Do Redistributive Policies Affect Economic Growth?
——An Empirical Study based on Canadian data

Zheng Liu
Department of Economics, Concordia University, Montreal, Canada
Email: jacquesliu@gmail.com
Received 2012

ABSTRACT
To reduce income inequality, redistributive policies are widely adopted by both federal and provincial governments in Canada. Quebec and Canada have a fairer society in OECD countries. However, their economic growth is slower than many other countries. This paper studies how these redistributive policies affect economic growth based on Canadian data for the first time. The growth model is based on standard augmented Solow model and includes several different self-defined policy indexes. Using high quality panel data spanning the period 1982 to 2009 calculated from statistic Canada’s website and Arellano-Bond panel technique, empirical analyses show that redistributive policy is negatively and significantly associated with economic growth. These findings are in accordance with many former literatures and may have important policy significance.

Keywords: Income Inequality and Economic Growth

1. Introduction
Canada, especially Quebec, is one of the best areas around the world on reducing income inequality. However, there is a debate whether Canadian benefit from a more equal society.

The key of the debate is whether income inequality will affect economic growth, which is a classical economic question. Since Nobel laureate Simon Kuznets introduced his Kuznet’s curve which shows inequality first rises and later falls with economic growth, many economists have made their contribution on this topic theoretically or empirically. These studies either find a negative or a positive coefficient on inequality. Alesina and Rodrik (1994), Benabou (1996), and William Easterly 2006, etc conclude inequality does cause underdevelopment. However, Li and Zou (1998), Kristin Forbers (2000), Mark Patridge (2004), Mark Frank (2009) and many authors argue that inequality is not harmful for growth. Robert Barro (1999) follows Kuznet’s tradition and concludes that inequality retards growth in poor countries but encourages growth in richer places. Qual (2001) also finds little relationship between inequality and growth.

Why income inequality has negative or positive influence on economic growth? Many theories have been proposed. On negative side, first, credit-market imperfection models suggest a transfer payment from rich to poor raises the average productivity of investment. In the presence of asymmetric information and limited access to credit, investments favor people with more assets. Secondly, high income inequality in a society will often bring social unrest even revolution. Poor people are easy to engage in crime, riots and disruptive activities which are a waste of social resources. Therefore, higher inequality means more social resource waste and lower economic growth. Thirdly, under Alesina and Rodrik’s political economy model, a system of majority voting tends to redistributive policies from rich to poor and many public expenditure programs. Poor people will lose interest to work again and rich people will have to prevent this redistributive policy through bribery. Often there is no enough investment in the economy and economic growth slows down.

There are also many theories on positive relationship. First, inequality brings higher level work effort and entrepreneurial energy and therefore promotes economic growth. Secondly, under political economy model, if public expenditure is mainly about public education which will increase aggregate human capital, this will increase economic growth. A third model argues that inequality enhances mobility and the concentration of high-ability workers in technologically advanced sectors; therefore promote technological progress and economic growth. Another model is about saving rates. Some economists believe that individual saving rates rise with the level of income. A transfer payment from rich to poor means lower aggregate rate of saving in an economy and
reduce economic growth.

Clearly, no either of above theories can fully explain the effect of income inequality on economic growth. This point accords with the ambiguous empirical findings in many literatures. Level of development, political institutions, and many other factors will decide whether it is positive or negative relation between income inequality and economic growth.

This paper will take Canada as an example. Veall (2012) in his most recent paper concludes that the surge in top share incomes in Canada over the last 30 years is clear. At the same time, redistributive policies – namely, mainly taxes and transfers – are widely used by both federal and provincial governments. The effect of redistributive policies of Canadian governments on economic growth is studied for the first time (according to our survey) in this paper. Section 2 explains our growth model and dataset. Section 3 estimates this model based on the panel technique developed by Allreno and Bond. Section 4 is some discussion followed by a simple conclusion.

2. Growth Model and Dataset

This paper estimates growth as a function of redistributive policy index, lagged per capita income, investment share, education, and working population growth rate using a model similar to the augmented Solow model in N.Gregory Mankiw, David Romer and David N. Weil’s (MRW) 1992 paper. Besides using per capita GDP growth rate instead of per capita income as the dependant variable, the only other change from MRW’s original model is the addition of policy index.

The growth model central to this paper is

\[
\text{Growth}_{i,t} = \beta \text{Income}_{i,t-1} + \beta \text{Education}_{i,t} + \beta \text{Investment}_{i,t} + \beta \text{Population}_{i,t} + \beta \text{Redistributive} \_ \text{Policy}_{i,t} + \mu + \delta g + u_{i,t} \tag{1}
\]

where represents each province in Canada and \( t \) represents each time period; \( \text{Growth}_{i,t} \) is average annual growth rate of per capita GDP for province \( i \) during period \( t \); \( \text{Income}_{i,t} \) is per capita GDP for province \( i \) in time period \( t-1 \); Unlike MRW’s paper using the percentage of the working-age population that is in secondary school as education index, this paper will use a new education index. In developed country such as Canada, there is very little difference on secondary school enrollment rate for each province. Therefore, it is not a good education index again. We decide to use school board expenditure as the new education index in this paper because it is easy to acquire long span data.

As in MRW’s paper, this paper uses average share of real investment (including government investment) in real GDP as investment index and average rate of growth of the working-age population (15-64) as population index. The same as in MRW’s paper, we use \( \text{log} + (\text{Assuming } g\_0.05) \) to be the population item.

Several redistributive policy indexes are used in this paper. The first one is change of Gini coefficient before and after execution of redistributive policies. Gini coefficient is widely used in many inequality-growth literatures. Traditionally, redistribution could be achieved by many methods like progressive income tax, capital tax, minimum wage laws, public expenditure of government and others. Gini coefficient, as a good indicator of income distribution, reflects result of redistributive policies directly, especially the change of Gini coefficient before and after execution of redistributive policies.

When there is no accurate Gini coefficient data available, which is possible in many developing countries, two other policy indexes are defined based on income distribution change. Figure 1 gives an illustration of income distribution change using data from Alberta, Ontario and Quebec in three different periods. Y-axis shows the ratio of change and X-axis shows quintile. The figure clearly shows Quebec has much stronger redistributive policy than Alberta. Based on figure 1, the second and the third policy index are the average slope and the regression slope (from polyfit operation) of the income distribution change line respectively. All three policy indexes are set to be positive.

This paper will use all these different policy indexes and give a comparison. At last, \( u_{i,t} \) is the error term of the model.

![Figure 1: Illustration of income distribution change](image)

Empirical studies of income inequality/economic growth problem are often limited by the available data. Most of the inequality/economic growth literatures use nations’ data to do regression. Patridge (1997) and Frank (2009) analyze data from states in the U.S.A. Several advantages to using states to examine inequality/growth issues are believed by the authors. First, there appears to be suffi-
cient variation in income distribution to produce differential economic outcomes. Secondly, large factor flows between states should magnify how small disparities in initial conditions affect economic growth. Thirdly, the great homogeneity of state-level data helps mitigate the possible omitted variable bias. Similarly, this paper uses data from 10 provinces from 1982 to 2009 in Canada. All data are from website of Statistics Canada (CANSIM). Clearly, high quality is another important advantage to use data from CANSIM.

3. Empirical Results

Empirical study of effect of inequality on economic growth often bases on different regression methods. There are either OLS regressions over a cross-section of nations or panel techniques over both periods and geographic regions. Forbes(2000) once pointed out to evaluate which technique is optimal, there are three factors: (1) the relationship between the region-specific effect and the regressors; (2) the presence of a lagged endogenous variable; (3) and the potential endogeneity of the other regressors. The same as Forbes, this paper will use Arellano-Bond method to study the income inequality/economic growth effect. Standard panel techniques like fixed effects and random effects have an important problem that there is a lagged endogenous income variable in the growth model. Manuel Arellano and Stephen R.Bond (1991) developed a special panel technique which corrects not only for the bias introduced by the lagged endogenous variable, but also permits a certain degree of endogeneity in the other regressors. This method can eliminate the region-specific effect and then uses all possible lagged values of each of the variables as instruments through first-difference. Table 1 lists the estimation result.

As Table 1 shows, the regression coefficients of all three policy indexes are negative and significant. In all three cases, the higher policy-index means more transfer payment for the poor and higher tax for the rich. This result is consistent with the simple theory model in Li and Zou (1998)’s paper. When public consumption enters utility function, income inequality has a positive relationship with economic growth. In Canada, especially in Quebec, government spends a lot on public services like education, public health, pension plan etc. It is easy for residents to lose incentive to work harder and it is hard for government to find enough capital to invest for economic growth.

Comparing three policy indexes from column 2 to column 4, no much difference is found although PC1 has a better Sargan test result. This shows Gini change is a better policy index compared to others in our regression. This result is also consistent with former panel data literatures where income inequality has a positive relationship with economic growth. As a comparison, column 5 gives the positive regression result on Gini coefficient in our model.

For the other explanatory variables, all the regression coefficients are significant. The coefficients of lagged per capita GDP and population growth rate are negative, and the coefficients of education and investment are positive. These results are consistent with results of augmented Solow model and other researchers’ findings.

In Allreno-Bond estimation, instruments setup is very important in order to get the ideal result. To improve efficiency of instruments, first is to choose endogenous and exogenous variables. In this paper we choose lagged per capita GDP and investment share to be endogenous and policy index as exogenous variable. Secondly, “system GMM” is introduced to further improvement instruments. The “system GMM” estimator uses the levels equation to obtain a system of two equations: one differentiated and one in levels. By adding the second equation additional instruments can be obtained. Third, there is a rule of thumb in Allreno-Bond method which is to keep the number of instruments less than or equal to the number of groups. In this paper, because we have only 10 provinces (groups), we must be very careful on selection of the instruments. Here “collapse” inside gmm() option is used to reduce the number of instruments.

Sensitivity analysis is usually an important step to examine the robustness of baseline regression results. To do sensitivity analysis, more variables are required to be added into the growth model. In this paper, only two more sensitive variables, international trade share and R&D expenditure share over GDP share, are considered. In step 1, both two sensitive variables are included. In step 2 and 3, only one sensitive variable is added each time.

Table 2 shows results of the simple sensitivity analysis. When sensitive variables are added, first, better Sargan test value shows that instruments setup improves. Secondly, more variables including sensitive variables become insignificant. Thirdly, the regression coefficients of policy index keeps negative and significant. This finding seems to confirm that the negative relationship between income inequality and economic growth is robust.

4. Discussion

It is important to note that there are still several questions for further discussion. These will be explained in this section.

(1) Measurement-error and omitted variable bias

Forbes (2000) pointed out that both measurement error and omitted variable bias will bring direction bias on the estimated policy index. Measurement error will also re-
Table 1: Baseline Regression Results

Number of obs = 270  Number of groups = 10  Number of instruments = 10

<table>
<thead>
<tr>
<th>Growth Rate</th>
<th>PC1</th>
<th>PC2</th>
<th>PC3</th>
<th>Gini</th>
</tr>
</thead>
<tbody>
<tr>
<td>GDP(-1)</td>
<td>-2383283(0.000)</td>
<td>-2064298(0.000)</td>
<td>-2037915(0.000)</td>
<td>-3968854(0.000)</td>
</tr>
<tr>
<td>In(School)</td>
<td>.0269738(0.006)</td>
<td>.0226871(0.017)</td>
<td>.0219274(0.023)</td>
<td>.0376422(0.000)</td>
</tr>
<tr>
<td>In(I/GDP)</td>
<td>.1230517(0.000)</td>
<td>.0968017(0.003)</td>
<td>.0945082(0.003)</td>
<td>.2088573(0.000)</td>
</tr>
<tr>
<td>In(n+g+δ)</td>
<td>-2021602(0.009)</td>
<td>-1073335(0.055)</td>
<td>-1108057(0.050)</td>
<td>-1986347(0.004)</td>
</tr>
<tr>
<td>In(PC)</td>
<td>-1314319(0.001)</td>
<td>-0548954(0.000)</td>
<td>-0572575(0.000)</td>
<td>1.398316(0.001)</td>
</tr>
<tr>
<td>_cons</td>
<td>1.530564(0.000)</td>
<td>1.677499(0.000)</td>
<td>1.734619(0.000)</td>
<td>2.97497(0.000)</td>
</tr>
<tr>
<td>Sargan test</td>
<td>0.207</td>
<td>0.116</td>
<td>0.107</td>
<td>0.948</td>
</tr>
</tbody>
</table>

Note: P-values are In parentheses. g+δ is assumed to be 0.05.

Sargan test is used to test if the null hypothesis of “the instruments as a group are exogenous” is true. Therefore, the higher the value the better.

Table 2: Sensitive Analysis

Number of obs = 270  Number of groups = 10  Number of instruments = 10

<table>
<thead>
<tr>
<th>Growth Rate</th>
<th>step 1</th>
<th>step 2</th>
<th>step 3</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>GDP(-1)</td>
<td>-.4765018(0.043)</td>
<td>-.5322517(0.012)</td>
<td>-.2352878(0.000)</td>
<td></td>
</tr>
<tr>
<td>In(School)</td>
<td>.1402053(0.827)</td>
<td>.0576037(0.022)</td>
<td>.1543243(0.137)</td>
<td></td>
</tr>
<tr>
<td>In(I/GDP)</td>
<td>-.0393416(0.842)</td>
<td>-.1062555(0.520)</td>
<td>.1600455(0.003)</td>
<td></td>
</tr>
<tr>
<td>In(n+g+δ)</td>
<td>-.5579005(0.077)</td>
<td>-.2326544(0.030)</td>
<td>-.3910543(0.034)</td>
<td></td>
</tr>
<tr>
<td>In(PC)</td>
<td>-.483134(0.044)</td>
<td>-.4332232(0.046)</td>
<td>-.2830407(0.035)</td>
<td></td>
</tr>
<tr>
<td>In(trade)</td>
<td>.4726888(0.292)</td>
<td>.5782205(0.150)</td>
<td>.124573(0.216)</td>
<td></td>
</tr>
<tr>
<td>In(R&amp;D)</td>
<td>.0562684(0.454)</td>
<td>.3099894(0.012)</td>
<td>.312937(0.843)</td>
<td></td>
</tr>
<tr>
<td>_cons</td>
<td>1.536825(0.532)</td>
<td>3.099894(0.012)</td>
<td>.312937(0.843)</td>
<td></td>
</tr>
<tr>
<td>Sargan Test</td>
<td>0.837</td>
<td>0.79</td>
<td>0.619</td>
<td></td>
</tr>
</tbody>
</table>

Note: P-values are in parentheses. g+δ is assumed to be 0.05.

duce the significance of results. This paper addresses these issues by selection of high quality data and panel technique.

All data used in our regression are calculated from basic data of Statistics Canada’s database. The homogeneity of these province-level data helps mitigate the omitted variable problem because provinces have relatively similar growth mechanisms and institutions. Also these data can be more accurate than counterparts in other cross-nation dataset which includes many data from developing countries.

Panel technique is also used to reduce omitted-variable bias. A key advantage of our Arellano-Bond fixed-effects model is to control for any time-invariant omitted variables. Especially, as Forbes (2000) pointed out that panel techniques can specifically estimate how redistributive policies predict economic growth rate because they use within-province time series variation.

(2) Small N Large T panel problem
The Arellano-Bond estimator was designed for small-T large-N panels. In large-T panels, the correlation of the lagged dependent variable with the error term will be insignificant. (Roodman, 2006) However, the panel in this paper is a typical small-N large T panel. This is the reason that a “collapse” option has to be used in the Al-lreno-Bond command. Otherwise, there will be too many instruments in the regression which bring insignificant result.

As Forbes (2000) and Patridge (2004) pointed out, one reason for the conflicting inequality-growth results in the literature is the different time periods in the study. Short run (5-10 years) methods like fixed-effect panel estimation and long run (25-30 years) methods like OLS can have different response. This paper uses a short run technique to study inequality/growth relationship in a long period. Further reassessment is needed to study if there are any negative effects using this method. However, because similar study is limited by only 10 provinces in Canada, other small-N large-T panel technique like SURE is worth being considered.
(3) Looking for negative factor
Equality has long been considered very important to the pursuit of long-term prosperity in aggregate term for society as a whole by World Bank. Easterly (2006) concludes that structural inequality caused by conquest, colonization, slavery and land distribution is unambiguously bad for economic growth. Only inequality made by market forces may have positive relationship with economic growth. Although we can say it is mainly market inequality in a society like Canada, it is interesting to find the negative factor of inequality on economic growth.

One channel that suggests income equality can promote economic growth is human capital accumulation through better education opportunities, more creativity. However, inequality also encourages individuals to invest more on education for a better future. In our growth model, it is not easy to find a good proxy for human capital. This paper uses school board expenditure to represent education. However, some better proxies like share of university degree holders over population do not support long span period. How to find a proxy to describe a worker’s education, creativity, incentive to work and others related to human capital therefore is an important task for further study of this inequality/growth relationship, especially for long run study.

5. Conclusion
Richard Wilkinson and Kate Pickett in their famous 2009 book “The Spirit Level: Why More Equal Societies Almost Always Do Better” concluded that countries that are most equal do best. This paper is motivated by the desire to provide an answer to the question if a more equal society does good to economic growth. In real life, on the contrary, a fairer society like Quebec has lower economic growth rate than societies with higher Gini coefficient like Alberta. What is the reason behind this? Many former literatures focus on cross-countries dataset or U.S state level data. However, these data may either have measurement error problem or are not suitable to Canadian experience. This paper, for the first time as we believe, constructs a Canadian income inequality/economic growth dataset and draws a negative relationship between redistributive policy and economic growth. This result is consistent to other literatures and the simple theory model introduced by Li and Zou (1998).

The Allreno-Bond panel technique used in this paper can help to mitigate omitted variable bias and is useful for policy analysis in nature. However, the small-N large-T panel in our regression will bring some potential problems to us. Further reassessment is needed.

Income inequality and economic growth generally have an ambiguous relationship. What we find in this paper depends on our estimation technique, our dataset. An important direction for further study is to find a good proxy for human capital accumulation. This will be the key to answer if a fairer society can promote economic growth in long run.

REFERENCES