation and testing. Stata journal, StataCorp LP, 7, 465-506.Bourguignon, F., 2003. The growth elasticity of poverty reduction: explaining heterogeneity across countries and time period. In . Eicher, T. S., Turnosky, S.J. (eds) *Inequality and Growth: theory and policy implications.* MIT Press, Cambridge, MA, London, 3-26

Bruno, M., Ravaillon, M., Squire, L., 1998. Equity and growth in developing countries: old and new perspectives on the policy issues. In: Tanzi, V., Chu, K. (eds) *Income Distribution and HIgh Quality Growth*, MIT Press, Cambridge, MA.

Carmignani, F., 2009. The distributive effects of institutional quality when government stability is endogenous. European Journal of Political Economy, forthcoming (available on line, doi:10.1016/j.ejpoleco.2009.03.003).

Chhibber, A., Nayyar, G., 2008. Pro-poor growth: explaining the crosscountry variation in the growth elasticity of poverty. International Journal of Development, 7, 160-176.

Clague, C., Keefer, P., Knack, S., Olson, M., 1999. Contract-Intensive Money: Contract Enforcement, Property Rights, and Economic Performance. Journal of Economic Growth, 4, 185-211.

Datt, G., Ravaillon, M., 1992. Growth and redistribution components of changes in poverty measures. Journal of Development Economics, 38, 75-295.Dollar, D., Kraay, A., 2002. Growth is good for the poor. Journal of Economic Growth, 7, 195-225.

Epualard, A., 2003. Macroeconomic performance and poverty reduction. IMF Working Paper N. 03/72.

Fields, G., 1989. Changes in poverty and inequality in developing countries. World Bank Research Observer, 4, 167-186.

Huang, H., Lin, Y., Yeh, C., 2009. Joint determinants of of inequality and growth. Economics Letters, 103, 163-166.

Kalwij, A., Verschoor, A., 2007. Not by growth alone: The role the distribution of income in regional diversity in poverty reduction. European Economic Review, 51, 805-829

Kakwani, N.,1993. Poverty and economic growth with application to Cote d'Ivoire. Review of Income and Wealth, 39, 121-139.

Kawkani, N., Khandker, S., Son, H., 2004. Pro-poor growth: concepts and measurement with country case studies. International Poverty Centre Working Paper 1. United Nations Development Programme. Kraay, A., 2006. When is growth pro-poor? Evidence from a panel of countries. Journal of Development Economics, 80, 198-227.

La Porta R., Lopez-de-Silanes F., Shleifer A., Vishny R., 1999. The quality of government. Journal of Law, Economics, and Organizations 15, 222-279.

Loyaza, N., Raddatz, C., 2009. The composition of growth matters for poverty alleviation. Journal of Development Economics, forthcoming (available on line, doi:10.1016/j.jdeveco.2009.03.008).

Lundberg, M., Squire, L., 2003. The simultaneous evolution of growth and inequality. The Economic Jurnal, 113, 326-344.

Mosley, P., Hudson, J., Verschoor, A., 2004. Aid, poverty reduction and the "new conditionality". The Economic Journal, 114, F217-F243.

Ravaillon, M., 1997. Can high-inequality developing countries escape absolute poverty? Economic Letters, 56, 51-57.

Ravaillon, M., 2001. Growth, inequality and poverty: looking beyond averages. World Development, 29, 1803-1815.

Ravaillon, M., 2004. Pro-poor growth: a primer. World Bank Policy Research Working Paper N. 3242

Squire, L., 1993. Fighting poverty. american Economic Review, Papers and Proceedings, 83, 377-382.

Vinod, T., Dailami, M., Dhareshwar, A., Kaufman, D, Kishor, N., Lopez, R., Wang, Y., 2000. *The Quality of Growth*. Oxford University Press for The World Bank

Wooldridge, J., 2003. *Econometric Analysis of Cross-Section and Panel Data*. The MIT Press.

Tables

Table 1: Summary statistics of poverty, growth, and elasticity

	$ph_\$1$	y_pc	$\frac{ph_{\$1}}{y \ pc}$
Full sample			
Mean	-1.24	1.32	-1.96
Median	-2.26	1.49	-0.94
Maximum	59.51	8.28	56.35
Minimum	-50.11	-11.39	-66.44
Standard deviation	19.04	3.33	13.80
Sample with $gini < 0$			
Mean	-6.99	1.72	-2.08
Median	-5.63	1.52	-1.67
Maximum	51.30	7.68	43.25
Minimum	-50.11	-3.01	-28.04
Standard deviation	14.85	2.55	9.70

Notes: The sample includes 151 spells covering 83 countries (see list in the Appendix). The variables are expressed as annualized percentage change over each spell.

	Ι	II	III	IV	V	VI
	$ph_\$1$	$ph_\$1$	$\frac{ph_{\$1}}{y_{pc}}$	$\frac{ph_{\$1}}{y_{pc}}$	y_pc	gini
$y \ pc$	-1.879***	-1.415***		0.089		-0.203***
gini		2.29***	-0.142		-0.277***	

Table 2: Correlations between poverty, growth, and elasticity

Notes: The coefficients are obtained from an OLS regression of each dependent variable on the listed regressors. The constant term is always included, but not reported in the table *, **, ***denote statistical significance at the 10%, 5%, and 1% confidence level respectively.

 Table 3: System estimates

	Ι	II	III	
	y_pc	gini	$ph_\$1$	
const	13.518^{***}	20.128^{***}	-0.403	
y_pc		-0.322***	-3.307***	
gini	-0.450***		0.801^{**}	
$ph_\$1$	-0.048***	0.018^{*}		
gov	0.088^{***}	0.160^{***}		
cpi	-0.029***	-0.005		
trade	0.019	0.017		
credit	0.119^{***}	0.044^{*}		
tyr	-0.013	0.012		
i_y_pc	0.352^{***}			
i_gini	-3.720***	-5.130***		
$legor_fr$		0.823^{***}		
lat_abst		-1.184^{**}		
ethnic		-0.546		
$y_pc * gov$			0.119	
y_pc*cpi			0.057^{**}	
$y_pc*trade$			0.458^{***}	
$y_pc*credit$			0.191^{***}	
$y_pc * tyr$			-0.568^{***}	
$y_pc*gini$		••	-0.151	
Observations		970		
Observations Source tost	50	270		
Sargan test	50.49 (44 d.f.)			
p_value		0.232		

Notes: GMM system estimates of equations 123. $i_y pc$, i_gini , $legor_fr$, lat_abst , and ethnic are exogenous. The set of instruments includes the exogenous variables plus lagged and initial values of all the other endogenous regressors. *, **, ***denote statistical significance at the 10%, 5%, and 1% confidence level respectively.

 Table 4: Single equation estimates

	Ι	II	III
	y_pc	gini	$ph_\$1$
const	19.218^{***}	18.823***	0.164
y_pc		-0.187***	-3.685***
gini	-0.607***		0.808^{**}
$ph_\$1$	-0.041***	0.014	
gov	0.088^{***}	0.111^{***}	
cpi	-0.042***	-0.011	
trade	-0.023	-0.004	
credit	0.129^{***}	0.008^{***}	
tyr	0.030	0.071^{**}	
i_y_pc	0.353^{***}		
i_gini	-5.174^{***}	-4.824***	
$legor_fr$		0.826^{***}	
lat_abst		-1.604^{**}	
ethnic		-0.726	
$y_pc * gov$			0.144
y_pc*cpi			0.057^{**}
$y_pc*trade$			0.453^{***}
$y_pc*credit$			0.196^{***}
$y_pc * tyr$			-0.460***
$y_pc * gini$			0.014
_			
Observations	90	90	90
Sargan test	15.98	13.29	20.29
p_value	0.383	0.425	0.207
$\overline{R2}$	0.692	0.835	0.524

Notes: 2SLS system estimates of equations 123. i_ypc , i_gini , $legor_fr$, lat_abst , and ethnic are exogenous. The set of instruments includes the exogenous variables plus lagged and initial values of all the other endogenous regressors. *, **, ***denote statistical significance at the 10%, 5%, and 1% confidence level respectively.