

What determines financial development?

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Abstract

This paper studies the fundamental determinants of cross-country differences in financial development. Two prominent tools for addressing model uncertainty, Bayesian Model Averaging and General-to-specific approaches, are jointly applied to investigate the financial development effects of a wide range of variables taken from various sources. The analysis suggests that the level of financial development in a country is determined by its institutional quality, macroeconomic policies, and geographic characteristics, as well as the level of income and cultural characteristics.

Keywords: Financial development; Principal component analysis; Model uncertainty; Bayesian Model Averaging; General-to-specific approach

JEL Classification: O16; E44

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1 Introduction

This paper attempts to examine systematically the factors that might account for cross-country differences in financial development. It employs two modern quantitative methods, Bayesian Model Averaging (BMA) and General-to-specific (Gets) approaches, to gauge the robustness of a selection of determinants of financial development. Special emphasis has been placed on the contributions that institutions, policy and geography may have in developing financial markets that provide funds for firms to undertake investment projects.

Firstly, take a look at some simple contrasts in the financial development experience. The United Kingdom and France have similar levels of GDP per capita, democratic institutions, and geographic characteristics in terms of latitude, access to the sea and distance from large markets. Nevertheless, they follow different legal traditions, reflected in different legal practices towards the protection of private property rights. In the 1990s the ratio of private credit to GDP in the UK (112%) was much higher than the same ratio in France (89%). Stock market capitalisation to GDP ratio in the UK was more than three times higher than that in France. How much of the difference in financial depth between the UK and France is due to the difference in their legal traditions and practices?

The financial development experience in Latin American countries provides an enlightening example of the possible role of macroeconomic policies in financial development given the similarities of geographic conditions, institutional development and cultural characteristics. After implementing market-oriented policies in the 1970s and establishing prudential regulations in the 1980s, Chile achieved a remarkable growth in financial intermediary development and stock market capitalisation, and has been regarded as the financial leader in Latin America since the mid-1980s. In the 1990s both the ratio of liquid liabilities to GDP and the ratio of private credit to GDP in Chile were fifty percentage points higher than those of Brazil, the second best country in the region. Stock market capitalisation as a fraction of GDP in Chile in the 1990s was 78%, at least three times larger than that in any other Latin American country. How much of the success of Chilean financial

development is due to better macroeconomic policies?

In the 1990s the ratio of credit issued to the private sector to GDP in Canada was 94%, more than four times higher than that in Mexico of 23%. Stock market capitalisation as a fraction of GDP in Canada in 1990s was 65%, more than two times higher than in Mexico (31%). Canada and Mexico share a number of similarities in terms of geographic endowments and institutional development. More specifically, both of them have access to the sea, have a long border with the biggest developed country, have a large land area and a democratic political system, etc. However, among others Canada and Mexico apparently differ in income level and latitude, which is associated with tropical cash crops in Mexico and grain in Canada. How much of the difference in financial depth between Canada and Mexico is due to the difference in income level and how much is due to their geographic endowment, and its effects on institutions?

Exploring what determines financial development has become an increasingly significant research topic in recent years. Examples are La Porta et al. (1997, 1998), Beck et al. (2003), Rajan and Zingales (2003) and Stulz and Williamson (2003) to mention a few. La Porta et al. (1997, 1998) have made a significant contribution to this topic concerning the legal determinants of financial development. By applying the settler mortality hypothesis of Acemoglu et al. (2001) to financial development, Beck et al. (2003) address how institutions matter for financial development. Rajan and Zingales (2003) interest groups theory argues that politics matter for financial development. Stulz and Williamson (2003) illustrate that culture matters although it may be tempered by openness. As to the role of policy, among others Huang and Temple (2005) study the importance of trade openness, whilst Chinn and Ito (2005) focus on the effect of financial openness.

Besides this, there is a large body of research aiming to identify the determinants of financial development, ranging from some emphasizing macroeconomic factors such as inflation, the income level (in terms of GDP per capita) and the saving rate, to others stressing institutional factors and geographic factors. Since the relevant economic theories provide limited guidance on the specification of a cross-country regression for financial development, it is not clear which of these factors, acting relatively independently, plays the

primary role in determining financial development when they are all taken into consideration. Formally speaking, there is a model uncertainty problem concerning which variables should be included in the model to capture the underlying data generating process.

When facing a situation where a vast literature suggests a variety of economic policy, political and institutional factors as determinants of long-run average growth rates, Levine and Renelt (1992) raised a concern over the robustness of existing conclusions in cross-section growth regressions. They found that only a few variables can be regarded as robust determinants of growth and almost all results are “fragile”. They suggested applying a version of “extreme bounds analysis” to the problem of model uncertainty. Motivated by this influential work, Sala-i-Martin (1997a, b), Fernandez et al. (2001), and Sala-i-Martin et al. (2004) are significant works among others that have investigated the contributions of various factors to cross-country growth. These works have emphasized the BMA methods as a potential technique for addressing model uncertainty.

Empirical research on the determinants of financial development encounters a similar model uncertainty problem to that on economic growth. This paper is the first attempt to study extensively the structural determinants of financial development using a large array of variables, by jointly applying BMA and the so-called LSE general-to-specific modelling approach (Gets), which is another modern method aiming to recover the true data-generating process. The Gets method has been recently developed and advocated by David Hendry and other practitioners (Hoover and Perez, 1999; Krolzig and Hendry, 2001 and Hendry and Krolzig, 2005 for example). To date, BMA and Gets have become more and more popular for the purpose of model selection, although the theory of econometric model selection is still underdeveloped.

Not only will this paper look at each individual factor, but it also pays special attention to the roles of institutions, policy and geography in the process of financial development¹. There has been substantial research on the role of institutions, policies and geography in the process of economic

¹The 39 potential determinants considered for this analysis are grouped under four headings: institutions, policy, geography and others. See Section 3.3 for details.

development in which much research regards institutions as the fundamental factor in long-run growth (Acemoglu et al. 2001, Dollar and Kraay 2003, Easterly and Levine 2003 and Rodrik et al. 2004). In particular, research by Easterly and Levine (2003) and Rodrik et al. (2004) highlights the dominant role of institutions over those of geography and policy. They argue that geography and policy affect economic development through institutions by influencing their quality, and the direct effect of geography and policy on development becomes weaker (or doesn't exist) once institutions are controlled for. Is this the case for financial development?

In three aspects, this paper exhibits distinct innovations and strengths. Firstly, it considers a wider assortment of economic, political, and geographic variables than any previous study. The second aspect is its joint application of the BMA and Gets procedures which combines the strengths of each method. By jointly applying two modern methods using a wide range of variables, more reliable conclusions can be expected. Thirdly, since as pointed out by Levine (2005) there is no uniformly accepted proxy for financial development currently available, this paper constructs a composite index of financial development using principal component analysis, which enables us to look at different dimensions of financial development including overall financial development, financial intermediary development, stock market development, financial efficiency development, and financial size development (usually called "financial depth").

The analyses based on the BMA and Gets procedures lead to the following findings: Institutions, macroeconomic policies and geography, when taken as groups, together with cultural characteristics and the income level of a country are significantly associated with the level of financial development. Of 39 variables taken individually, legal origins, government quality index, a trade policy index, land area and initial GDP and initial population are found to be important determinants of financial development.

The following section provides a brief review on the origins of financial development. Section 3 includes a description of the data. Section 4 discusses the empirical strategy and is followed by the empirical results of both BMA and Gets in section 5. Section 6 summarizes the conclusions.

2 Potential determinants of financial development

Research on the role of financial development in growth can be traced back at least to Schumpeter (1912) who points out the role of a country's banking system for economic development. The inherent functions of financial systems, including mobilizing savings to their highest valued use, acquiring information, evaluating and monitoring investment projects, and enabling individuals to diversify away idiosyncratic risk, have been widely believed to encourage productive investment and therefore total factor productivity². Given the broad consensus on the substantial role of financial development in economic growth, it is of great practical importance to understand the origins of financial development. Recent years have witnessed burgeoning research in this context. This section briefly outlines the main possible determinants of financial development, including institutional factors, macroeconomic factors, geographic factors and others, that have been studied in the literature.

2.1 Institutions

Research on the role of institutions in financial development has been considerable, especially research on the effects of the legal and regulatory environment on the functioning of financial markets. A legal and regulatory system involving protection of property rights, contract enforcement and good accounting practices has been identified as essential for financial development. Most prominently, La Porta et al. (1997, 1998) have argued that the origins of the legal code substantially influence the treatment of creditors and shareholders, and the efficiency of contract enforcement. They document that countries with a legal code like Common Law tend to protect private property owners, while countries with a legal code like French Civil Law tend to care more about the rights of state and less about the rights of the masses. Countries with French Civil Law are said to have comparatively inefficient contract enforcement and higher corruption, and less well-developed financial systems, while countries with British legal origin achieve higher levels of financial development. Among others, Mayer and Sussman (2001)

²See Levine (1997, 2005) for a review.

emphasize that regulations concerning information disclosure, accounting standards, permissible practice of banks and deposit insurance do appear to have material effects on financial development.

Beck et al. (2003)'s application of the settler mortality hypothesis of Acemoglu et al. (2001) to financial development is another significant work in this context. They argue that the extractive colonizers in an inhospitable environment aimed to establish institutions that privileged small elite groups rather than private investors, while the settler colonizers in more favorable environments were more likely to create institutions that supported private property rights and balanced the power of the state, therefore favouring financial development. Both the legal origin theory of La Porta et al. (1997, 1998) and Beck et al. (2003)'s application are related to colonisation, but the former is more concerned with how colonisation determines the national approaches to property rights and financial development, whereas the latter is more about the channel via which colonisation influences financial development.

The “new political economy” approach recently developed regards “regulation and its enforcement as a result of the balance of power between social and economic constituencies” (Pagano and Volpin, 2001). It centres on self-interested policy makers who can intervene in financial markets on either overall regulation or individual cases for purposes such as career concerns and group interests. Rajan and Zingales (2003) emphasize the role of interest groups, the incumbent industrial firms and the domestic financial sector, in the process of financial development. They argue that, in the absence of openness, incumbents have strong incentives to block the development of a more transparent and competitive financial sector which undermines the incumbents' vested interests and relationships. When both trade openness and financial openness are encouraged, the incumbents have incentives to support financial development from which more funds can be sought to meet foreign competition and new rents can be generated to partially compensate their loss of incumbency.

Inspired by the above literature, Huang (2005a) studies whether political liberalisation intending to limit the influence of the elite group (or interests group) over policy making, widen suffrage in the political system

and respect basic political rights and civil liberties is important for financial development. This work reveals that political liberalisation is typically followed by a higher level of financial development at least in the short-run. However, Huang (2005b) suggests that, once democracy has been established and enhanced, the extent of democracy may exert negative effects on the likelihood of reform aimed at financial development.

Generally speaking, institutions might have a profound impact on the supply side of financial development. The level of institutional development in a country to some extent determines the sophistication of the financial systems.

2.2 Policy

The policy view highlights the importance of some macroeconomic policies, openness of goods markets and financial liberalisation in promoting financial development. The significant effect of policy on financial development could be working through either the demand side or supply side of financial development.

Some major national macroeconomic policies such as maintaining lower inflation and higher investment have been documented to be conducive to financial development. Huybens and Smith (1999) theoretically and Boyd et al. (2001) empirically investigate the effects of inflation on financial development and conclude that economies with higher inflation rates are likely to have smaller, less active, and less efficient banks and equity markets. Huang (2005c) empirically investigates the existence and direction of causality between private investment and financial development on a panel dataset of 43 developing countries over the period 1970-1998. He shows positive causal effects going in both directions.

Some recent work has supported the view that policies which encourage openness to external trade tend to boost financial development (Do and Levchenko, 2004; Huang and Temple, 2005). Huang and Temple (2005) utilise the cross-country and time-series variation in openness and financial development, finding a positive effect of increases in goods market openness on financial development.

Apart from these, research has been carried out to study the effects of financial liberalisation on financial development over the past three decades following the McKinnon-Shaw model (McKinnon, 1973; Shaw, 1973), which concludes that, while financial repression reduces the quantity and quality of aggregate investment, financial liberalisation can foster economic growth by increasing investment and its productivity. The positive link between domestic financial liberalisation and financial development is supported by evidence (World Bank, 1989) though domestic financial liberalisation is not without risks (Demirgüç-Kunt and Detragiache, 1998). Research on the positive correlation between external financial liberalisation, especially capital account openness, and financial development is discussed in the panel data studies of Bailliu (2000) and Chinn and Ito (2005) though potential destabilizing effects may also exist. Claessens et al. (1998) present evidence that opening banking markets improves the functioning of national banking systems and the quality of financial services, with positive implications for banking customers and lower profitability of domestic banks. Laeven (2000) examines whether the liberalisation of the banking sector may help reduce financial restrictions and the external cost of capital premium, stimulating investment and financial development. Bekaert et al. (2002) provide evidence that opening up the stock market to foreign investors renders stock returns more volatile and more highly correlated with the world market return.

2.3 Geography

There is less work directly addressing the potential correlation between geography and financial development in comparison to that for policy and institutions. However, much research attention has been paid to the importance of geography for economic development, emphasizing three aspects in particular.

The first group is concerned with the correlation between latitude and economic development. Countries closer to the equator typically have a more tropical climate. On the one hand, research by Kamarck (1976), Diamond (1997), Gallup et al. (1999) and Sachs (2003a, b) suggests that tropical

location may directly lead to poor crop yields and production due to adverse ecological conditions such as fragile tropical soils, unstable water supply and prevalence of crop pests. On the other hand, tropical location can be characterised as an inhospitable disease environment, which is believed to be a primary cause for “extractive” institutions (Acemoglu et al., 2001).

A second strand of research relates to countries being landlocked, distant from large markets or having only limited access to coasts and ocean-navigable rivers (Sachs and Warner, 1995a, b, 1997; Easterly and Levine, 2003; Malik and Temple, 2005). As natural barriers to external trade and knowledge dissemination, geographic isolation and remoteness to some extent determine the scale and structure of external trade in which those countries engage. The countries’ potential to enter a large economic market and exploit economies of scale may be limited by particular geographic circumstances. The ability to develop a competitive manufacturing sector may be constrained when some intermediate products for the manufactured goods need to be imported from distant markets. The export of a limited range of primary commodities is therefore the main feature of external trade for these countries, determining their vulnerability to external shocks.

The last strand of research focuses on the link between resource endowment and economic development. Diamond (1997) suggests that countries with a richer endowment of grains species have more potential for high-yielding food crops and technological development. Isham et al. (2002) argue that a developing country’s natural resource endowment affects its economic development through an unique channel, namely natural resource endowment is linked to different export structures, different export structures determine institutional capacities towards coping with external shocks and finally the institutional quality is reflected in the level of GDP per capita. Easterly and Levine (2003) argue that the natural endowment of tropics, germs and crops indirectly influences income through their impacts on institutions.

In general, geography mainly works through the demand side of financial development, although it may affect the supply side of financial development by influencing the quality of institutions. For instance, the production of particular agricultural products or primary goods, and exploitation of some

natural resources could affect the demand for external finance.

2.4 Other variables

Other variables considered as determinants of financial development are economic growth, income level, population level, religious, language and ethnic characteristics and so on. Greenwood and Jovanovic (1990) and Saint-Paul (1992) document that as the economy grows, the costs of financial intermediation decrease due to intensive competition, inducing a larger scale of funds available for productive investment. The importance of income levels for financial development has been addressed in Levine (1997, 2003, 2005). In considering the banking sector development in transition economies, Jaffee and Levonian (2001) demonstrate that the level of GDP per capita and the saving rate have positive effects on the banking system structure as measured by bank assets, number, branches and employees for 23 transition economies.

Stulz and Williamson (2003) stress the impact of differences in culture, proxied by differences in religion and language, on the process of financial development. They provide evidence that culture predicts cross-country variation in protection and enforcement of investor rights, especially for creditor rights. The evidence also shows that the influence of culture on creditor rights protection is mitigated by the introduction of trade openness.

3 The data

This section describes the sample of countries on which this study is undertaken, and the measures of financial development and potential determinants. Appendix Table 1 contains the description and sources of these variables and Appendix Table 2 presents summary statistics.

3.1 Samples

Firstly, transition economies and small economies with a population of less than 500,000 in 1990 are excluded from the sample. The information on the transition economies and population size are from the World Bank Global

Development Network Database (GDN) and the Penn World Table 6.1 due to Heston et al. (2002), respectively.

This study mainly investigates key determinants of five specific indices of financial development, which will be discussed in more depth below. For each financial development index, there are three samples on which the investigation is based: the whole sample, a developing country sample and a smaller sample for which the La Porta et al. (1998) data are available. The whole sample is the main focus of the analysis. The developing countries in the settler mortality dataset of Acemoglu et al. (2001) form the main part of the developing country sample here. Looking at the La Porta et al. (1998) sample makes it possible to examine whether differences in legal tradition, reflected in the protection of shareholders' rights and creditors' rights, determine cross-country differences in financial development. The countries included are listed in Appendix Table 3.

3.2 Measures of financial development

Since there is no single aggregate index for financial development in the literature, we use principal component analysis based on widely-used indicators of financial development to produce new aggregate indices.

Essentially the principal components analysis takes N specific indicators and produces new indices (the principal components) X_1, X_2, \dots, X_N that are mutually uncorrelated. Each principal component, as a linear combination of the N indicators, captures a different dimension of the data. Typically the variances of several of the principal components are low enough to be negligible, and hence the majority of the variation in the data will then be captured by a small number of indices. This paper uses the first principal component, which accounts for the greatest amount of the variation in the original set of indicators in the sense that the linear combination corresponding to the first principal component has the highest sample variance subject to the constraint that the sum-of-squares of the weights placed on each existing indicator is equal to one.

The conventional measures of financial development on which the prin-

principal component analysis is based are as follows³:

The first measure, Liquid Liabilities (*LLY*), is one of the major indicators to measure the size, relative to the economy, of financial intermediaries, including three types of financial institutions: the central bank, deposit money banks and other financial institutions. It is calculated as the liquid liabilities of banks and nonbank financial intermediaries (currency plus demand and interest-bearing liabilities) over GDP.

The second indicator, Private Credit (*PRIVO*), is defined as the credit issued to the private sector by banks and other financial intermediaries divided by GDP excluding credit issued to government, government agencies and public enterprises, as well as the credit issued by the monetary authority and development banks. It measures general financial intermediary activities provided to the private sector.

The third one, Commercial-Central Bank (*BTOT*), is the ratio of commercial bank assets to the sum of commercial bank and central bank assets. It proxies the advantage of financial intermediaries in channelling savings to investment, monitoring firms, influencing corporate governance and undertaking risk management relative to the central bank.

Next are two efficiency measures for the banking sector. Overhead Costs (*OVC*) is the ratio of overhead costs to total bank assets. The Net Interest Margin (*NIM*) equals the difference between bank interest income and interest expenses, divided by total assets. A lower value of overhead costs and net interest margin is frequently interpreted as indicating greater competition and efficiency.

The last are three indices for stock market development⁴. Stock Market Capitalization (*MCAP*), the size index, is the ratio of the value of listed domestic shares to GDP.

Total Value Traded (*TVT*), as an indicator to measure market activity,

³The description of these measures are heavily drawn from Demirgüç-Kunt and Levine (1996, 1999).

⁴Since data for the efficiency of the bond market are not available while data for the size of the bond market are mainly available for developed countries in the World Bank's Financial Structure and Economic Development Database (2005), to avoid resulting in smaller sample sizes in the principal component analysis, bond market development is not included here. A simple analysis of the determinants of bond market development (for a smaller sample) is presented in Appendix Table 8.

is the ratio of the value of domestic shares traded on domestic exchanges to GDP, and can be used to gauge market liquidity on an economy-wide basis.

Turnover Ratio (*TOR*) is the ratio of the value of domestic share transactions on domestic exchanges to total value of listed domestic shares. A high value of the turnover ratio will indicate a more liquid (and potentially more efficient) equity market.

The data are obtained from the World Bank's Financial Structure and Economic Development Database (2005) and averaged over 1990-2001. Any measure for which fewer than three years of data are available is treated as a missing value.

Appendix Table 4 presents the eigenvalues, proportion explained, and the eigenvector of each first principal component from which the new indices of financial development are defined. It reports the sample variance of each first principal component (linear combination), the proportion of the variance in the raw data the first principal component accounts for, and the coefficient (weight) of each existing standardised measure in the linear combination.

(1) The first is a measure of overall financial development, denoted by *FD*. This is based on eight components, namely Liquid Liabilities, Private Credit, Commercial-central Bank, Overhead Cost, Net Interest Margin, Stock Market Capitalisation, Value Traded and Turnover. The first principal component accounts for 49% of the variation in these seven indicators. In Appendix Table 4 the coefficients of each financial indicator for *FD* indicate the negative correlations between the Overhead Cost and Net Interest Margin and *FD*, and the positive correlations between the rest and *FD*.

(2) A second measure, *FDBANK*, captures the extent of bank-based intermediation. It uses five indicators, Liquid Liabilities, Private Credit, Commercial-central Bank, Overhead Costs and Net Interest Margin. *FDBANK* accounts for 61% of the variation in these five indicators.

(3) A third measure, *FDSTOCK* is a measure of stock market development, based on Stock Market Capitalization, Value Traded and Turnover. *FDSTOCK* accounts for 66% of the variations in these financial indices.

(4) A fourth measure, *FDEFF*, captures financial efficiency. Four indicators of financial efficiency used are Overhead Cost, Net Interest Margin, Value Traded and Turnover. *FDEFF* accounts for 54% of the total varia-

tion in these indicators. Lower values of this index indicate a higher level of financial efficiency.

(5) A fifth measure, *FDSIZE*, based solely on Liquid Liabilities and Stock Market Capitalization, captures the size of financial system (also called “financial depth”). The first principal component of these two measures accounts for 81% of the variation.

3.3 The potential determinants

Potential determinants of financial development are widely selected from various sources. To discover the structural determinants of financial development, they are either those “predetermined” like fixed factors, or those “evolving slowly over time” like some institutional factors which are averaged over 1960-89. All variables that could potentially cause serious endogeneity problems are excluded⁵. The candidate determinants are grouped into four categories as showed in Appendix Table 1. The problem of missing data has been addressed by using a set of fixed factors as independent variables to impute the missing data. The fixed factors used include some regional dummies, dummies for income levels and geographic factors for which we have a complete set of data. The imputation information is presented in Appendix Table 5.

3.3.1 Institutional variables

This analysis firstly considers legal origin dummies from the GDN due to the work by La Porta et al. (1997, 1998) on the legal determinants of financial development. The relevant variables are the common law legal origin dummy (COMLEG) for countries with British legal origin and a civil law legal origin dummy (CIVLEG) for countries with French, Germany and Scandinavian legal origins. Two variables below closely related to the financial system itself are also considered⁶. Taken from the dataset of La Porta et al. (1998),

⁵Measures of financial liberalization and financial openness are not used here due to the concern that the effects of other variables on financial development may work through them.

⁶Accounting standards data in La Porta et al. (1998) forms another interesting variable, but this variable has to be excluded due to its limited country coverage.

SRIGHT is the aggregate index for shareholders' rights ranging from 0 to 6, while CRIGHT is the aggregate index for creditors' right ranging from 0 to 4. These variables directly measure the extent to which the government protects the rights of shareholders and creditors.

In addition, this research makes use of some general institutional indicators. POLITY2 and DURABLE are taken from the PolityIV Database (Marshall et al., 2003), and averaged over 1960-89. POLITY2 is an index of democracy, seeking to reflect government type and institutional quality based on freedom of suffrage, operational constraints and balances on executives, and respect for other basic political rights and civil liberties. It is called the "combined polity score", equal to the democracy score minus the autocracy score. The democracy and autocracy scores are derived from six authority characteristics (regulation, competitiveness and openness of executive recruitment; operational independence of chief executive or executive constraints; and regulation and competition of participation). Based on these criteria, each country is assigned a democracy score and autocracy score ranging from 0 to 10. Accordingly, POLITY2 ranges from -10 to 10 with higher values representing more democratic regimes. DURABLE is an index of political stability, using the number of years since the last regime transition or independence. The next variable is FREE, the average of the indexes of civil liberties and political rights from the Freedom House Country Survey (2003) over 1972-89. Higher ratings indicate better civil liberties and political rights such as freedom to develop views, institutions and personal autonomy apart from government. I also employ KKZ and PCI. The KKZ measure from Kaufmann et al. (1999) is a widely-used indicator of the quality of government in a broader sense, derived by averaging six measures of government quality: voice and accountability, political stability and absence of violence, government effectiveness, light regulatory burden, rule of law, and freedom from graft. The variable PCI, measuring narrowly the constraints on the executive, is derived by Henisz (2000). The last institutional variable I use is EURO1900, the percent of population that was European or of European descent in 1900, taken from Acemoglu et al. (2001).

Although missing values for EURO1900, SRIGHT, CRIGHT as well as the market share of state-owned media (discussed below) are imputed, the

variable EURO1900 only appears in the developing country sample while the others only appear in the La Porta sample.

3.3.2 Policy variables

To examine whether macroeconomic policy variables explain cross-country variation in financial development, this research makes extensive use of five economic volatility indicators and three trade openness indicators. It uses output volatility and inflation volatility to capture macroeconomic mismanagement and fluctuations. The output volatility measure (SDGR) is defined as the standard deviation of the annual growth rate of real, chain-weighted GDP per capita over 1960-89 from the Penn World Table 6.1. Inflation volatility (SDPI) is defined as the standard deviation of the annual inflation rate over 1960-89 from the World Development Indicators (2004). Taken from the GDN, the volatility of the black market premium (SDBMP), volatility of the terms of trade (SDTT) and trading partners' output volatility (SDTP) are used to reflect the extent of external shocks. SDBMP is defined as the standard deviation of the annual black market premium (BMP) over 1960-89. SDTT is defined as the standard deviation of the first log-differences of a terms of trade index for goods and services. SDTP is the standard deviation of trading partners' GDP per capita growth (weighted average by trade share).

To assess the role of trade openness, this research uses dummies for fuel, non-fuel primary goods exporting countries (EXPPRIM) and manufactured goods exporting countries (EXPMANU) from the GDN. A trade openness policy index, TOPEN, available from the database of Harvard University's Center for International Development (Gallup et al., 1999), is utilised to measure the extent of openness to external trade in the presence of government intervention over 1965-90, while the trade share proposed by Frankel and Romer (1999), denoted by CTRADE, is employed to capture natural openness to external trade. CTRADE is derived by summing up all bilateral trade with all potential trading partners from a bilateral trade equation that controls for population and land area of the home country and trading partners, the distance between any two trade partners and whether or not

the home country is landlocked.

3.3.3 Geographic variables

To examine the role of geography, this study takes six regional dummies from the GDN for East Asia and Pacific countries (REGGEAP), Middle East and North African countries (REGMENA), West Europe and North American countries (REGWENA), South Asian countries (REGSA), Sub Saharan African countries (REGSSA) and Latin America and Caribbean countries (REGLAC), respectively. It also uses the following two geographic variables from the GDN. The landlock variable (LANDLOCK) is a dummy value that takes the value of 1 if the country has no coastal access to the ocean, and 0 otherwise. There are 17 countries that are landlocked in the whole sample. Absolute latitude (LATITUDE) equals the absolute distance from the equator of a country. The closer to the equator the countries are, the more tropical climate they have⁷. Latitude has an institutional interpretation since smaller absolute latitudes are associated with more unfavorable environments, which are associated with weaker institutions according to the settler mortality hypothesis of Acemoglu et al. (2001). The land area (AREA) in square kilometers for each country, taken from Hall and Jones (1999), is in logs.

This study also makes use of three additional geographic variables. One is POP100CR from the database of Harvard University's Center for International Development. It is the 1994 share of population within 100km of a coast or navigable river for a country. Another is MINDIST, based on data from Jon Haveman's International Trade website. This captures the minimum distance from three capital-goods-supplying centers in the world (USA, Japan and EU represented by Belgium). The study uses the logarithm of the minimum distance from three capital-goods-supplying centres plus one. These variables might be highly correlated with external trade and manufacturing since lack of access to coasts or ocean-navigable rivers and geographic remoteness constitute natural disadvantages to external trade. A further variable for geographic endowment is a dummy for the point source

⁷To some extent, absolute latitude serves as an alternative indicator for the zero-one tropical dummy in the GDN.

natural resource exporting countries (RESPOINT) from Isham et al. (2002), who find that, in comparison to manufacturing exporters, and exporters of “diffuse” natural resources (e.g. wheat, rice, animals) and coffee/cocoa natural resources, the exporting countries of “point source” natural resources (e.g. oil, diamonds, plantation crops) are more likely to have severe social and economic divisions, and less likely to develop socially cohesive mechanisms and effective institutional capacities to manage shocks.

3.3.4 Other variables

Other variables included in this analysis are initial income (GDP90), initial population (POP90), an ethnic fractionalisation index (ETHNIC), an ethnic polarisation index (ETHPOL), a religious fractionalisation index (RELIGION), a language fractionalisation index (LANGUAGE), a European first language index (EURFRAC), and the market share of state-owned media, either TV or newspapers (MEDSHARE).

The inclusion of the level of GDP per capita in 1990 (GDP90) is stimulated by work such as Greenwood and Smith (1997) on the feedback from growth in the economy to the development of financial markets. Population size is also closely related to indexes of financial development since small countries tend to have higher ratios of liquid liabilities and private credit, having the potential to affect the overall results substantially. GDP90 and POP90, the level of the population in 1990, are from the GDN and used in logs.

The variables ETHNIC, RELIGION and LANGUAGE, taken from Alesina et al. (2003), characterise social divisions and cultural differences, as does the variable ETHPOL, which is taken from Reynal-Querol and Montalvo (2005) to capture the extent to which a large ethnic minority faces an ethnic majority in a society. The EURFRAC measure, taken from Hall and Jones (1999), is the fraction of population speaking one of the major languages of Western Europe (English, French, German, Portuguese, or Spanish) as a mother tongue. To some extent, this variable reflects not only the culture of the country, but also the history of colonisation. It is therefore closely linked to some other variables like EURO1900, CIVLEG and COMLEG.

The market share of stated-owned media (MEDSHARE) is from Djankov et al. (2003) which shows that greater state ownership of the media is associated with less political and economic freedom, inferior governance, less developed capital markets and poor health outcomes. Djankov et al. (2003) consider two kinds of media state ownership. One is press state ownership, the market share of state-owned newspapers out of the aggregate market share of the five largest daily newspapers (by circulation), and the other is television state ownership, the market share of state-owned television stations out of the aggregate market share of the five largest television stations (by viewer). The index used here is the average of the two media state ownerships.

4 Empirical Strategy

This section discusses the empirical strategies for dealing with model uncertainty faced by research on the determinants of financial development, with the central focus placed on Bayesian Model Averaging and General-to-specific approaches.

As summarised in section 2, substantial research has been done to explore the origins of financial development, leading to a large number of candidate determinants. Essentially the associated theories, developed under specific settings, are not mutually exclusive, raising concern over the robustness of these candidate determinants in any cross-section regression used to explain financial development.

Usually, the uncertainty about the composition of a regression model is called “model uncertainty”. To handle the model uncertainty issue, a number of econometric methodologies have been proposed and widely debated. Among others, the Extreme Bounds Analysis (EBA), Bayesian Model Averaging (BMA) and General-to-specific (Gets) are the most famous methods.

The EBA proposed by Leamer (1983, 1985) regards a variable to be robust if its extreme bounds lie strictly one side or the other side of zero, where the extreme bounds for the coefficients of a particular variable are defined as “the lowest estimate of its value minus two times its standard error and the highest estimate of its value plus two times its standard error, respectively”.

The interval formed by two extreme bounds constitutes the maximum scope a variable may vary in the presence or absence of other variables, and indicates the confidence one may have in the coefficient estimates.

The EBA method has been widely applied in the context of cross-section growth regression since then. A version of EBA was used by Levine and Renelt (1992) to consider alterations of the conditioning information set in a robustness analysis of cross-section growth regressions. More specifically, they include three “free” variables⁸, one “focus” variable and at most three “doubtful” variables in all models so as to reduce the number of regressions needed to calculate the extreme bounds. By this method, they find almost all variables can be regarded as fragile determinants of growth. Sala-i-Martin (1997a, b) criticizes the standard of robustness employed by Levine and Renelt (1992) as too restrictive and suggests a different version of extreme bounds analysis by saying that a variable is robust as long as 95% or more of the distribution of estimates lays one side of zero. By this methodology, relatively more variables are found to be robustly related to growth. By applying a modified version of EBA to the Levine and Renelt (1992) dataset, Temple (2000) argues that “the most common objections to EBA are either misplaced or easily taken into account by a careful presentation of the results”. He also regards the EBA as “a useful means of providing information on the sensitivity of results to alternative modelling choices, including the extent of uncertainty in model specification, parameter estimates and standard errors”.

The methodology of Sala-i-Martin (1997 a, b) is not Bayesian although it uses weights proportional to the likelihoods of each model. Fernandez et al. (2001) re-examine the Sala-i-Martin (1997a, b) data set using a full BMA explained below and Markov Chain Monte Carlo techniques to deal with the huge range of possible models. The full BMA of Fernandez et al. (2001) requires fully specifying the prior distribution for every parameter conditional on each possible model and calculates the average of the parameter estimates across all possible models by using corresponding posterior model probabilities as weights. Their research has produced findings in support of

⁸They are real GDP per capita in 1960, the primary school enrolment rate in 1960 and the average investment share of GDP during 1960-89.

the conclusions of Sala-i-Martin (1997a, b). However, fully specifying the prior distribution for all potential parameters is very difficult and “essentially arbitrary” (Sala-i-Martin et al. 2004) when the number of possible regressors is large. Sala-i-Martin et al. (2004) propose a version of BMA called Bayesian Averaging of Classical Estimate (BACE), in which diffuse priors are assumed for the parameters and only one other prior, relating to the expected model size, is required. This approach has generated evidence in favour of Sala-i-Martin (1997a, b)’s original findings as well.

The third strand of research on model uncertainty is the general-to-specific modelling strategy (Gets) associated with the LSE methodology. It starts from the most general unrestricted model (GUM) which is assumed to characterise the essential data generating process (DGP), applies standard testing procedures to eliminate statistically insignificant variables and ends up with a “congruent” final model, which should be free of significant misspecification.

Recent developments in computing technology render these methods rather easy to adopt. A computer programme for the Bayesian approach to model uncertainty has been developed by Raftery (1995). A computer algorithm designed for implementing the general-to-specific approach, called PcGets, has been developed by Krolzig and Hendry (2001).

Undoubtedly, the econometric theory and methods of model selection are still underdeveloped. Although the BMA and Gets procedures have respective advantages in handling model uncertainty, neither of them is without limits and exempted from criticism. As argued by Granger and Hendry (2005) and echoed by Hansen (2005), none of the model selection methods currently available is immune from four possible conceptual errors of model selection methods: parametric vision, the assumption of a true data generating process, evaluation based on fit and ignoring the impact of model uncertainty on inference. This research chooses to jointly apply the BMA and Gets procedures to handle model uncertainty in this context. The combination of Gets and BMA analyses has the advantage of incorporating their merits while circumventing some of their limitations. In what follows, I set out the methods of BMA and Gets in more detail.

4.1 Bayesian Model Averaging

The sketch of the Bayesian approach to model uncertainty in this section heavily relies on Raftery (1995), Sala-i-Martin et al. (2004) and Malik and Temple (2005).

Essentially, BMA treats parameters and models as random variables and attempts to summarise the uncertainty about the model in terms of a probability distribution over the space of possible models. More specifically, it is used to average the posterior distribution for the parameters under all possible models, where the weights are the posterior model probabilities. To evaluate the posterior model probability, the BMA uses the Bayesian Information Criterion to approximate the Bayes factors that are needed to compute the posterior model probability discussed in detail below.

Typically, the number of possible models, 2^p given p candidate variables, is large. Most applications of BMA to larger datasets do not average over all possible models, but use a search algorithm to identify the subset of models with greatest relevance. The Occam’s Window and Markov Chain Monte Carlo techniques can be adopted for this purpose. The Markov Chain Monte Carlo techniques developed by Hoeting et al. (1996) have the advantage of selecting variables and identifying outliers simultaneously, but require a larger sample size relative to the regressor set, and so this method will be applied in Table 1 only. The simpler version of BMA used elsewhere in this study follows Raftery (1995) which only focuses on the subset defined by the Occam’s Window technique and treats all the worst-fitting models outside the subset as having zero posterior probability. Embodying the principle of parsimony⁹, the use of the Occam’s Window technique considerably reduces the number of possible models, and in the meantime encompasses the inherent model uncertainty present. Once the Occam’s Window technique excludes the relatively unlikely models, the posterior model probabilities for the well-fitting models are then calculated.

⁹The Occam’s Window approach can be divided into two types, corresponding to two approaches. One is the symmetric Occam’s Window in which models “much less likely than the most likely model” are excluded, the other is the strict Occam’s Window in which the models having “more likely submodels nested within them” are excluded from the subset left in the symmetric Occam’s Window (Raftery, 1995).

The posterior model probability is a critical concept for BMA. Below is a brief summary on how to derive it.

We suppose there are many models, $\{M_1, \dots, M_k\}$ for the data D . Every model is specified by a vector of d unknown parameters $\theta_i = (\theta_{i1}, \theta_{i2}, \dots, \theta_{id})$, $i = 1, 2, \dots, K$. These models may be nested or not. Bayesians treat the unknown parameters as random variables.

Let Δ denote a quantity of interest such as a parameter. The posterior distribution of Δ given data D is derived according to

$$P(\Delta|D) = \sum_{i=1}^K P(\Delta|D, M_k)P(M_k|D) \quad (1)$$

where $P(M_k|D)$ are the posterior model probabilities, and $P(\Delta|D, M_k)$ is the posterior distribution of Δ given the data D and model M_k .

The equation contains all information needed to make inference about Δ , indicating that the posterior distribution of Δ given data D is a weighted average of its posterior distributions given data D and a specific model. The weights are the posterior model probabilities, $P(M_k|D)$, which can be obtained by Bayes' theorem,

$$P(M_k|D) = \frac{P(D|M_k)P(M_k)}{\sum_{i=1}^K P(D|M_i)P(M_i)} \quad (2)$$

where $P(M_k)$ is the prior probability of model i ($i = 1, 2, \dots, K$), and $P(D|M_i)$ is the probability of the data given M_i , also called the integrated (marginal) likelihood for model M_i or marginal (predictive) probability of the data given M_i .

To represent no prior preference for any model, each model will start on an equal footing, that is $P(M_1) = P(M_2) = \dots P(M_k) = \frac{1}{K}$. Therefore the posterior model probabilities $P(M_k|D)$ can be rewritten as

$$P(M_k|D) = \frac{P(D|M_k)}{\sum_{i=1}^K P(D|M_i)} \quad (3)$$

To identify the value of $P(D|M_k)$, it is useful to compare model M_k with

a baseline model. A null model (M_0) in which no independent variables are included is usually used as a baseline model¹⁰.

Let B_{k0} be the Bayes factor for model M_k against model M_0 , that is

$$B_{k0} = \frac{P(D|M_k)}{P(D|M_0)} \quad (4)$$

then,

$$2 \log B_{k0} = 2 \log P(D|M_k) - 2 \log P(D|M_0) \quad (5)$$

Due to the introduction of the BIC (Bayesian Information Criterion) approximation to the Bayesian factor, “ $2 \log B_{k0}$ ” can be expressed as the approximation of the difference between BIC_0 and BIC_k , the values of BIC for the null model, M_0 , and model M_k , respectively,

$$2 \log B_{k0} \approx BIC_0 - BIC_k \quad (6)$$

The fact that $BIC_0 = 0$ yields the approximation for the posterior probability $P(D|M_k)$, which is

$$P(D|M_k) \propto \exp\left(-\frac{1}{2}BIC_k\right) \quad (7)$$

The posterior model probabilities $P(M_k|D)$ can be then written as

$$P(M_k|D) \approx \frac{\exp\left(-\frac{1}{2}BIC_k\right)}{\sum_{i=1}^K \exp\left(-\frac{1}{2}BIC_i\right)} \quad (8)$$

At this point, we are ready to implement a systematic form of inference for different quantities of interest. For example, when the interest is one of the regression parameters being present, positive or negative, what we need to do is to sum up the posterior model probabilities for all models in which the parameter is non-zero, positive or negative. In the section on empirical results below, the output of the BMA analysis includes the posterior inclusion probabilities for variables and a sign certainty index.

¹⁰A saturated model (M_s) in which each data point is fitted exactly can be also used as a baseline model.

The posterior inclusion probability for any particular variable is the sum of the posterior model probabilities for all of the models including that variable. The higher the posterior probability for a particular variable, the more robust that determinant for financial development appears to be. For posterior inclusion probabilities greater than 0.20, a sign certainty index rather than sign certainty probability is presented, clearly suggesting the relationship being either positive or negative.

The computational procedure for the Occam’s Window technique is implemented using the bicreg software for S-Plus or R written by Adrian Raftery and revised by Chris Volinsky. The program used to calculate the sign certainty index is developed by Malik and Temple (2005) based on the original bicreg code.

4.2 General-to-specific approach

Hoover and Perez (1999) make important advances in practical modelling, like the multiple-path approach to Gets model selection. Based on these, PcGets has been developed to embody the principles of the underlying theory of general-to-specific reductions extensively discussed in Hendry (1995).

The selection of models by PcGets roughly includes three stages¹¹:

The first stage concerns the estimation and testing of the GUM. The GUM should be formulated carefully based on previous empirical and theoretical findings, institutional knowledge and data characteristics. The specification of the GUM should be sufficiently general with a relatively orthogonal parameterization for the N candidate regressors. The next step is to conduct a mis-specification test for “congruence” of the initial GUM. The congruence of the initial GUM is maintained through the selection process to ensure a congruent final model. Once the congruence of the GUM is established, pre-search reduction tests are conducted at a loose significance level. The statistically-insignificant variables are eliminated both in blocks and individually, and the GUM reformulated as the baseline for the next stage.

¹¹The summary below is heavily drawn from Hoover and Perez (1999), Krolzig and Hendry (2001), Hendry and Krolzig (2005) and Granger and Hendry (2005).

The second stage is the search process. Many possible reduction paths are investigated to avoid path-dependent selection. The terminal model emerges from each path when all reduction diagnostic tests are valid and all remaining variables are significant. At the end of the path searches, all distinct terminal models are collected and tested against their union to find an undominated encompassing contender. If a unique model results, it is selected; otherwise, the “surviving” terminal models form a union as a new starting point for reduction. The search process continues until a unique model occurs, or the union coincides with the original GUM or a previous union. If a union made up of mutually encompassing and undominated models result, PcGets employs the Bayesian Information Criterion to select the unique final model.

The third stage is the post-search evaluation. At this stage PcGets use post-selection reliability checks to evaluate the significance of variables in the final model selected in two overlapping subsamples.

Obviously, the choice of critical values for pre-selection, selection encompassing tests and sub-sample post-selections is important for the success of the PcGets algorithm. It provides two basic strategies, the liberal and conservative strategy, for the levels of significance, degree of pretesting and so on. The liberal strategy tries to equate the probability of deleting relevant and retaining irrelevant variables, whilst the conservative strategy tries to reduce the chance of retaining irrelevant variables. The choice of different strategies hence affects the chance of either retaining irrelevant variables or dropping relevant variables. Throughout the paper, PcGets is conducted with a more liberal strategy than the default setting of the “liberal strategy” as presented in the Appendix Table 6¹², aiming to keep all promising variables in the final model. The final conclusion is then based on the intersection of the BMA and PcGets results.

¹²Since any variable removed at the presearch stage is permanently eliminated, the F presearch testing (top-down) at step 1 in the “liberal strategy” default setting has been increased from 0.75 to 1, so as not to risk omitting any potential factor which might not be significant in the GUM.

5 Empirical results (I): Overall financial development

This section begins studying the determinants of various indexes of financial development. The BMA and Gets methods are applied and compared on three different samples (the whole sample, the developing country sample and the La Porta sample) for each index. This section, the central contribution of this analysis, studies the determinants of overall financial development (FD). Section 6 is concerned with the determinants for four specific indexes of financial development, followed by a study of the determinants of bond market development.

5.1 Some stylized facts

As a starting point, it might be useful to look at some stylized facts on the links between some important institutional, policy and geographic variables and FD. These figures are based on the whole sample.

Figure 1 presents two scatter plots for the links between institutions and financial development. Better institutional quality, captured by *KKZ*, and a more democratic regime, captured by *POLITY2*, are associated with an increase in FD. The trade policy index denoted by *TOPEN* and Frankel-Romer trade share denoted by *CTRADE* are positively related to FD in Figure 2. The upper chart of Figure 3 indicates that countries closer to the main world market centers achieve a higher level of FD, while the lower chart shows that financial markets in countries further from the equator are relatively more advanced.

Figure 4 portrays the evolution of averaged liquid liability (*LLY*) over 1960-2003 by different country groups. Note from the upper-left chart that countries in all income groups experienced an increase in *LLY*, though higher-income countries remain at a higher level of financial development than lower-income countries throughout. The upper-right chart shows considerable differences in averaged *LLY* between manufactured goods exporting countries and primary goods exporters in which the latter remain at lower levels or at least partially financial repressed. The lower-left chart shows that the level of *LLY* in West Europe and North American countries

was much higher and more stable than that in other country groups. The development process of LLY in East Asia and Pacific countries was much more pronounced relative to that in any other country group. In the last chart, the development performance of LLY in common law countries was in general much more gradual with the whole process stretching over four decades compared to that in civil law countries, which experienced surges in the 1970s and late 1990s, but a decline in the late 1980s.

The figures above have shown some interesting facts on the determinants of FD. However, a clear conclusion on the robustness of any variable presented cannot be readily drawn. The task of the subsequent section is to examine these links systematically.

5.2 What are the main determinants of FD?

As mentioned in the introduction, much research regards institutions as the fundamental factor in long-run growth, and some even argue that the only effect of geography and policy on development is via institutions (Acemoglu et al., 2001; Dollar and Kraay, 2003; Easterly and Levine, 2003; and Rodrik et al., 2004). Before proceeding to study the main determinants of overall financial development (FD), this section starts by testing the hypothesis of whether any of three determinants (institutions, policy and geography), considered as a whole, dominates the other two.

Table 1 reports the BMA results for determinants of FD, which is measured over 1990-99, for 64 countries in the whole sample. All possible explanatory variables are grouped into 4 blocks in the order of “other” variables, geographic variables, policy variables and institutional variables. In addition to including the “other” variables, the models 1-3 include any two of the three blocks (geographic variables, policy variables and institutional variables) to examine the combined effects of any two types of determinants on FD¹³.

The BMA analysis yields posterior inclusion probabilities (either “PIPs” or “MC3”)¹⁴, the total posterior model probabilities for the set of mod-

¹³The effect of institutions, policy or geography on financial development is also examined in isolation. The results are not reported here, but are available from the author.

¹⁴MC3 is essentially a shorthand for the Markov Chain Monte Carlo technique, which

els which include a given variable of interest, and the sign certainty index (“Sign”) of a relationship discussed above. The PIPs are the posterior inclusion probabilities calculated by using Occam’s window to select models due to Raftery (1995). A sign certainty index is provided where the PIPs are above 0.2. The *MC3* denotes the posterior inclusion probabilities computed by using the Markov Chain Monte Carlo techniques due to Hoeting et al. (1996), which conduct variable selection and outlier identification simultaneously. Any *MC3* greater than 0.2 is in bold.

Looking at the first block of “other” variables across models, we note that initial income, GDP90, appears to be important in almost all models with a high posterior probability of inclusion, meaning that the level of GDP per capita is fundamental in explaining the cross-country variation in FD. Other variables in this block exhibit varying explanatory power for FD. Model 1 and 2 present the effect of geography on FD when policy and institutions are controlled for, respectively. The effect of geography on FD doesn’t seem to disappear when the institutional variables are present, implying that the usual claim that geography works through institutions is not necessarily true in this context. The two BMA methods show that two regional dummies (REGSSA and REGLAC) appear to be closely related to FD, meaning a number of developing countries in these regions are associated with higher levels of financial development in 1990s, conditional on other variables. The regional dummy REGEAP and land area (AREA) also appear to be important predictors of FD when institutions are controlled for. Similarly, policy has a significant effect on FD in the presence of geography and institutions (model 1 and 3). Among others, at least EXPPRIM is significant in both cases. Neither does the usual claim that policy works through institutions by affecting their quality apply to this context. Model 2 and 3 show that the role of institutions is not altered when geography and policy are controlled for. Note that most of the institutional variables appear to be significant predictors of FD, in particular, the KKZ (governance index) and PCI (political constraints index) have a posterior probability of inclusion close to 1.

is only applied in this table as a robustness test.

Overall, Table 1 has demonstrated that geography, institutions and policy as a group are all important in the process of financial development, although their effects may be picked up by varied predictors when conditioning on other factors is in place. These results clearly suggest that it would be more appropriate to include all of them in the models.

Table 2 contains a thorough study of determinants of FD by using BMA and Gets on three samples in which the above conclusion in terms of geography, institutions and policy being all important is embodied. In the whole sample, there are 64 observations together with a set of 35 potential explanatory variables in all. In the sample of developing countries there are 44 observations and a set of 35 potential explanatory variables which is different from the variable set for the whole sample by the exclusion of REGWENA and the inclusion of EURO1900. The La Porta sample has 40 observations, and it has 3 additional variables (MEDSHARE, SRIGHT and CRIGHT) on top of the variable set for the whole sample.

The BMA analysis reports posterior probabilities of inclusion (PIPs) and the sign certainty index (“Sign”) discussed above. The PcGets analysis produces the coefficients and t-values for possible determinants in the final model. The PcGets also reports the residual sum of squares (RSS); the equation standard error or residual standard deviation (sigma); the squared multiple correlation coefficient (R^2) and its values adjusted for degree of freedom (R^2_{adj}); the log-likelihood value and three information criteria (AIC, HQ and SC). The output also includes three mis-specification tests (Chow test, Normality test and Heteroscedasticity test)¹⁵. As mentioned earlier, PcGets is conducted with a relatively liberal setting presented in Appendix Table 6. The PcGets results in Table 2 are the final models for three samples, respectively, in Appendix Table 7, which clearly shows the variables included in the GUM and in the final model.

In Table 2, the BMA analysis for the whole sample yields a subset inclusive of 4 “other” variables (GDP90, POP90, ETHPOL and EURFRAC), 2 geographic variables (REGGEAP and AREA), 4 policy variables (CTRADE,

¹⁵The Chow tests are F tests and used to test parameter constancy. The Normality test, a Chi-squared statistic, is used to check the normality of the distribution of the residuals. The Heteroscedasticity test is for unconditional heteroscedasticity.

EXPPRIM, SDBMP and SDPI) and 5 institutional variables (CIVLEG, COMLEG, DURABLE, KKZ and PCI). Given no misspecification tests reject, the PcGets analysis for the whole sample yields a subset inclusive of 3 “other” variables (GDP90, POP90, and EURFRAC), 2 geographic variables (LATITUDE and AREA), 1 policy variable (SDTT) and 3 institutional variables (CIVLEG, KKZ and PCI). Both the BMA and Gets further confirm that direct effects of institutions, policy and geography on FD exist.

The BMA and Gets analyses on the whole sample unanimously suggest 3 “other” variables (GDP90, POP90, and EURFRAC), 1 geographic variable (AREA), and 3 institutional variables (CIVLEG, KKZ and PCI) are the main determinants for FD. The results from the developing country sample and the La Porta sample¹⁶ show that RESPOINT, TOPEN and SDPI should be in the models for FD. The analysis based on the La Porta sample provides some indication that the market share of stated-owned media (MEDSHARE), shareholders’ rights (SRIGHT) and creditors’ right (CRIGHT) are potential predictors for overall financial development.

In general, the results from the whole sample, strengthened by the results from two subsamples, provide strong evidence in support of the view discussed earlier that geography (represented by AREA and RESPOINT), policy (captured by TOPEN and SDPI) and institutions (captured by CIVLEG, KKZ and PCI) are all important for FD. The links between FD and AREA, TOPENDIX and KKZ suggest that countries with a smaller land area, more open trade policy and stronger institutions have a higher level of financial development. The initial GDP and population level are also found to be important for FD.

In Table 1 and 2, the BMA procedure has yielded posterior probabilities of inclusion for all candidate variables. A natural question to ask is about the structure of the models, especially the models with higher explanatory power. Table 3 lists the structure of the top ten models for FD in the whole sample in terms of posterior model probabilities, serving as a concrete illustration of model selection. A noteworthy point is that all these models have more than 10 possible predictors with geographic, policy and institutional

¹⁶Many experiments suggest that the results are sensitive to the inclusion of the variable EXPMANU.

and “other” variables present in all models. However, one should be aware of the dramatic model uncertainty, reflected by less than 5% posterior model probabilities for all top ten “best” models, which indicates the potential value of the BMA and Gets procedures for model selection as a systematic response to model uncertainty.

Moving on one step further, OLS regressions are used to estimate some of the best performing models in Table 4. The best model, that is the model with highest posterior probability in Table 3, is presented in column (4). The effects of GDP90 and POP90 on FD suggest that countries with a higher income level and a larger population size have more potential to develop financial markets. The relationship between regional dummies and FD show the East Asia and Pacific countries are positively associated with FD. The effect of country size on FD is negative. This is perhaps supported by the claim that land area determines transportation costs of manufactured exports, efficiency of supply chains and the flow of information and technology. A positive and statistically significant coefficient on CTRADE suggests trade openness may exert a significant role on FD. CIVLEG is found to have much stronger negative effects on FD than COMLEG, but this result should be regarded with some care. Countries with a common law legal origin in general witness better financial development as shown by the subsequent tables. The relationships between institutional factors, KKZ and PCI, and FD highlight the importance of efficient and transparent institutions, and a free and just society for FD. The effect is very strong, as shown by the standardized coefficients in the lower section of the table: a one standard deviation change in KKZ translates into more than 0.5 standard deviation of the FD measure. Stronger effects of PCI are observed.

In sum, the analysis above has suggested that institutions, policy and geography - taken as a group - jointly explain a substantial proportion of the variation in FD. When taken individually, at least AREA, TOPEN, CIVLEG, COMLEG, KKZ, PCI, GDP90 and POP90 are found to have significant influences on FD measure.

6 Empirical results (II): Specific financial developments

This section turns to study briefly the determinants of four specific indexes for financial development derived by using principal component analysis, namely, financial intermediary development (FDBANK), stock market development (FDSTOCK), financial efficiency development (FDEFF) and financial size development (FDSIZE). Bond market development (FDBOND) is also studied afterwards. The three samples are investigated for each index in which EURO1900 is only available for the developing country sample while SRIGHT, CRIGHT and MEDSHARE are only available for the La Porta dataset sample.

As in the previous section, the Gets model search is conducted with a relatively liberal strategy presented in Appendix Table 6.

The determinants of financial intermediary development (FDBANK) are reported in Table 5. The whole sample has 91 observations, the developing country sample has 70 observations and the La Porta sample has 40 observations¹⁷. The BMA and Gets analyses on the whole sample suggest FDBANK is positively related to initial income. East Asia and Pacific countries, Middle East and North African countries, and South Asian countries witness relative success in financial intermediary development. MINDIST is suggested to be important as well. The trade open policy index (TOPEN) and Frankel-Romer index (CTRADE) are significantly positively signed, suggesting financial intermediary development is boosted by more open trade policies. Three institutional variables (POLITY2, KKZ and PCI) are suggested to be determinants for FDBANK, consistent with a conventional view that better institutions are associated with better financial intermediary development. The analyses based on the developing country sample and La Porta sample in general confirm the findings for GDP90, REGEAP, REGMENA, TOPEN, KKZ and PCI. In addition, the analyses from the La Porta sample show that the shareholders' right and creditors' rights may be closely related to financial intermediary development.

¹⁷Many experiments suggest that the results are sensitive to the inclusion of the variable EXPMANU.

The determinants of stock market development (FDSTOCK) are reported in Table 6. The whole sample has 81 observations, the developing country sample has 50 observations and the La Porta sample has 49 observations. The BMA and Gets analyses on the whole sample indicate that FDSTOCK is positively related to the initial population and the ethnic polarisation index, while it is negatively related to the European first language fractionalisation index (EURFRAC). East Asia and Pacific countries experience a rise in stock market development. Land area is also important for FDSTOCK. Among other policy factors, TOPEN and SDGR are almost suggested by two methods to be in the model - this finding is also supported in the developing country sample and La Porta sample. The usual claim concerning the positive impacts of open trade policy on financial development applies here. The significantly negative effect of output volatility on FDSTOCK means that macroeconomic mismanagement might exert an adverse effect on FDSTOCK. Three institutional variables (DURABLE, KKZ and PCI) are suggested to be the main determinants for FDSTOCK. The analyses based on the developing country sample and the La Porta sample supports the idea that more open trade policies and better institutions promote stock market development.

The determinants of financial efficiency (FDEFF) are reported in Table 7. The whole sample has 79 observations, the developing country sample has 48 observations and the La Porta sample has 49 observations. Note that the lower value of FDEFF is associated with a higher level of financial efficiency development as discussed in Section 3.2. The BMA and Gets analyses on the whole sample suggest that RELIGION is significantly related to FDEFF. East Asia and Pacific countries, South Asian countries, Middle East and North African countries tend to have more efficient financial markets. Financial markets are more efficient in countries where institutional quality (captured by KKZ) is higher. The results from two subsamples show that initial GDP and population are also important for FDEFF.

The determinants of financial size development (FDSIZE), also called financial depth, are reported in Table 8. The whole sample has 73 observations, the developing country sample has 51 observations and the La Porta sample has 42 observations. The BMA and Gets analyses on the

whole sample suggest that financial depth in a country is positively related to the initial population. The West Europe and North American countries including most developed countries seem to witness a decline in financial depth. Countries with a larger land area experience relatively less financial size development. Countries with a more open trade policy are found to have better financial development in terms of size. Financial depth is also associated with a stable political system (captured by DURABLE) and less political constraints on the executive (captured by PCI). Most of these findings are supported by analyses based on a developing country sample and the La Porta sample. In addition, the analyses from the La Porta sample show that financial depth might be closely related to shareholders' rights.

We now turn to the case of bond market development. Since only size measures for bond market development, bond market capitalisation, are available in the World Bank Financial Development and Financial Structure Database (2005) with incomplete data for many developing countries, the above financial development measures do not include indexes of bond market development. Appendix Table 8 presents the specific BMA and PcGets analyses for bond market development, denoted by FDBOND, which is the sum of the private and public bonds share over GDP in 1990s. The analyses are based on the La Porta sample of 35 countries subject to data availability. The results show that initial GDP level (GDP90), language fractionalisation index (LANGUAGE), East Asia and Pacific countries (REGEAP), population density in coastal area (POP100CR) and terms of trade volatility (SDTT) and governance index (KKZ) may influence bond market development. The results support previous findings in terms of institutions, policy and geography being important for financial development, but further study critically depends on the availability of additional data.

7 Conclusions

The level of financial development in a country is primarily determined by its institutional quality, government policies, geographic endowments, its income level and cultural characteristics. The analysis reaches this conclusion by a joint application of recently developed Bayesian Model Averaging and

General-to-specific methods using 39 variables, which are more than any previous study. The joint application adopted here has the potential for incorporating their merits and minimising their limits, showing advantages in mitigating arbitrary choices and increasing precision in model selection. Furthermore, conducted using a large variable set, the joint application is expected to produce more reliable findings.

To explore the structural causes of financial development, the variables considered here are either predetermined or evolving slowly over time. Of 39 variables selected, 8 variables (initial income, initial population, land area, open trade policy, civil law countries, common law countries, a governance index and a political constraint index) are found to be associated with financial development. The implications of the findings are interesting and may deserve further thoughts. For example, more open trade policies are associated with greater financial development, and better institutional quality and higher levels of civil liberties and political rights are also associated with higher levels of financial development. The finding that the legal origins influence financial development supports the emphasis on legal determinants of financial development of La Porta et al. (1998).

Admittedly, the selection of variables can never be exhaustive although the number of economic, political, and geographic variables included is much more than in any previous research. This highlights the significance of placing emphasis on the role of institutions, policy and geography in financial development. The institutions, policy, geography, income level and cultural characteristics in a country constitute the economic environment in which entrepreneurs make decisions for their investment projects and external financing, consumers decide on consuming and saving, and the financial intermediaries channel the funds from savers to borrowers. Essentially, good institutional quality leads to the efficient supply of external finance while ill-functioning institutions as well as particular cultural characteristics mainly form structural impediments to the supply of external finance. On the demand side, sound economic policies, industrialisation and economic growth mainly stimulate the demand for more and cheaper credit. So does an advantageous geographic endowment, which facilitates external trade and manufacturing. On the contrary, macroeconomic mismanagement could dis-

courage the demand for external finance. Adverse geographic characteristics may become structural impediments to demand. The fundamental roles of institutions, policy and geography on financial development appear on either the demand side or supply side of financial development.

Findings on institutions, policy and geography as a whole being all important for financial development have significant implications for developing financial markets. The significant effects of the structural factors which are relatively time-invariant means that any effort of the government to better institution quality, implement more open trade and sound macroeconomic policies, and improve geographic infrastructure can stimulate financial development in the long run. Further research, as in Abiad and Mody (2005), is needed to explore what causes governments to undertake financial reforms aimed at financial development.

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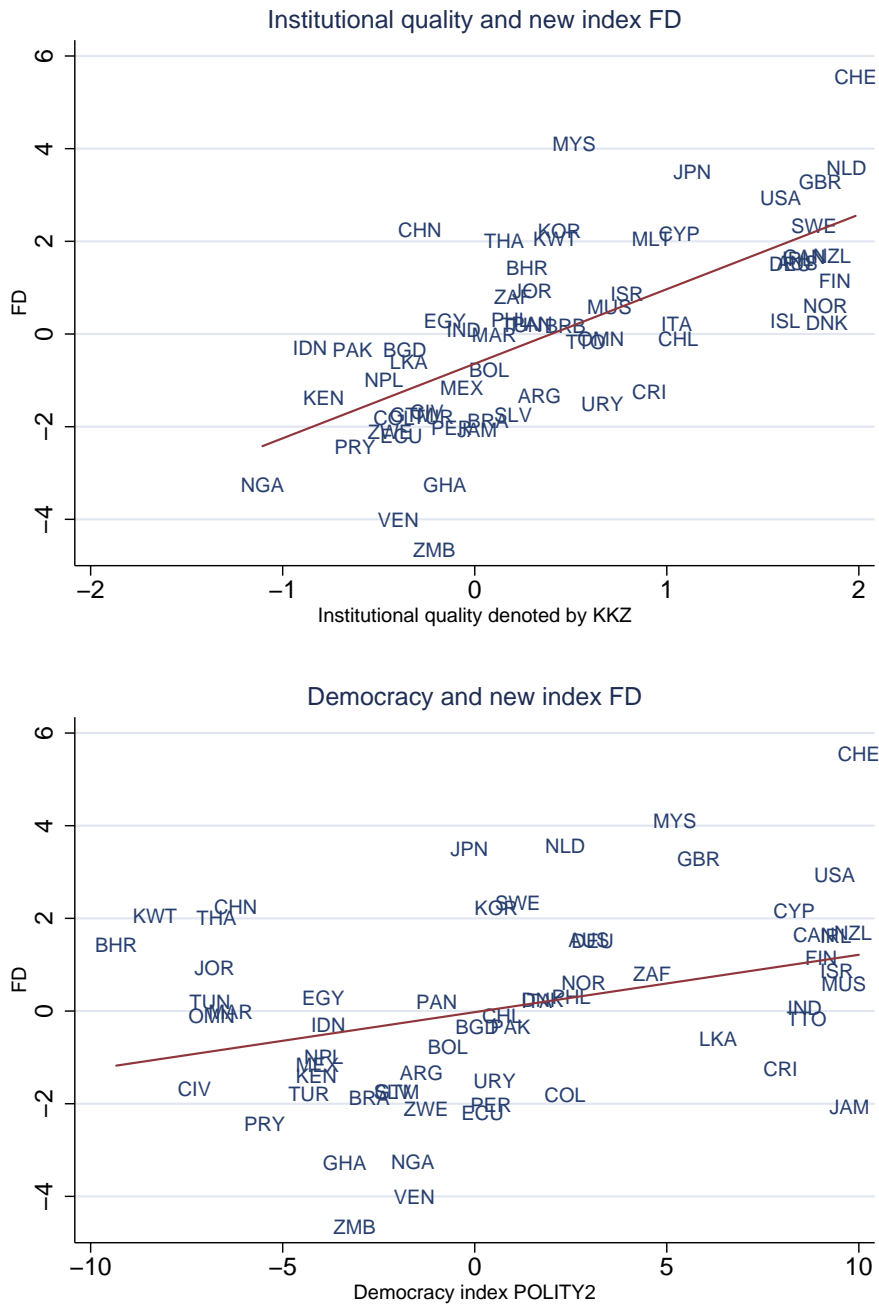
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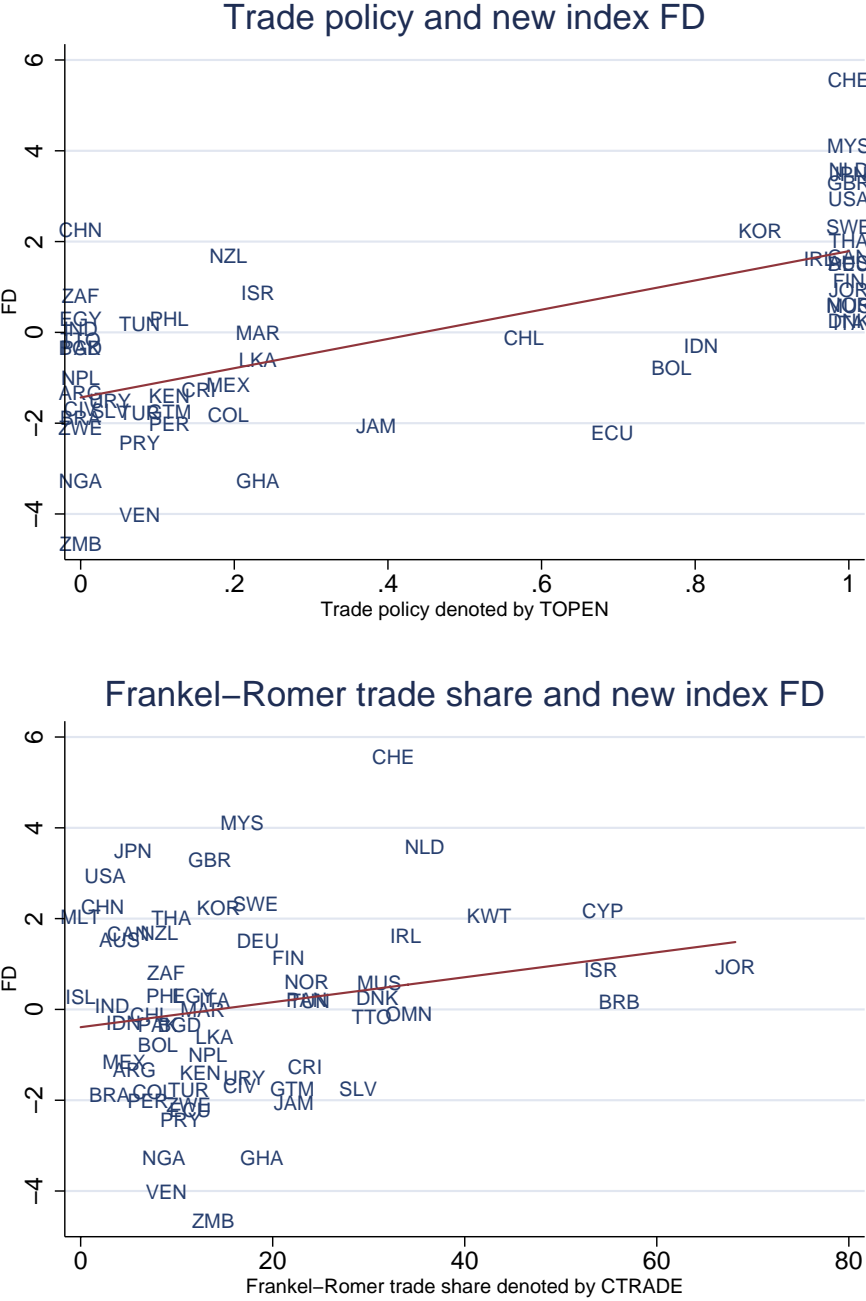
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Figure 1: Scatter plots of institutions and financial development



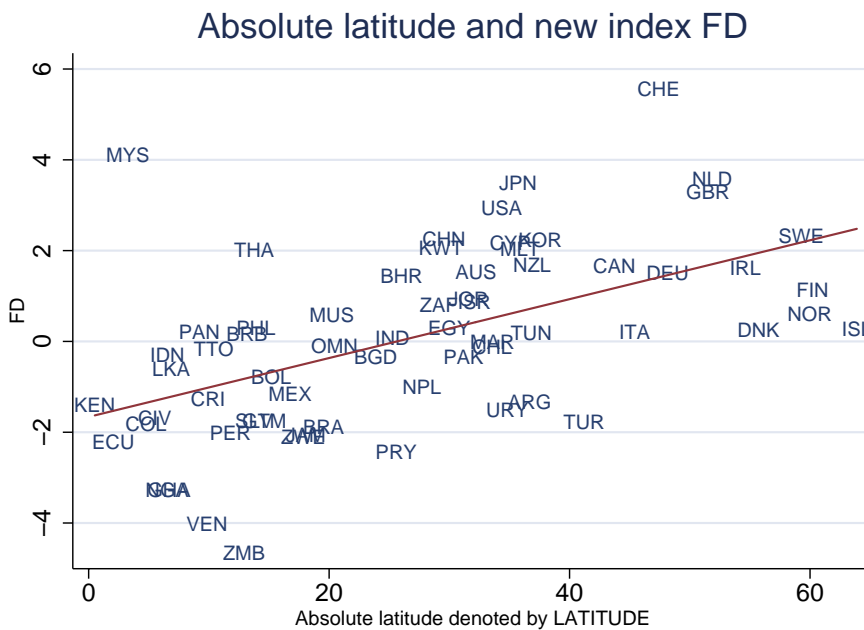
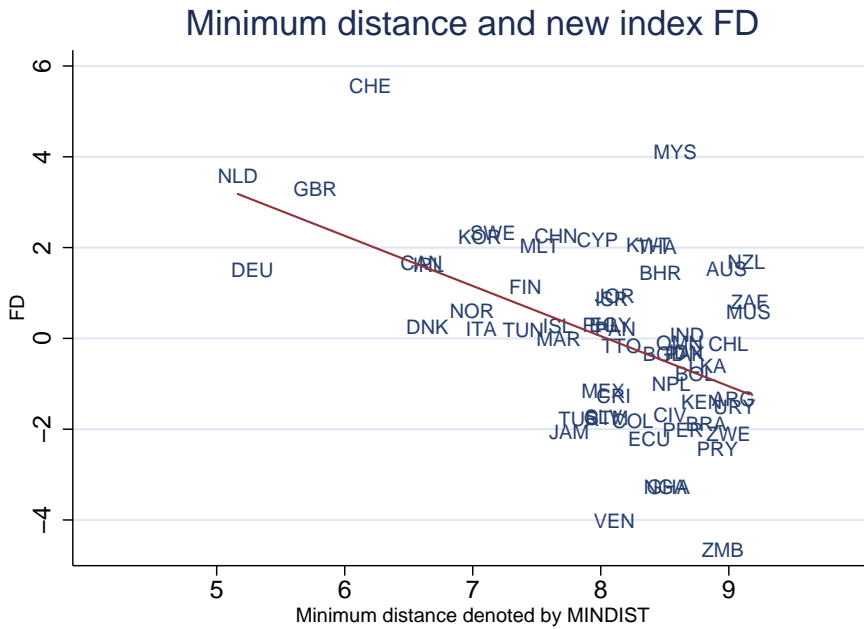
Note: Variables and data sources are described in Appendix Table 1. These figures show scatter plots of the institutional quality denoted by KKZ, and the democracy index POLITY2, against the new index FD.

Figure 2: Scatter plots of policy and financial development



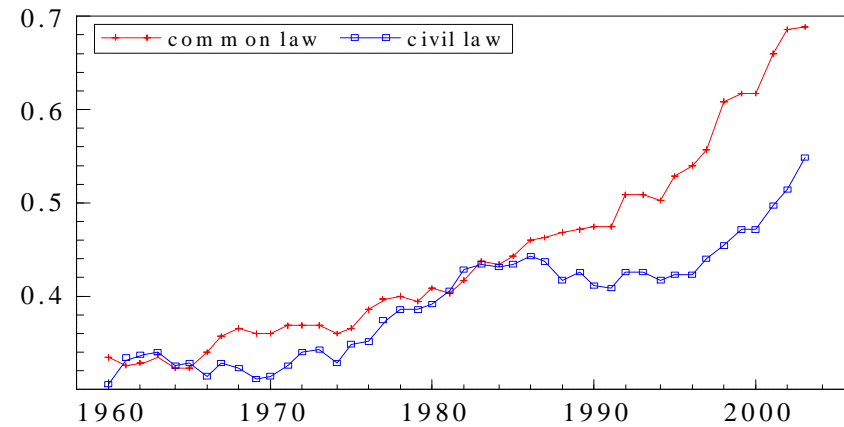
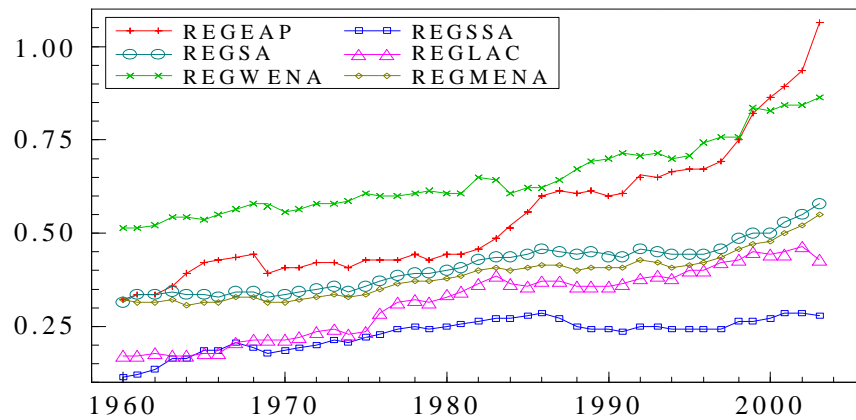
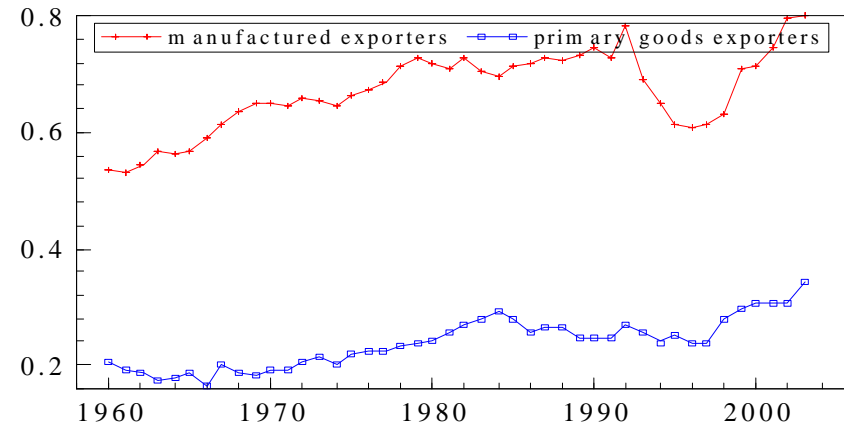
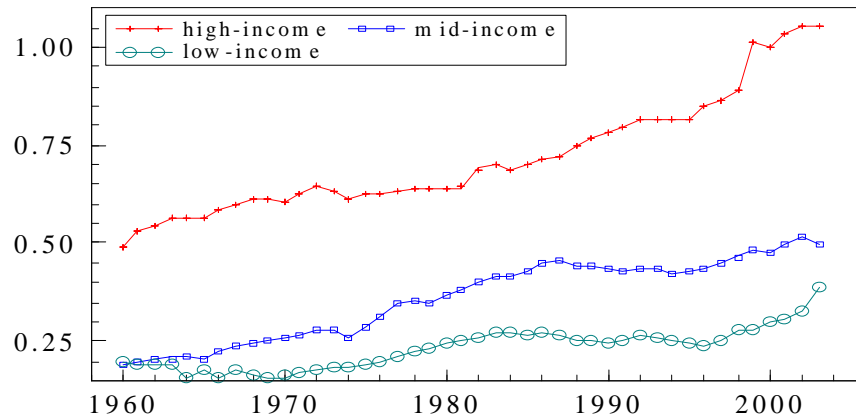
Note: Variables and data sources are described in Appendix Table 1. These figures show scatter plots of the trade policy index from Gallup et al. (1999), and the trade share constructed by Frankel and Romer (1999), against the new index FD.

Figure 3: Scatter plots of geography and financial development



Note: Variables and data sources are described in Appendix Table 1. These figures show scatter plots of the logarithm of minimum distance, and the absolute latitude, against the new index FD.

Figure 4: Averaged Liquid liability by different country group over 1960-2003



Note: Variable descriptions are from Appendix Table 1. These figures plot the averaged liquid liabilities by different income groups in the upper-left chart, different exporting countries in the upper-right chart, different regions in the lower-left chart and countries with different law traditions in the lower-right chart over 1960-2003.

Table 1. Determinants of FD by using BMA.

Sample	Whole			Whole			Whole		
Countries	64			64			64		
Model	1			2			3		
Variable	PIPs	Sign	MC3	PIPs	Sign	MC3	PIPs	Sign	MC3
CONSTANT	1.000	(-)		1.000	(-)		1.000	(-)	
GDP90	0.466	(+)	0.342	0.941	(+)	0.488	0.744	(+)	0.689
POP90	0.000		0.026	0.969	(+)	0.839	0.941	(+)	0.517
ETHPOL	0.004		0.039	0.649	(+)	0.070	0.906	(+)	0.035
ETHNIC	0.000		0.028	0.014		0.962	0.000		0.333
RELIGION	1.000	(+)	0.029	0.056		0.071	0.099		0.029
LANGUAGE	0.000		0.056	0.036		0.995	0.000		0.962
EURFRAC	0.000		0.992	0.706	(-)	0.040	0.982	(-)	0.089
REGAEP	0.186		0.132	0.726	(+)	0.958			
REGMENA	0.314	(+)	0.176	0.006		0.060			
REGSA	0.186		0.110	0.053		0.037			
REGSSA	0.879	(-)	0.946	0.642	(-)	0.049			
REGLAC	0.872	(-)	0.942	0.385	(-)	0.071			
REGWENA	0.204	(-)	0.175	0.036		0.094			
LANDLOCK	0.000		0.065	0.003		0.208			
LATITUDE	0.000		0.056	0.386	(-)	0.063			
AREA	0.073		0.034	0.975	(-)	0.831			
MINDIST	0.030		0.032	0.012		0.623			
POP100CR	0.051		0.027	0.400	(-)	0.056			
RESPOINT			0.037	0.005		0.025			
TOPEN	0.850	(+)	0.927				0.045		0.309
CTRADE	0.099		0.049				0.215	(+)	0.050
EXPMANU	0.000		0.030				0.000		0.025
EXPPRIM	0.409	(-)	0.175				0.999	(-)	0.964
SDGR	0.000		0.031				0.000		0.069
SDBMP	0.252	(-)	0.192				0.000		0.126
SDPI	0.076		0.030				0.064		0.053
SDTP	0.000		0.023				0.398	(+)	0.228
SDTT	0.329	(-)	0.201				0.045		0.026
CIVLEG				0.589	(-)	0.740	0.461	(-)	0.867
COMLEG				0.361	(-)	0.358	0.128		0.467
POLITY2				0.291	(+)	0.051	0.258	(+)	0.050
DURABLE				0.300	(+)	0.058	0.022		0.031
FREE				0.020		0.069	0.006		0.084
KKZ				0.988	(+)	1.000	1.000	(+)	0.999
PCI				0.963	(-)	0.996	0.924	(-)	0.953
EURO1900									
MEDSHARE									
SRIGHT									
CRIGHT									

Note: The dependent variable FD is the aggregate index of overall financial development over period, 1990-99. The variable description is in Appendix Table 1. The Bayesian Model Averaging (BMA) yields the posterior probabilities of inclusion (either "PIPs" or "MC3"), the total posterior model probabilities for all models including a given variable, and the sign certainty index of a relationship ("Sign"). A sign is given to the PIPs greater than 0.2. No sign govern means the sign of estimated relationship being uncertain. Any MC3 greater than 0.2 is in bold. The PIPs is due to Raftery (1995) while the MC3 is due to Hoeting et al. (1994) which also identify the outliers.

Table 2. Determinants of FD

Sample Method Variable	Whole Sample				Developing Country Sample				La Porta Sample			
	BMA		PcGets		BMA		PcGets		BMA		PcGets	
	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value
CONSTANT	1.000	(-)	-10.563	-4.074	1.000	(-)	-15.723	-5.932	1.000	(+)		
GDP90	0.946	(+)	1.391	4.403	1.000	(+)	2.049	7.192	0.012			
POP90	0.996	(+)	0.705	5.856	0.994	(+)	0.248	2.855	1.000	(+)	1.314	8.610
ETHPOL	0.999	(+)			0.007				0.039		3.131	4.652
ETHNIC	0.000				0.027				0.942	(+)		
RELIGION	0.004				0.492	(-)			0.067			
LANGUAGE	0.009				0.998	(+)	1.487	2.941	0.054			
EURFRAC	0.998	(-)	-1.287	-3.694	0.002				0.978	(-)	-3.988	-6.216
REGEAP	0.999	(+)			0.001				0.093		-4.224	-3.847
REGMENA	0.001				0.998	(-)	3.353	7.548	0.868	(+)		
REGSA	0.027				0.062				0.189		-3.000	-4.524
REGSSA	0.029				0.465	(-)			0.053			
REGLAC	0.019				0.008				0.069		2.435	3.317
REGWENA	0.002								0.965	(-)	-7.789	-5.614
LANDLOCK	0.009				0.005				0.993	(+)	2.562	3.979
LATITUDE	0.011		-0.041	-2.948	0.322	(-)			0.051			
AREA	0.992	(-)	-0.417	-4.188	0.817	(-)			1.000	(-)	-0.416	-4.095
MINDIST	0.003				0.046				0.241	(-)		
POP100CR	0.000				0.377	(+)			0.043		0.015	2.592
RESPOINT	0.023				1.000	(-)	3.353	7.548	0.434	(-)	-2.960	-4.633
TOPEN	0.001				1.000	(+)	1.990	5.192	0.985	(+)	1.854	3.512
CTRADE	0.492	(+)			0.903	(+)			0.050		0.034	2.421
EXPMANU	0.000				0.064						1.246	4.264
EXPPRIM	0.996	(-)			0.002				0.077		1.416	2.597
SDGR	0.000				0.005				0.980	(-)	-0.571	-3.875
SDBMP	0.346	(-)			0.658	(-)			0.120		0.004	4.876
SDPI	0.700	(+)			1.000	(+)	0.001	2.193	0.504	(-)	-0.035	-5.076
SDTP	0.033				0.445	(-)			0.177		-1.390	-2.263
SDTT	0.130		-0.024	-3.270	0.000				0.037		0.036	3.222
CIVLEG	0.529	(-)	-0.927	-3.396	1.000	(-)	-4.562	-5.239	0.074			
COMLEG	0.205	(+)			1.000	(+)	-3.445	-3.879	0.074		1.687	4.452
POLITY2	0.058				0.992	(-)	0.164	4.382	0.059			
DURABLE	0.764	(+)			0.000				0.686	(+)		
FREE	0.002				0.035				0.035		-0.453	-3.033
KKZ	1.000	(+)	1.846	5.064	0.370	(+)			0.992	(+)	4.849	9.156
PCI	0.995	(-)	-5.363	-5.184	1.000	(-)	-5.391	-4.026	0.995	(-)	-13.547	-7.002
EURO1900					0.002							
MEDSHARE									0.405	(-)		
SRIGHT									0.753	(+)		
CRIGHT									0.066		-0.374	-3.169
RSS				51.11				17.06				3.78
sigma				0.97				0.73				0.50
R^2				0.80				0.86				0.98
Radj^2				0.77				0.82				0.94
LogLik				7.20				20.84				47.21
AIC				0.09				-0.40				-1.11
HQ				0.22				-0.22				-0.73
SC				0.42				0.08				-0.05
Chow test 1												
Chow test 2			0.66	0.68			1.35	0.28				
Normality test			8.46	0.01			1.76	0.41			1.35	0.51
Hetero test												

Note: The dependent variable FD is the aggregate index of overall financial development over period, 1990-99. The variable description is in Appendix Table 1. There are 64 observations in the whole sample, 44 observations in the developing country sample and 40 observations in the La Porta sample. The BMA analysis yields the posterior probabilities of inclusion (PIPs) and the sign certainty index of a relationship (Sign). No sign given means the sign of estimated relationship being uncertain. each variable. The PcGets analysis yield coefficient and t-value for the variables in the final model. See p31 for the description of PcGets output.

Table 3. Top 10 Models and Their Posterior Probabilities for FD

	1	2	3	4	5	6	7	8	9	10
GDP90	*	*	*	*	*	*	*	*	*	*
POP90	*	*	*	*	*	*	*	*	*	*
ETHPOL	*	*	*	*	*	*	*	*	*	*
EURFRAC	*	*	*	*	*	*	*	*	*	*
REGEAP	*	*	*	*	*	*	*	*	*	*
AREA	*	*	*	*	*	*	*	*	*	*
CTRADE			*	*	*		*	*		
EXPPRIM	*	*	*	*	*	*	*	*	*	*
SDBMP			*	*	*		*		*	
SDPI	*	*	*	*	*		*	*	*	*
SDTT										
CIVLEG	*	*	*	*		*			*	
COMLEG					*			*		*
DURABLE	*		*	*	*		*	*	*	*
KKZ	*	*	*	*	*	*	*	*	*	*
PCI	*	*	*	*	*	*	*	*	*	*
PMP	0.048	0.042	0.042	0.037	0.030	0.028	0.028	0.028	0.028	0.025

Note: This table presents the top ten models for FD, ranked by their posterior model probability (PMP) in the whole sample. The variable description is in Appendix Table 1.

Table 4. Geography, policy, institutions and FD

	(1)	(2)	(3)	(4)
CONSTANT	-15.159 [5.87]**	-8.220 [2.95]**	-10.874 [3.33]**	-8.056 [3.16]**
GDP90	1.312 [6.25]**	0.990 [2.65]*	1.000 [2.93]**	0.958 [3.01]**
POP90	0.521 [3.66]**	0.584 [4.75]**	0.371 [3.12]**	0.512 [4.72]**
ETHPOL	1.117 [1.88]	1.584 [2.89]**	1.029 [1.65]	1.496 [3.17]**
EURFRAC	-0.801 [2.32]*	-1.138 [3.84]**	-1.143 [3.68]**	-1.100 [4.16]**
REGEAP	1.961 [4.61]**	1.277 [3.29]**		1.239 [3.92]**
AREA	-0.177 [1.58]	-0.457 [4.72]**		-0.412 [4.41]**
CTRADE	0.044 [4.07]**		0.025 [2.01]	
EXPPRIM	-0.609 [1.54]		-0.970 [3.11]**	-0.943 [4.06]**
SDBMP	-0.001 [3.79]**		0.000 [0.22]	
SDPI	0.001 [4.12]**		0.001 [3.51]**	0.001 [4.50]**
SDTT	-0.010 [1.20]		-0.010 [0.93]	
CIVLEG		-1.159 [2.22]*	-1.712 [3.33]**	-0.600 [2.49]*
COMLEG		-0.656 [1.28]	-0.998 [1.96]	
DURABLE		0.018 [1.66]	0.011 [0.73]	0.017 [1.54]
KKZ		1.489 [4.40]**	1.237 [3.30]**	1.445 [5.12]**
PCI		-4.006 [4.29]**	-3.791 [3.98]**	-4.258 [4.90]**
Standardised coefficients				
ETHPOL	0.49	0.72	0.45	0.68
EURFRAC	-0.46	-0.62	-0.63	-0.61
AREA	-0.15	-0.29		-0.26
CTRADE	-0.04		-0.05	
SDBMP	-0.06		-0.06	
SDPI	-0.06		-0.06	-0.06
SDTT	-0.07		-0.07	
DURABLE		-0.05	-0.06	-0.05
KKZ		0.68	0.55	0.66
PCI		-2.05	-1.94	-2.17
Observations	64	64	64	64
R-square	0.740	0.820	0.790	0.860

Note: The models are estimated by OLS. The dependent variable is FD, over 1990-99. The t-values are reported in brackets. Variable descriptions are from Appendix Table 1. The standardised coefficients show the change of a standard deviation of FD due to a one standard deviation change in a variable for those other than initial GDP and population, binary variables.

Table 5. Determinants of FDBANK

Sample Method Variable	Whole Sample				Developing Country Sample				La Porta Sample			
	BMA		PcGets		BMA		PcGets		BMA		PcGets	
	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value
CONSTANT	1.000	(-)	-2.523	-1.579	1.000	(-)	-4.218	-2.614	1.000	(+)		
GDP90	0.832	(+)	0.449	2.445	0.413	(+)	0.568	2.598	0.175			
POP90	0.017				0.000				0.858	(+)		
ETHPOL	0.012				0.008				0.692	(-)		
ETHNIC	0.001				0.000				0.852	(+)		
RELIGION					0.004				0.203	(-)		
LANGUAGE	0.000				0.000				0.060			
EURFRAC	0.000				0.000				0.326	(-)	-0.967	-2.751
REGEAP	1.000	(+)	1.979	6.770	0.929	(+)	1.879	5.125	0.268	(+)		
REGMENA	1.000	(+)	1.327	4.115	0.928	(+)	2.025	4.534	1.000	(+)		
REGSA	1.000	(+)	2.123	5.019	0.926	(+)			0.344	(+)		
REGSSA	0.001				0.072				0.336	(+)		
REGLAC	0.000				0.072				0.144			
REGWENA	0.001								0.767	(-)	-1.750	-3.295
LANDLOCK	0.008		-0.224	-0.817	0.259	(-)			0.734	(+)		
LATITUDE	0.000				0.000				0.048			
AREA	0.000				0.000				1.000	(-)		
MINDIST	0.804	(-)	-0.205	-2.630	0.000				0.343	(-)	-0.078	-2.270
POP100CR	0.001				0.000				0.257	(-)		
RESPOINT	0.361	(-)			0.321	(-)	-0.680	-2.674	0.770	(-)		
TOPEN	0.703	(+)	0.790	2.118	0.682	(+)			0.969	(+)	1.942	4.213
CTRADE	0.862	(+)	0.020	2.730	0.158				0.949	(-)		
EXPMANU	0.095				0.184							
EXPPRIM	0.126				0.134				0.083			
SDGR	0.000				0.000				0.940	(-)		
SDBMP	0.005				0.035				0.175			
SDPI	0.001		0.000	0.383	0.035				0.165			
SDTP	0.000				0.000				0.129			
SDDT	0.109				0.143				0.062			
CIVLEG	0.043				0.013				0.116			
COMLEG	0.182		-0.367	-1.743	0.035				0.107		0.691	2.305
POLITY2	0.335	(+)	0.060	2.079	0.059		0.113	2.709	0.037			
DURABLE	0.000				0.000				0.028			
FREE	0.031				0.033				0.024			
KKZ	1.000	(+)	0.733	3.628	1.000	(+)	0.753	2.737	1.000	(+)	1.349	4.850
PCI	0.828	(-)	-3.433	-3.622	0.430	(-)	-4.340	-2.944	0.832	(-)		
EURO1900					0.045							
MEDSHARE									0.130			
SRIGHT									0.662	(+)		
CRIGHT									0.729	(-)		
RSS				51.30				51.41				29.21
sigma				0.82				0.94				0.93
R^2				0.82				0.65				0.76
Radj^2				0.79				0.61				0.72
LogLik				26.08				8.82				6.29
AIC				-0.27				-0.02				-0.01
HQ				-0.11				0.08				0.08
SC				0.12				0.23				0.24
Chow test 1			1.63	0.07								
Chow test 2			2.94	0.01			2.27	0.05				
Normality test			0.65	0.72			1.45	0.48			0.93	0.63
Hetero test			20.25	0.51								

Note: The dependent variable FDBANK is the index of financial interdediary development over period, 1990-99. Variable description is in Appendix Table 1. There are 91 observations in the whole sample, 70 observations in the developing country sample and 40 observations in the La Porta sample. The BMA analysis yields the posterior probabilities of inclusion (PIPs) and the sign certainty index of a relationship (Sign). No sign given means the sign of estimated relationship being uncertain. each variable. The PcGets analysis yields coefficient and t-value for the variables in the final model. See p31 for the description of PcGets output.

Table 6. Determinants of FDSTOCK

Sample Method Variable	Whole Sample				Developing Country Sample				La Porta Sample			
	BMA		PcGets		BMA		PcGets		BMA		PcGets	
	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value
CONSTANT	1.000	(-)			1.000	(-)	-8.131	-6.444	1.000	(-)		
GDP90	0.183				1.000	(+)	0.645	4.911	0.045		-0.669	-3.836
POP90	1.000	(+)	0.435	8.414	1.000	(+)	0.290	7.118	0.919	(+)	0.732	4.701
ETHPOL	0.985	(+)	0.791	3.379	0.000				0.832	(+)	1.490	2.897
ETHNIC	0.000				0.014				0.009			
RELIGION	0.008				0.974	(+)	0.989	3.567	0.232	(+)		
LANGUAGE	0.039				0.005				0.046			
EURFRAC	0.210	(-)	-0.611	-3.016	0.977	(-)			0.783	(-)	-0.932	-2.783
REGEAP	0.937	(+)	-1.301	-3.001	0.056				0.016		-0.833	-1.689
REGMENA	0.062		-1.653	-4.024	0.021		-1.714	-2.791	0.008			
REGSA	0.035		-2.485	-5.434	0.351	(-)	-1.623	-2.807	0.721	(-)	-2.367	-3.686
REGSSA	0.022		-1.427	-3.245	0.892	(-)	-2.098	-3.419	0.034			
REGLAC	0.000		-1.128	-2.828	0.012		-1.562	-3.196	0.000			
REGWENA	0.063		-1.716	-3.664					0.033		-1.258	-2.367
LANDLOCK	0.000				0.007				0.004			
LATITUDE	0.000				0.000				0.040			
AREA	0.985	(-)	-0.245	-6.484	0.000				0.935	(-)	-0.462	-4.611
MINDIST	0.120				0.000				0.722	(+)	0.173	2.361
POP100CR	0.037				0.006				0.000			
RESPOINT	0.007				0.532	(-)			0.000			
TOPEN	0.191		0.578	2.686	1.000	(+)	0.901	4.497	0.098		0.647	1.361
CTRADE	0.000				0.000				0.081			
EXPMANU	0.901	(+)			0.003				0.165		0.602	1.976
EXPPRIM	0.003				0.000				0.065		0.858	1.948
SDGR	0.141		-0.129	-3.320	0.244	(-)	0.000	-3.172	0.555	(-)	-0.351	-3.908
SDBMP	0.014				0.736	(-)			0.252	(-)		
SDPI	0.000				0.068				0.076			
SDTP	0.745	(+)			0.001				1.000	(+)	1.847	3.524
SDTT	0.000				0.000				0.000			
CIVLEG	0.002				0.615	(-)			0.006			
COMLEG	0.111		0.582	4.032	0.133		0.321	2.212	0.005			
POLITY2	0.010				0.045				0.036			
DURABLE	0.216	(+)	0.013	3.003	0.040				0.688	(+)	0.022	2.619
FREE	0.134				0.164		0.161	2.983	0.051			
KKZ	0.976	(+)	0.701	4.687	0.463	(+)			0.894	(+)	1.503	4.310
PCI	0.245	(-)	-1.828	-4.349	0.000				0.006			
EURO1900					0.006		-0.813	-1.681				
MEDSHARE									0.249	(-)		
SRIGHT									0.108			
CRIGHT									0.002			
RSS				14.02				4.27				17.87
sigma				0.47				0.34				0.74
R^2				0.87				0.84				0.78
Radj^2				0.84				0.78				0.68
LogLik				71.04				61.53				24.71
AIC				-1.33				-1.90				-0.36
HQ				-1.13				-1.70				-0.12
SC				-0.83				-1.37				0.26
Chow test 1			1.20	0.33								
Chow test 2			0.88	0.54			1.47	0.23			0.15	0.96
Normality test							9.04	0.01			8.52	0.01
Hetero test			31.22	0.22								

Note: The dependent variable FDSTOCK is the index of stock market development over period, 1990-99. The variable description is in Appendix Table 1. There are 81 observations in the whole sample, 50 observations in the developing country sample and 49 observations in the La Porta sample. The BMA analysis yields the posterior probabilities of inclusion (PIPs) and the sign certainty index of a relationship (Sign). No sign given means the sign relationship being uncertain each variable. The PcGets analysis yields coefficient and t-value for the variables in the final model. See p31 for description of PcGets output.

Table 7. Determinants of FDEFF

Sample Method Variable	Whole Sample				Developing Country Sample				La Porta Sample			
	BMA		PcGets		BMA		PcGets		BMA		PcGets	
	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value
CONSTANT	1.000	(+)			1.000	(+)	9.145	4.200	1.000	(+)		
GDP90	0.010				0.996	(-)	-1.456	-6.307	0.243	(+)	0.661	4.845
POP90	0.041				0.856	(-)	-0.536	-5.097	0.672	(-)	-0.411	-3.726
ETHPOL	0.000				0.987	(+)			0.034			
ETHNIC	0.000				0.027				0.004			
RELIGION	0.375	(-)	-0.789	-1.839	0.061				0.479	(-)		
LANGUAGE	0.000				0.996	(-)	-0.975	-2.279	0.170			
EURFRAC	0.028				0.336	(-)			0.894	(+)	1.039	3.902
REGAAP	0.989	(-)	-1.397	-5.028	0.029				0.075			
REGMENA	0.962	(-)	-1.685	-5.377	1.000	(-)	-2.613	-7.077	0.716	(-)	-0.894	-2.535
REGSA	0.986	(-)	-2.326	-5.240	1.000	(-)			0.707	(-)		
REGSSA	0.033				0.006				0.029			
REGLAC	0.037				0.045				0.225	(+)		
REGWENA	0.777	(-)							0.071			
LANDLOCK	0.000				0.021				0.031			
LATITUDE	0.000				0.055				0.001			
AREA	0.021				0.851	(+)	0.411	4.515	0.289	(+)	0.160	2.168
MINDIST	0.066				0.024				0.015			
POP100CR	0.000				0.003				0.028			
RESPOINT	0.103		0.667	2.473	0.996	(+)	1.379	5.697	0.019			
TOPEN	0.052				0.996	(-)			0.030			
CTRADE	0.000				0.088				0.000			
EXPMANU	0.140				0.007				0.270	(-)		
EXPPRIM	0.926	(+)			0.000				0.001			
SDGR	0.002				0.089				0.755	(+)	0.149	2.443
SDBMP	0.021				0.000				0.000			
SDPI	0.134		0.000	-0.999	0.996	(-)	-0.001	-4.142	0.023			
SDTP	0.086		-0.332	-1.144	0.082				0.932	(-)	-2.084	-5.550
SDIT	0.058				0.003				0.121			
CIVLEG	0.034		1.773	3.340	0.996	(+)	2.828	3.756	0.000			
COMLEG	0.102		2.127	3.725	0.996	(+)	2.133	2.789	0.000			
POLITY2	0.000				0.981	(-)	-0.144	-4.539	0.000			
DURABLE	0.000				0.059				0.000			
FREE	0.985	(-)			0.027				0.125			
KKZ	1.000	(-)	-1.119	-8.260	0.768	(-)			1.000	(-)	-2.044	-8.788
PCI	0.000				0.996	(+)	6.226	5.693	0.703	(+)		
EURO1900					0.159							
MEDSHARE									0.003			
SRIGHT									0.001			
CRIGHT									0.000			
RSS				40.53				14.68				19.89
sigma				0.77				0.64				0.70
R^2				0.74				0.82				0.82
Radj^2				0.70				0.76				0.79
LogLik				26.36				28.43				22.09
AIC				-0.41				-0.68				-0.57
HQ				-0.29				-0.51				-0.46
SC				-0.11				-0.22				-0.27
Chow test 1												
Chow test 2			2.27	0.01			0.70	0.60			0.45	0.77
Normality test			16.62	0.34			0.11	0.95			2.82	0.24
Hetero test												

Note: The dependent variable FDEFF is the index of financial efficiency development over period, 1990-99. The variable description is in Appendix Table 1. There are 79 observations in the whole sample, 48 observations in the developing country sample and 49 observations in the La Porta sample. The BMA analysis yields the posterior probabilities of inclusion (PIPs) and the sign certainty index of a relationship (Sign). No sign given means the sign of estimated relationship being uncertain each variable. The PcGets analysis yields coefficient and t-value for the variables in the final model. See p31 for the description of PcGets output.

Table 8. Determinants of FDSIZE

Sample Method Variable	Whole Sample				Developing Country Sample				La Porta Sample			
	BMA		PcGets		BMA		PcGets		BMA		PcGets	
	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value	PIPs	Sign	Coeff	t-value
CONSTANT	1.000	(-)			1.000	(-)	-11.170	-4.993	1.000	(-)		
GDP90	1.000	(+)			0.809	(+)	0.356	2.073	0.825	(+)		
POP90	0.999	(+)	0.286	3.708	0.961	(+)	0.247	4.185	0.624	(+)		
ETHPOL	0.068				0.282	(+)			0.009			
ETHNIC	0.000				0.093		1.558	3.456	0.138			
RELIGION	0.000				0.808	(+)			0.000			
LANGUAGE	0.342	(+)			0.392	(+)			0.346	(+)		
EURFRAC	0.049		-0.535	-2.043	0.043				0.005			
REGEAP	0.351	(+)			0.000				0.505	(+)		
REGMENA	0.552	(+)			1.000	(+)			0.111			
REGSA	0.016		-0.568	-1.324	0.018		-1.621	-4.294	0.137			
REGSSA	0.003				0.152		-0.607	-2.369	0.003			
REGLAC	0.013				0.042				0.045			
REGWENA	0.554	(-)	-1.224	-3.319					0.640	(-)	-1.209	-2.479
LANDLOCK	0.000				0.105				0.987	(+)		
LATITUDE	0.000				0.002		0.043	4.284	0.063			
AREA	0.999	(-)	-0.267	-4.605	0.002				0.379	(-)		
MINDIST	0.008				0.016		0.511	2.483	0.594	(-)		
POP100CR	0.040				0.269	(+)	0.012	3.817	0.080			
RESPOINT	0.001				0.039		-0.388	-2.045	0.034			
TOPEN	0.987	(+)	1.353	4.065	1.000	(+)	1.313	4.984	0.616	(+)	1.971	5.253
CTRADE	0.000				0.076				0.843	(+)		
EXPMANU	0.000				0.005				0.001			
EXPPRIM	0.003				0.034				0.020			
SDGR	0.000				0.590	(-)			0.768	(-)		
SDBMP	0.218	(-)			0.491	(-)	-0.001	-3.310	0.032			
SDPI	0.000				0.000				0.854	(-)	-0.008	-2.716
SDTP	0.000				0.000				0.824	(-)		
SDTT	0.018				0.183				0.019			
CIVLEG	0.757	(-)			0.895	(-)	-0.752	-4.225	0.010			
COMLEG	0.243	(+)			0.029				0.010			
POLITY2	0.000				0.012				0.000			
DURABLE	0.244	(+)	0.022	3.009	0.000				0.005			
FREE	0.004				0.249	(+)			0.002			
KKZ	0.155		0.697	3.231	0.467	(+)			0.097			
PCI	0.888	(-)	-1.348	-1.921	0.008				0.879	(-)		
EURO1900					0.000		-2.359	-4.514				
MEDSHARE									0.015			
SRIGHT									1.000	(+)		
CRIGHT									0.001			
RSS				39.78				9.38				46.94
sigma				0.79				0.50				1.10
R^2				0.65				0.75				0.42
Radj^2				0.61				0.66				0.39
LogLik				22.15				43.18				-2.34
AIC				-0.36				-1.14				0.25
HQ				-0.25				-0.94				0.30
SC				-0.08				-0.61				0.38
Chow test 1			0.80	0.74								
Chow test 2			1.17	0.34			1.11	0.37				
Normality test			8.90	0.01			0.29	0.87			8.45	0.01
Hetero test												

Note: The dependent variable FDSIZE is the index of financial size development over period, 1990-99. The variable description is in Appendix Table 1. There are 73 observations in the whole sample, 51 observations in the developing country sample and 42 observations in the La Porta sample. The BMA analysis yields the posterior probabilities of inclusion (PIPs) and the sign certainty index of a relationship (Sign). No sign given means the sign of estimated relationship being uncertain each variable. The PcGets analysis yields coefficient and t-value for the variables in the final model. See p31 for the description of PcGets output.

Appendix Table 1. The variables

VARIABLE	DESCRIPTION	SOURCE
<u>Dependent variables</u>		
FD	Index for overall financial development. The first principal component of private credit(PRIVO), liquidity liability (LLY), commercial-central bank (BTOT), overhead cost (OVC), net interest margin (NIM), stock market capitalisation (MCAP), total value traded (TVT) and turnover ratio (TOR) in 1990s.	World Bank's Financial Structure and Economic Development Database (FSED), 2005
FDBANK	Index for financial intermediary development. The first principal component of PRIVO, LLY, BTOT, OVC and NIM in 1990s.	FSED, 2005
FDSTOCK	Index for stock market development. The first principal component of MCAP, TVT and TOR in 1990s.	FSED, 2005
FDEFF	Index for financial efficiency development. The first principal component of OVC, NIM, TVT and TOR in 1990s.	FSED, 2005
FDSIZE	Index for financial size development (financial depth). the first principal component of LLY and MCAP in 1990s.	FSED, 2005
FDBOND	index for bond market developpment, the sum of private bond and public bond share over GDP in 1990s.	FSED, 2005
<u>Policy variables</u>		
TOPEN	The proportion of years that a country is open to trade during 1965-90, by the criteria in Sachs and Warner (1995). A country is considered to be open if it meets minimum criteria on four aspects of trade policy: average tariffs must be lower than 40 percent, quotas and licensing must cover less than 40 percent of total imports, the black market premium must be less than 20 percent, and export taxes should be moderate.	Gallup et al. (1999)
CTRADE	Natural log of the Frankel-Romer measure of predisposition to external trade	Frankel and Romer (1999)
EXPMANU	Dummy for manufactured good exporting countries	Global Development Network Database in World Bank (GDN), 2002
EXPPRIM	Dummy for fuel, non-fuel primary good exporting countries	Global Development Network Database in World Bank (GDN), 2002

SDGR	Standard deviation of annual growth of real, chainweighted GDP per capita, 1960-89	Penn World Table 6.1 (PWT61) (Heston et al., 2002)
SDPI	Standard deviation of annual inflation (PI), 1960-89	World Development Indicators (WDI), 2004
SDBMP	Standard deviation of annual black market premium (BMP), 1960-89	GDN
SDTP	Standard deviation of trading partners' GDP per capita growth (% weighted average by trade share).	GDN
SDTT	Standard deviation of the first log-differences of a terms of trade index for goods and services.	GDN

Institutional variables

COMLEG	The dummy for British legal origin	GDN
CIVLEG	Legal origin dummy for French, German and Scandinavian	GDN
POLITY2	Index of democracy. It is called combined polity score, the democracy score minus the autocracy score. The democracy and autocracy scores are derived from the six authority characteristics (regulation, competitiveness and openness of executive recruitment; operational independence of chief executive or executive constraints; and regulation and competition of participation). Based on these criterion, each country is assigned a democracy score and autocracy score ranging from 0 to 10, accordingly, the POLITY2 ranges from -10 to 10 with higher values representing more democratic regimes. averaged over 1960-89.	PolityIV Database (Marshall et al., 2003)
DURABLE	Index of political stability based on the number of years since the last (3-point or greater) regime transition, averaged over 1960-89.	PolityIV Database (Marshall et al, 2003)
FREE	The average of indexes of civil liberties and political rights over 1972-89. The basic components of the index of civil liberties are (1) freedom of expression and belief, (2) association and organizational rights, (3) rule of law and human rights, (4) personal autonomy and economic rights. Rescaled from zero to one, with higher values indicating better civil liberties. The basic components of the index of political rights are (1) free and fair elections; (2) those elected rule; (3) there are competitive parties or other competitive political groupings; (4) the opposition has an important role and power; (5) the entities have self-determination or an extremely high degree of autonomy." Rescaled from zero to one, with higher values indicating better political rights.	Freedom House (FH), www.freedomhouse.org, 2003
KKZ	Average of six measures of institutional development: voice and accountability, political stability and absence of violence, government effectiveness, light regulatory burden, rule of law, and freedom from graft.	Kaufmann et al. (1999)

PCI	Political Constraints Index is a structurally derived measure of the feasibility of policy change (the extent to which a change in the preferences of any one actor may lead to a change in government policy).	Henisz (2000), 2002 version
EURO1900	The percent of population that was European or European descent in 1900.	Acemoglu et al (2001)
MEDSHARE	The index of media owned by the government, the average of the market share of state-owned newspapers and state-owned television stations. Market share of state-owned newspapers is the market share owned by the state out of the aggregate market share of the five largest daily newspapers (by circulation). Market share of state-owned television stations is the market share owned by the state out of the aggregate market share of the five largest television stations (by viewership)	Djankov et al. (2003)
SRIGHT	An index aggregating the shareholder rights which we labeled as “anti-director rights.” The index is formed by adding 1 when: (1) the country allows shareholders to mail their proxy vote to the firm; (2) shareholders are not required to deposit their shares prior to the General Shareholders’ Meeting; (3) cumulative voting or proportional representation of minorities in the board of directors is allowed; (4) an oppressed minorities mechanism is in place; (5) the minimum percentage of share capital that entitles a shareholder to call for an Extraordinary Shareholders’ Meeting is less than or equal to 10 percent (the sample median); or (6) shareholders have preemptive rights that can only be waived by a shareholders’ vote. The index ranges from 0 to 6.	La Porta et al. (1998)
CRIGHT	An index aggregating creditor rights. The index is formed by adding 1 when: (1) the country imposes restrictions, such as creditors’ consent or minimum dividends, to file for reorganization; (2) secured creditors are able to gain possession of their security once the reorganization petition has been approved (no automatic stay); (3) the debtor does not retain the administration of its property pending the resolution of the reorganization; (4) secured creditors are ranked first in the distribution of the proceeds that result from the disposition of the assets of a bankrupt firm. The index ranges from 0 to 4.	La Porta et al. (1998)
<u>Geographic variable</u>		
REGAP	Region dummy for East Asia and Pacific countries	GDN
REGMENA	Region dummy for Middle East and North African countries	GDN
REGSA	Region dummy for South Asian countries	GDN
REGSSA	Region dummy for Sub Sahara African countries	GDN

REGLAC	Region dummy for Latin America and Caribbean countries	GDN
REGWENA	Region dummy for West Europe and North American countries	GDN
LANDLOCK	Dummy for landlocked countries	GDN
LATITUDE	Latitude--absolute distance from equator	GDN
AREA	Area (in log) in square kilometers from World Bank (1997), except for Taiwan and Mexico from CIA (1997), with submerged land subtracted out.	Gallup et al. (1999)
POP100CR	Proportion of the population in 1994 within 100km of the coastline or ocean-navigable river.	Gallup et al. (1999)
MINDIST	The log of minimum distance from three capital-goods-supplying centres plus one.	Jon Haveman's International trade data. www.eiit.org
RESPOINT	Dummy for point source exporting countries.	Isham et al (2002)
<u>Other variables</u>		
GDP90	Log of real GDP per capita (chain) in 1990	PWT61
POP90	Log of total population in 1990	PWT61
ETHPOL	Index of ethnic polarisation	Reynal-Querol and Montalvo (2005)
ETHNIC	Index of ethnic fractionalization	Alesina et al.(2003)
RELIGION	Index of religious fractionalization	Alesina et al.(2003)
LANGUAGE	Index of language fractionalization	Alesina et al.(2003)
ERUFRAC	Index of the "first" language variables, corresponding to the fraction of the population speaking one of the major languages of Western Europe: English, French, German, Portuguese, or Spanish.	Hall and Jones (1999)
INCLW	low income countries	GDN
INCMID	upper middle and lower-middle income countries	GDN
INCHIGH	high income OECD and non-OECD countries	GDN

Appendix Table 2: Descriptive Statistics

Geography

	FD	LANDLOCK	LATITUDE	AREA	MINDIST	POP100CR	RESPOINT
FD	1.000						
LANDLOCK	-0.163	1.000					
LATITUDE	0.536	-0.059	1.000				
AREA	-0.095	-0.009	0.096	1.000			
MINDIST	-0.514	0.121	-0.503	0.007	1.000		
POP100CR	0.376	-0.583	0.293	-0.429	-0.340	1.000	
RESPOINT	-0.387	0.027	-0.296	0.108	0.252	-0.215	1.000

Institution

	FD	TOPEN	CTRADE	EXPMANU	EXPPRIM	SDGR	SDBMP	SDPI	SDTP	SDTT
FD	1.000									
TOPEN	0.660	1.000								
CTRADE	0.196	0.207	1.000							
EXPMANU	0.447	0.429	0.039	1.000						
EXPPRIM	-0.463	-0.482	-0.131	-0.320	1.000					
SDGR	-0.321	-0.359	0.213	-0.199	0.461	1.000				
SDBMP	-0.142	-0.092	-0.016	-0.063	0.122	0.121	1.000			
SDPI	-0.086	-0.017	-0.049	-0.056	0.189	0.047	0.716	1.000		
SDTP	-0.112	-0.130	-0.108	0.050	0.113	0.077	0.164	0.139	1.000	
SDTT	-0.411	-0.424	-0.133	-0.181	0.415	0.388	0.116	0.060	0.060	1.000

Policy

	FD	CIVLEG	COMLEG	POLITY2	DURABLE	FREE	KKZ	PCI
FD	1.000							
CIVLEG	-0.071	1.000						
COMLEG	0.037	-0.942	1.000					
POLITY2	0.325	-0.255	0.319	1.000				
DURABLE	0.451	-0.131	0.111	0.429	1.000			
FREE	-0.374	-0.002	-0.088	-0.725	-0.474	1.000		
KKZ	0.675	0.050	-0.018	0.525	0.506	-0.711	1.000	
PCI	0.318	0.032	0.031	0.722	0.404	-0.892	0.644	1.000

Others

	FD	GDP90	POP90	ETHPOL	ETHNIC	RELIGION	LANGUAGEURFRAC	
FD	1.000							
GDP90	0.627	1.000						
OPO90	0.070	-0.014	1.000					
ETHPOL	-0.169	-0.178	-0.107	1.000				
ETHNIC	-0.358	-0.516	-0.006	0.540	1.000			
RELIGION	0.151	-0.103	-0.030	0.207	0.231	1.000		
LANGUAGE	-0.159	-0.463	0.084	0.294	0.673	0.268	1.000	
EURFRAC	-0.082	0.346	-0.060	0.154	-0.138	0.081	-0.462	1.000

Appendix Table 3: The list of countries in the full sample

East Asia & Pacific		Middle East & North Africa		South Asia	
AUS	Australia	BHR	Bahrain	BGD	Bangladesh
CHN	China	DZA	Algeria	IND	India
FJI	Fiji	EGY	Egypt, Arab Rep.	LKA	Sri Lanka
HKG	Hong Kong, China	GRC	Greece	NPL	Nepal
IDN	Indonesia	IRN	Iran, Islamic Rep.	PAK	Pakistan
JPN	Japan	ISR	Israel		
KOR	Korea, Rep.	JOR	Jordan		
MAC	Macao	KWT	Kuwait		
MNG	Mongolia	LBN	Lebanon		
MYS	Malaysia	MAR	Morocco		
NZL	New Zealand	MLT	Malta		
PHL	Philippines	OMN	Oman		
PNG	Papua New Guinea	PRT	Portugal		
SGP	Singapore	QAT	Qatar		
THA	Thailand	SAU	Saudi Arabia		
TWN	Taiwan, China	TUN	Tunisia		
VNM	Vietnam				
Sub Sahara Africa		Latin America & Caribbean		West Europe & North America	
BDI	Burundi	ARG	Argentina	AUT	Austria
BEN	Benin	BOL	Bolivia	BEL	Belgium
BFA	Burkina Faso	BRA	Brazil	CAN	Canada
BWA	Botswana	BRB	Barbados	CHE	Switzerland
CIV	Cote d'Ivoire	CHL	Chile	CYP	Cyprus
CMR	Cameroon	COL	Colombia	DEU	Germany
ETH	Ethiopia	CRI	Costa Rica	DNK	Denmark
GHA	Ghana	DOM	Dominican Republic	ESP	Spain
KEN	Kenya	ECU	Ecuador	FIN	Finland
MDG	Madagascar	GTM	Guatemala	FRA	France
MLI	Mali	GUY	Guyana	GBR	United Kingdom
MOZ	Mozambique	HND	Honduras	IRL	Ireland
MRT	Mauritania	HTI	Haiti	ISL	Iceland
MUS	Mauritius	JAM	Jamaica	ITA	Italy
MWI	Malawi	MEX	Mexico	LUX	Luxembourg
NAM	Namibia	NIC	Nicaragua	NLD	Netherlands
NGA	Nigeria	PAN	Panama	NOR	Norway
RWA	Rwanda	PER	Peru	SWE	Sweden
SDN	Sudan	PRY	Paraguay	USA	United States
SEN	Senegal	SLV	El Salvador		
SLE	Sierra Leone	TTO	Trinidad and Tobago		
SWZ	Swaziland	URY	Uruguay		
TGO	Togo	VEN	Venezuela		
UGA	Uganda				
ZAF	South Africa				
ZMB	Zambia				
ZWE	Zimbabwe				

Appendix Table 4. The eigenvalue, proportion and eigenvector of each first principal component.

Measure	Eigenvalue	Proportion	LLY	PRIVO	BTOT	OVC	NIM	MCAP	TVT	TOR
FD	3.922	0.490	0.411	0.454	0.278	-0.357	-0.368	0.364	0.357	0.157
FDBANK	3.063	0.613	0.479	0.479	0.357	-0.437	-0.471			
FDSTOCK	1.986	0.662						0.535	0.676	0.506
FDEFF	2.160	0.540				0.546	0.561		-0.467	-0.411
FDSIZE	1.612	0.806	0.707					0.707		

Notes: The financial development measures are described in Table 1. The first principal component is the linear combination of the measures selected.

The eigenvalues are the variances of the (first) principal components. The eigenvectors give the coefficients of the standardised variables.

LLY = the ratio of liquid liabilities of financial system (currency plus demand and interest-bearing liabilities of banks and nonbanks) to GDP;

PRIVO = the ratio of credits issued to private sector by banks and other financial intermediaries to GDP;

OVC = the ratio of overhead costs to total assets of the banks; NIM = the bank interest income minus interest expense over total assets;

MCAP = the ratio of the value of domestic shares traded on domestic exchange to GDP;

TVT = the ratio of the value of domestic shares traded on domestic exchange to GDP;

TOR = the ratio of the value of domestic shares traded on domestic exchange to total value of listed domestic shares

Appendix Table 5: Imputation

Variables	Variables used to impute the missing data											
ETHPOL	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	RELIGION					
ETHNIC	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	RELIGION					
LANGUAGE	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	RELIGION					
EURFRAC	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	RELIGION					
CTRADE	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	INCLOW	INCMID	INCHIGH	EXPMANU	EXPPRIM	LANDLOCK
TOPEN	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	INCLOW	INCMID	INCHIGH	EXPMANU	EXPPRIM	LANDLOCK
AREA	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LANDLOCK	LATITUDE				
MINDIST	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LANDLOCK	LATITUDE				
POP100CR	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LANDLOCK	LATITUDE				
POP90	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LANDLOCK					
RESPOINT	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LANDLOCK	LATITUDE				
POLITY2	CIVLEG	COMLEG	LATITUDE									
DURABLE	CIVLEG	COMLEG	LATITUDE									
FREE	CIVLEG	COMLEG	LATITUDE									
KKZ	CIVLEG	COMLEG	LATITUDE									
PCI	CIVLEG	COMLEG	LATITUDE									
EURO1900	CIVLEG	COMLEG	LATITUDE	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA			
SRIGHT	CIVLEG	COMLEG	LATITUDE									
CRIGHT	CIVLEG	COMLEG	LATITUDE									
MEDSHARE	CIVLEG	COMLEG	LATITUDE									
GDP90	INCLOW	INCMID	INCHIGH	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LATITUDE		
SDGR	INCLOW	INCMID	INCHIGH	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LATITUDE		
SGBMP	INCLOW	INCMID	INCHIGH	CIVLEG	COMLEG	LATITUDE						
SDPI	INCLOW	INCMID	INCHIGH	CIVLEG	COMLEG	LATITUDE						
SDTP	INCLOW	INCMID	INCHIGH	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LATITUDE		
SDTT	INCLOW	INCMID	INCHIGH	REGGEAP	REGMENA	REGSA	REGSSA	REGLAC	REGWENA	LATITUDE		

Appendix Table 6: Setting for PcGets

expert significance:	0.075	0.075	0.75	0.075	0.01	0.005		
expert presearch:	0.75	1	0.5	0.075	0.075	0.05	0.05	1
expert block search:	1	1	1	1	1	1	1	1
expert choose specific:		"hq"						
expert split sample:	0.075	0.75	0.2	0.4	0.4			
expert outlier dection:		2.56						
expert tests:	1	1	0	1	0	0	1	
expert test options:	0.5	0.9	12	1	4	1	4	
set detect outliers:	"1"							
set0lagorder:	"0"							
set0topdown:	"1"							
set0bottomup:	"1"							
setsplitsample:	"1"							
setstrategy:	"expert",	1						
setreporting:	"0"							
estimate:	"GETS",	1	1	n	1			("n" denotes the sample size)

Note: A change has been made to the "liberal strategy" default setting by increasing the F pre-search testing (top-down) at step 1 from 0.75 to 1.

Appendix Table 7. Determinants of FD by using PcGets

Sample	Full Sample				Developing Country Sample				La Porta Sample			
Model	GUM		Final Model		GUM		Final Model		GUM		Final Model	
Variable	Coeff	t-value	Coeff	t-value	Coeff	t-value	Coeff	t-value	Coeff	t-value	Coeff	t-value
CONSTANT	-8.112	-1.338	-10.563	-4.074	-16.154	-1.151	-15.723	-5.932	0.000	0.000		
GDP90	1.326	2.660	1.391	4.403	2.250	2.823	2.049	7.192	-0.878	-0.316		
POP90	0.566	2.549	0.705	5.856	0.768	1.816	0.248	2.855	1.044	1.733	1.314	8.610
ETHPOL	0.497	0.511			-1.290	-0.671			2.592	0.672	3.131	4.652
ETHNIC	-0.774	-0.557			0.282	0.137			0.111	0.032		
RELIGION	-0.214	-0.188			-0.386	-0.156			2.183	0.400		
LANGUAGE	1.132	1.138			2.920	1.847	1.487	2.941	1.469	0.492		
EURFRAC	-0.702	-0.940	-1.287	-3.694	2.010	0.797			-4.480	-2.036	-3.988	-6.216
REGCAP	0.938	0.659			1.486	0.334			-4.249	-0.841	-4.224	-3.847
REGMENA	0.798	0.600			5.417	1.011	3.353	7.548	-0.477	-0.104		
REGSA	0.311	0.227			2.864	0.457			-4.670	-0.853	-3.000	-4.524
REGSSA	-0.275	-0.178			1.664	0.309			-1.473	-0.193		
REGLAC	0.029	0.022			0.275	0.079			3.615	1.283	2.435	3.317
REGWENA	-0.388	-0.262							-9.044	-1.455	-7.789	-5.614
LANDLOCK	-0.266	-0.325			1.098	0.772			2.585	1.188	2.562	3.979
LATITUDE	-0.021	-0.805	-0.041	-2.948	-0.046	-1.241			0.071	0.644		
AREA	-0.421	-2.383	-0.417	-4.188	-0.246	-0.725			-0.260	-0.564	-0.416	-4.095
MINDIST	-0.036	-0.258			-0.272	-0.183			-0.232	-0.508		
POP100CR	-0.010	-0.824			0.010	0.726			0.015	0.471	0.015	2.592
RESPOINT	-0.269	-0.500			-0.774	-0.916	3.353	7.548	-3.148	-1.723	-2.960	-4.633
TOPEN	0.608	0.712			2.046	1.220	1.990	5.192	3.184	1.237	1.854	3.512
CTRADE	0.013	0.640			0.036	1.495			0.046	0.487	0.034	2.421
EXPMANU	-0.097	-0.198			-1.975	-0.349			0.848	0.958	1.246	4.264
EXPPRIM	-0.378	-0.792			0.362	0.478			1.844	1.164	1.416	2.597
SDGR	-0.026	-0.203			-0.126	-0.636			-0.794	-1.154	-0.571	-3.875
SDBMP	0.000	-0.631			-0.001	-1.315			0.004	1.485	0.004	4.876
SDPI	0.001	1.273			0.001	1.053	0.001	2.193	-0.031	-1.513	-0.035	-5.076
SDTP	0.120	0.232			-0.435	-0.519			-2.270	-0.825	-1.390	-2.263
SDIT	-0.008	-0.805	-0.024	-3.270	-0.015	-0.720			0.033	0.827	0.036	3.222
CIVLEG	-2.353	-1.307	-0.927	-3.396	-6.899	-1.458	-4.562	-5.239	10.563	0.332		
COMLEG	-1.885	-1.037			-5.840	-1.083	-3.445	-3.879	12.554	0.390	1.687	4.452
POLITY2	0.051	0.927			0.151	1.528	0.164	4.382	0.083	0.326		
DURABLE	0.014	0.903			-0.025	-0.781			0.010	0.248		
FREE	0.001	0.004			0.168	0.596			-0.542	-0.424	-0.453	-3.033
KKZ	1.191	1.974	1.846	5.064	0.099	0.119			3.425	1.245	4.849	9.156
PCI	-4.827	-2.095	-5.363	-5.184	-6.769	-1.870	-5.391	-4.026	-13.777	-1.472	-13.547	-7.002
EURO1900					3.579	0.633						
MEDSHARE									0.627	0.152		
SRIGHT									-0.282	-0.320		
CRIGHT									-0.413	-1.098	-0.374	-3.169
RSS		26.44		51.11		5.37		17.06		1.03		3.78
sigma		0.97		0.97		0.82		0.73		1.02		0.50
R^2		0.90		0.80		0.96		0.86		0.99		0.98
Radj^2		0.77		0.77		0.77		0.82		0.77		0.94
LogLik		28.28		7.20		46.26		20.84		73.12		47.21
AIC		0.24		0.09		-0.47		-0.40		-1.71		-1.11
HQ		0.72		0.22		0.08		-0.22		-1.11		-0.73
SC		1.46		0.42		0.99		0.08		-0.06		-0.05
Chow test 1	0.00	0.00			0.00	0.00			0.00	0.00		
Chow test 2	1.34	0.28	0.66	0.68	7.45	0.39	1.35	0.28	0.00	0.00		
Normality tes	0.17	0.92	8.46	0.01	1.77	0.41	1.76	0.41	2.59	0.27	1.35	0.51
Hetero test												

Note: The dependent variable FD is the index of overall financial development over period, 1990-99. The variable description is in Appendix Table 1. The PcGets analysis yields coefficient and t-value for the variables in the final model. There are 64 observations in the whole sample, 44 observations in the developing country sample and 40 observations in the La Porta sample.

Appendix Table 8. Determinants of FDBOND

Sample Method Variable	La Porta's Dataset			
	BMA		PcGets	
	PIPs	Sign	Coeff	t-value
CONSTANT	1.000	(-)	-6.273	-4.754
GDP90	0.678	(+)	0.648	4.481
POP90	0.007			
ETHPOL				
ETHNIC				
RELIGION				
LANGUAGE	0.634	(+)	0.855	3.939
EURFRAC	0.127			
REGEAP	0.665	(-)	-0.529	-4.297
REGMENA				
REGSA	0.024			
REGSSA	0.245	(+)		
REGLAC				
REGWENA	0.122			
LANDLOCK	0.012			
LATITUDE				
AREA				
MINDIST	0.424	(-)		
POP100CR	0.663	(+)	0.011	3.820
RESPOINT	0.087			
TOPEN	0.236	(+)		
CTRADE				
EXPMANU	0.013			
EXPPRIM	0.000			
SDGR	0.062			
SDBMP	0.000			
SDPI	0.026			
SDTP	0.022			
SDTT	0.628	(+)	0.018	2.906
CIVLEG				
COMLEG	0.012			
POLITY2	0.032			
DURABLE	0.108			
FREE	0.019			
KKZ	0.620	(-)	-0.345	-2.642
PCI	0.781	(+)		
EURO1900				
MEDSHARE	0.022			
SRIGHT	0.135			
CRIGHT	0.161			
RSS				1.76
sigma				0.25
R^2				0.69
Radj^2				0.62
LogLik				52.37
AIC				-2.59
HQ				-2.48
SC				-2.28
Chow test 1				
Chow test 2			1.49	0.24
Normality test			3.79	0.15
Hetero test				

Note: The dependent variable FDBOND is the index of bond market development over period, 1990-99. The variable description is in Appendix Table 1. This study is based on La Porta dataset with 35 countries. The BMA analysis yields posterior probabilities of inclusion (PIPs), the total posterior model probabilities for all models including a given variable, and the sign certainty index of a relationship (Sign). No sign given means the sign of estimated relationship being uncertain each variable. The PcGets analysis yields coefficient and t-value for the variables in the final model.