# A New International Database on Financial Fragility 

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ERSA working paper 534

## August 2015

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July 28, 2015


#### Abstract

We present a new database on financial fragility for 124 countries over 1998 to 2012. In addition to commercial banks, our database incorporates investment banks and real estate and mortgage banks, which are thought to have played a central role in the recent financial crisis. Furthermore, it also includes cooperative banks, savings banks and Islamic banks, that are often thought to have different risk appetites than do commercial banks. As a result, the total value of financial assets in our database is around $50 \%$ higher than that accounted for by commercial banks alone. We provide eight different measures of financial fragility, each focussing on a different aspect of vulnerability in the financial system. Alternative selection rules for our variables distinguish between institutions with different levels of reporting frequency.


## 1 Introduction

In both research and policy making, there is a pressing need for a comprehensive international database that identifies the characteristics of financial systems that are vulnerable to crises. The analysis of such a database would enhance our understanding of the principal mechanisms through which crises are initiated and propagated. Existing financial sector datasets (e.g. Beck et al., 2000; Cihak et al., 2013) focus on the commercial banking sector,

[^0]but recent financial crises have highlighted the pivotal role played by investment banks and real estate and mortgage banks. In our database, investment banks and real estate and mortgage banks each account for $9 \%$ and $5 \%$ respectively of total bank assets, while cooperative banks and savings banks together account for another $20 \%$. Moreover, we provide a greater range of measures of financial fragility than do the existing datasets. Our database uses bank-level data from the Bureau van Dijk's Bankscope in order to construct the country-level financial fragility variables. ${ }^{1}$ It incorporates all deposit-taking institutions and also investment banks, since the activities of investment banks are not always separate from those of commercial banks in all countries, and investment banking activities are known to have played a major role in the most recent financial crisis.

The paper is organised as follows: section 2 presents the data preliminaries, section 3 discusses the selection rules, section 4 provides the aggregation methodology, section 5 analyses the newly constructed data, and section 6 concludes.

## 2 The Bank-Level Data

### 2.1 Data definitions and summary statistics

Bankscope covers 18 different types of financial institution. ${ }^{2}$ Our countrylevel data are constructed using information about six of these types, accounting for 23,287 out of the 29,366 institutions listed. There are five types of deposit-taking institution - commercial banks, co-operative banks, Islamic banks, real estate and mortgage banks, ${ }^{3}$ and savings banks - plus investment banks. The annual data span 1998-2012 inclusive. Table 1 shows the number of banks of each type and the proportion of assets accounted for by each type. It can be seen that commercial banks account for around two thirds of banking assets in the database; we show in section 5.5 that neglecting other types of banks can lead to mis-measurement of the overall level of financial fragility in the economy. Moreover, taking on board additional bank types can allow for a more in-depth analysis of the sources of financial fragility,

[^1]including examining the role played by different types of banks. ${ }^{4}$ For example, we show in section 5.5 that in the Netherlands and Germany, where cooperative and savings banks account for a substantial proportion of banking assets, the overall level of financial fragility is lower when these types of banks are included. ${ }^{5}$

Table 2 shows data based on the eight pre-defined activity levels in

Table 1: Frequency and asset shares of different banking institutions

| Specialisation | Frequency | Percent | Asset Share |
| :--- | ---: | ---: | ---: |
| Commercial Banks | 15,574 | 66.88 | 66.68 |
| Cooperative Banks | 3,507 | 15.06 | 10.80 |
| Investment Banks | 1,012 | 4.35 | 8.61 |
| Islamic Banks | 67 | 0.29 | 0.10 |
| Real Estate and Mortgage Banks | 399 | 1.71 | 4.98 |
| Savings Banks | 2,728 | 11.71 | 8.83 |
| Total | $\mathbf{2 3 , 2 8 7}$ | $\mathbf{1 0 0}$ | $\mathbf{1 0 0}$ |

Bankscope. Banks are grouped into two broad categories: 'active' and 'inactive'. Active banks include both those continuing to report to Bankscope and those known to be active but no longer reporting; some active banks are in receivership. A bank is classed as inactive if it has become bankrupt or has been liquidated, dissolved, or dissolved and merged. However, some banks have become inactive for no specified reason. Each year, different banks become active or cease to be active, so the panel of active banks is unbalanced. In any given year, approximately two thirds of the banks in the dataset are active and operating, and therefore in a position to report financial data.

One challenge in interpreting the data is deciding how to treat banks which are coded as active but report little or no accounting information. We do not know whether these banks are in financial distress, or whether there has been a change in the nature of their activity. It is possible that such banks ceased to report in an attempt to hide their financial problems. This implies a potential bias in the data for which there is no straightforward remedy. More broadly, banks are under no compulsion to report to Bankscope, so those which do report may not constitute a representative sample. For this reason we will present alternative country-level measures based on different rules for selecting individual banks into our sample, which are discussed below. We recommend that researchers employing our data use a range of alternative measures in order to check the robustness of their results.

[^2]Table 2: Bank status

| Status and Code | Bank <br> Frequency | Bank <br> Percent |
| :--- | ---: | ---: |
| 1: Active | 15,312 | 65.76 |
| 2: Active (receivership) | 487 | 2.09 |
| 3: Active (left Bankscope) | 685 | 2.94 |
| 4: Bankrupt | 38 | 0.16 |
| 5: Dissolved | 1,041 | 4.47 |
| 6: Dissolved and merged | 5,386 | 23.13 |
| 7: Liquidation | 254 | 1.09 |
| 8: No reason provided for inactivity | 84 | 0.36 |
| Total | $\mathbf{2 3 , 2 8 7}$ | $\mathbf{1 0 0}$ |

Table 3 provides a list of all 36 financial indicators available in Bankscope, along with their summary statistics. ${ }^{6}$ The variables highlighted in bold are used to construct the country-level data. The motivation for selecting these variables is provided later in this section; detailed definitions of the variables appear in Table 4 and in the Appendix. All of the ratios in Table 3 are calculated using standard Bankscope definitions.

Table 3 shows that even the most frequently reported variable (the return on annual average assets, ROAA) has only 127,365 observations across all institutions and years. This is far less than the theoretical maximum of 349,305 , highlighting the unbalanced nature of the panel. Other variables have far fewer observations: for example, there are only 27,473 observations of net charge-offs divided by average gross loans. Overall, however, most of the variables have over 100,000 observations, and there are only seven with fewer than 50,000.

The seven bank-level variables of financial fragility (that are used to construct the country-level data) are as follows: equity divided by total assets, impaired loans divided by total gross loans, the cost to income ratio, returns on average assets, net loans divided by total assets, liquid assets divided by total assets, and net charge-offs ${ }^{7}$ divided by average gross loans.

[^3]Table 3: Summary statistics of the financial variables downloaded

| Variable | Observations | Mean | Standard Deviation | Minimum | Maximum |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Total Assets (\$ US millions) | 120,285 | 10,400 | 84,400 | 0 | 3,070,000 |
| Deposits and Short Term Funding (\$ US millions) | 119,225 | 6,845 | 54,200 | 0 | 2,570,000 |
| Equity (\$ US millions) | 120,247 | 593 | 4,407 | -99,600 | 180,000 |
| Net Income (\$ US millions) | 119,526 | 45 | 978 | -50,100 | 261,000 |
| Loans (\$ US millions) | 100,248 | 5,838 | 39,400 | -185 | 1,840,000 |
| Gross Loans (\$ US millions) | 100,234 | 5,966 | 40,300 | 0 | 1,880,000 |
| Total Customer Deposits (\$ US millions) | 97,460 | 5,935 | 46,600 | 0 | 2,170,000 |
| Deposits from Banks (\$ US millions) | 80,229 | 1,769 | 14,300 | 0 | 718,000 |
| Total Liabilities and Equity (\$ US millions) | 102,275 | 12,200 | 91,400 | -6 | 3,070,000 |
| Profit Before Tax (\$ US millions) | 101,513 | 75 | 1,444 | -44,600 | 371,000 |
| Tax (\$ US millions) | 97,790 | 24 | 426 | -14,300 | 110,000 |
| Net Interest Revenue (\$ US millions) | 101,166 | 197 | 3,339 | -3,005 | 975,000 |
| Tier 1 Capital (\$ US millions) | 30,928 | 1,444 | 7,097 | -64,300 | 486,000 |
| Total Capital (\$ US millions) | 33,378 | 1,706 | 8,164 | -6,678 | 227,000 |
| Tier 1 Ratio (\%) | 31,085 | 17.33 | 31.19 | -747.38 | 962.18 |
| Total Capital Ratio (\%) | 42,168 | 21.78 | 46.33 | -766.94 | 989.84 |
| Impaired Loans divided by Gross Loans (\%) | 49,504 | 6.06 | 10.17 | 0.00 | 312.55 |
| Loan Loss Reserves divided by Gross Loans (\%) | 64,007 | 4.65 | 10.00 | 0.00 | 833.33 |
| Loan Loss Reserves divided by Impaired Loans (\%) | 46,375 | 130.98 | 162.11 | 0.00 | 998.88 |
| Impaired Loans divided by Equity (\%) | 50,088 | 44.84 | 77.28 | 0.00 | 993.42 |
| Net Charge Offs divided by Average Gross Loans (\%) | 27,473 | 0.93 | 7.61 | -251.43 | 811.32 |
| Equity divided by Total Assets (\%) | 102,194 | 11.65 | 19.63 | -992.86 | 100.00 |
| Equity divided by Liabilities (\%) | 101,575 | 19.00 | 53.77 | -275.00 | 999.85 |
| Capital Funds divided by Total Assets (\%) | 64,047 | 10.71 | 16.34 | -898.88 | 100.00 |
| Capital Funds divided by Net Loans (\%) | 62,180 | 27.21 | 66.56 | -856.25 | 994.03 |
| Capital Funds divided by Liabilities (\%) | 63,839 | 16.37 | 47.54 | -107.14 | 992.31 |
| Net Interest Margin (\%) | 126,899 | 3.90 | 10.76 | -987.50 | 966.67 |
| Return on Average Assets (\%) | 127,365 | 0.69 | 4.95 | -540.48 | 185.57 |
| Return on Average Equity (\%) | 127,265 | 6.80 | 27.51 | -998.29 | 997.05 |
| Cost to Income Ratio (\%) | 125,730 | 70.77 | 38.47 | 0.00 | 988.89 |
| Net Loans divided by Total Assets (\%) | 126,061 | 57.92 | 20.66 | 0.00 | 100.00 |
| Net Loans divided by Deposits and Short Term Funding (\%) | 124,514 | 80.60 | 58.22 | -150.46 | 999.36 |
| Net Loans divided by Total Deposits and Borrowing (\%) | 115,138 | 70.40 | 35.07 | -135.52 | 957.58 |
| Liquid Assets divided by Deposits and Short Term Funding (\%) | 125,794 | 31.30 | 58.02 | -4.61 | 999.41 |
| Liquid Assets divided by Total Deposits and Borrowing (\%) | 116,058 | 25.11 | 43.85 | -3.77 | 997.63 |
| Liquid Assets divided by Total Assets (\%) | 100,162 | 21.36 | 18.67 | 0.00 | 100.00 |

Table 4: Core measures related to financial fragility

| CAMEL Measure | Variable | Bankscope Code | Definition of Variable | Proprtional/Inverse |
| :---: | :---: | :---: | :---: | :---: |
| Capitalisation | $\frac{\text { Equity }}{\text { Total Assets }}$ | $\frac{2055}{2060}$ | $\frac{\text { Equity }}{\text { Total Liabilities + Equity }}$ | (-) |
| Asset Quality | $\frac{\text { Impaired Loans }}{\text { Gross Loans }}$ | $\frac{2170}{2000+2070}$ | mpiared Loans <br> Loans + Loan Loss Reserves | (+) |
| Managerial Efficiency | $\frac{\text { Cost }}{\text { Income }}$ | $\frac{2090}{2080+2085}$ | Overhead Cost <br> $\overline{\text { Net Interest Revenue + Other Operating Income }}$ | (+) |
| Earnings | $\frac{\text { Net Income }}{\text { Average Total Assets }}$ | $\frac{2115}{\text { Average } 2025}$ | $\frac{\text { Net Income }}{\text { Total Assets }}$ | (-) |
| Liquidity I | $\frac{\text { Net Loans }}{\text { Total Assets }}$ | $\frac{2000}{2025}$ | $\frac{\text { Loans }}{\text { Total Assets }}$ | (+) |
| Liquidity II | $\frac{\text { Liquid Assets }}{\text { Total Assets }}$ | $\frac{2075}{2025}$ | $\frac{\text { Liquid Assets }}{\text { Total Assets }}$ | (-) |
| Risk exposure | $\frac{\text { Net Charge Offs }}{\text { Average Gross Loans }}$ | $\frac{2150}{2000+2070}$ | $\frac{\text { Net Charge Offs }}{\text { Loans + Loan Loss Reserves }}$ | (+) |
| Notes: The variables in the definition section in the final column (hence, the numbers in the third column) may be further disaggregated. The disaggregated values may be found in the Appendix so the reader may find what the composition of say " 2000 (net loans)" truly is. |  |  |  |  |

These variables were chosen to reflect the key areas of the CAMELS bank rating system (capitalisation, asset quality, managerial efficiency, earnings, liquidity, and sensitivity to $\mathrm{risk}^{8}$ ).

Using these seven bank-level variables we construct eight country-level measures of financial fragility: bank capitalisation, asset quality, managerial efficiency, the return on average assets, two alternative measures of liquidity, a measure of risk exposure, and a general financial stability measure (a Z-score) ${ }^{9}$. These measures are summarised in Table 4; more detail on the definition of the variables appears in the Appendix. The first five measures below are our "core" measures.

We measure bank capitalisation (102,194 observations) as the ratio of equity to total assets. The mean of this ratio is $11.7 \%$ and the median is $7.9 \%$; at the 99 th percentile the ratio is $82 \%$. 33 banks have a ratio of $100 \%$, and 275 banks with a negative ratio for at least one year.

Asset quality (49,504 observations) is measured as impaired loans divided by gross loans. The mean of this ratio is $6.1 \%$ and the median is $3.0 \%$;
of fragility. Net charge-offs should be inversely related to the quality of loan screening and so positively related to the degree of fragility. In a case where a bank reclaims some of the bad loans at a later date, net charge offs will be negative, indicating a reduction in fragility.
${ }^{8}$ Although we do not compute a direct measure of sensitivity to risk, net charge-offs divided by average gross loans can be viewed as a proxy for sensitivity.
${ }^{9}$ The Z-score is the only country-level indicator that does not have a corresponding indicator at the bank-level, because its construction utilises the variability of returns across banks (see Section 4.2.).
the ratio exceeds $20 \%$ only at the 95 th percentile. 438 institutions report a value of $0 \%$, but the maximum value exceeds $100 \%$ : as Table 4 shows, the ratio includes loan loss reserves in the denominator, and when such reserves are negative the ratio may exceed $100 \%$.

Managerial efficiency (125,730 observations) is measured as the cost-to-income ratio. A management which deploys its resources efficiently will look to maximise its income and reduce its operating costs, so a larger ratio implies a lower level of efficiency. The mean of this ratio is $71 \%$ and the median is $68 \%$. The ratio does not exceed 100 until approximately the 95th percentile. There are 73 observations (from 39 banks) with a figure of zero. The ratio exceeds $100 \%$ for 3,318 banks and 6,450 bank-year observations. It exceeds $500 \%$ for 180 banks, and the largest value is $989 \%$.

The return on average assets ( 127,365 observations) is used to measure an institution's earnings capacity. An institution has to make an appropriate return on assets to replenish or increase capital, fund expansion from retained earnings, or to generate profit that will be paid out as dividends. The mean return is $0.7 \%$ and the median is $0.5 \%$. At the 10th percentile the value is $0 \%$, and at the 90 th percentile $2.2 \%$. Only 36 banks have a value exceeding $50 \%$; the maximum value is $186 \%$. There are 5,345 banks $(12,144$ bank-year observations) with negative returns, and 72 banks ( 98 observations) with returns below $-50 \%$.

Our first measure of liquidity is net loans divided by total assets (126,061 observations), which is inversely related to liquidity. The mean of this variable is $57.9 \%$ and the median is $61.2 \%$. 307 banks ( 1,026 bank-year observations) report a value of $0 \%$, and two banks (five bank-year observations) report a value of $100 \%$. At the 10th percentile the value is $29 \%$ and at the 90 th percentile it is $81 \%$.

Our second liquidity measure is liquid assets divided by total assets (100,162 observations). This variable has a mean of $21.4 \%$ and a median of $15.7 \%$. The value at the 5 th percentile is $2.5 \%$, while the value at the 95 th percentile is $62.1 \%$. In 6 different banks in 6 different countries (Malaysia, the Netherlands, Russia, the United Kingdom, the United States and Uruguay) the ratio is $100 \%$.

Our measure of risk exposure is net charge-offs as a fraction of total loans (27,473 observations). This variable has a mean of $0.9 \%$ and a median of $2.4 \%$. The variable can be negative when banks recover debt that was originally written off, and this is a common occurrence in the data. At the 10 th percentile the value is $-0.08 \%$, while at the 90 th percentile its value is $2.4 \%$. There are 16 cases (five banks in 12 countries) in which banks are charging off over $100 \%$ of their average loans.

A final indicator of financial fragility is the $\boldsymbol{Z}$-score. This variable is
not taken directly from Bankscope, but is constructed at the country level using Bankscope data. The higher the Z-score, the more financially sound a country is. The construction of the Z-score is discussed in more detail below.

The construction of national aggregate data also makes use of the total annual average asset value of each bank, as a measure of the relative size of each institution. ${ }^{10}$ The distribution of this variable is highly skewed: its mean value is USD 10.4 billion and its median is USD 454 million. The minimum is zero and the maximum is USD 3.1 trillion. At the 10th percentile the value is USD 45 million while at the 90th percentile the value is USD 8.9 billion.

### 2.2 The geographical distribution of banks

The database covers 124 countries. The number of institutions varies substantially across countries, and Table 5 notes the 25 countries with the largest and smallest number of banks in the database. Overall, there is a strong positive correlation between the number of banks and a country's level of economic development, and Africa accounts for only 796 out of the 23,287 banks in the database. ${ }^{11}$ Some of the African countries are very small both in terms of GDP and in terms of population, and have a formal financial sector which is very rudimentary. Moreover, three of the African countries, South Africa, Nigeria and Kenya, account for over $25 \%$ of all banks from the continent. Many of the African banks report data for only a handful of years.

The infrequency of data from banks in developing countries (and particularly from African banks) creates potential problems in the construction of our dataset. In countries where bank penetration is low, country-level data may be driven by a very small number of banks. (Moreover, Bankscope does not include data from every bank in a country, and its selection may not be representative of the population of banks.) For this reason, the dataset of Beck et al. (2000) is based on a rule that in any one year excludes countries with fewer than three banks. However, the application of such a rule to our database would mean that four countries would be excluded entirely. This trade-off is discussed in more detail below.

[^4]Table 5: Countries with the largest and smallest number of banks

| Bank |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- |
| Rank | Country | Name | Frequency | Bank <br> Rank | Country <br> Name |
| 1 | United States of America | 10,671 | 1 | Guinea Bissau | Frequency |
| 2 | Germany | 2,605 | 2 | Central African Republic | 1 |
| 3 | Russia | 1,186 | 3 | Djibouti | 2 |
| 4 | Japan | 929 | 4 | Equatorial Guinea | 2 |
| 5 | Italy | 928 | 5 | Eritrea | 3 |
| 6 | Switzerland | 545 | 6 | Sao Tome and Principe | 3 |
| 7 | France | 512 | 7 | Cape Verde | 4 |
| 8 | United Kingdom | 377 | 8 | Congo | 4 |
| 9 | Austria | 369 | 9 | Lesotho | 4 |
| 10 | Spain | 269 | 10 | Chad | 5 |
| 11 | Brazil | 233 | 11 | Guinea | 5 |
| 12 | Ukraine | 191 | 12 | Liberia | 5 |
| 13 | China | 187 | 13 | Seychelles | 5 |
| 14 | Norway | 170 | 14 | Swaziland | 5 |
| 15 | Denmark | 149 | 15 | Gabon | 6 |
| 16 | Argentina | 134 | 16 | Madagascar | 6 |
| 17 | Indonesia | 132 | 17 | Burundi | 7 |
| 18 | Sweden | 130 | 18 | Niger | 7 |
| 19 | Canada | 124 | 19 | Togo | 7 |
| 20 | Belgium | 114 | 20 | Namibia | 8 |
| 21 | India | 106 | 21 | Rwanda | 8 |
| 22 | Malaysia | 105 | 22 | Benin | 9 |
| 23 | Australia | 102 | 23 | Gambia | 9 |
| 24 | Hong Kong | Nigeria | 24 | Burkina Faso | 10 |
| 25 | 96 | 25 | Kyrgyzstan | 10 |  |

## 3 Selection Rules Used in Constructing National Aggregates

A bank's entry into or exit from the database might be correlated with changes in its level of fragility, which introduces potential biases in a national aggregate measure based on all available data. In this case, restricting the sample to banks reporting consistently through time is likely to reduce the bias in the measurement of changes in national aggregates. However, the restriction is likely to result in a sample that is less representative in terms of aggregate levels. For this reason, the five core aggregate measures of financial fragility (bank capitalisation, asset quality, managerial efficiency, ROAA, and net loans divided by total assets) are constructed in five different ways, each way involving a different selection rule. We recommend that empirical applications using our data include a comparison of results for the five alternative measures, as a robustness check.

One potential selection rule would be to use the largest five or 10 banks (or largest $10 \%$ of banks) in a country, but then any systematic correlation between bank size and fragility would lead to biases in aggregate measures. Moreover, the large disparities in the number of banks per country mean that this rule is unlikely to produce consistent measures across countries.

Note also that there are substantial variations in individual bank size over the sample period, which introduces additional complications in the application of a rule based on bank size. Instead, we implement the following five alternative selection rules.

The first selection rule is to include all available observations for individual banks. If a bank reports the value of a particular variable in a given year then this value is used in the construction of the national aggregate, regardless of the frequency with which that bank reports data. This rule generates the "base sample" indicated in the uppermost part of the flow chart in Figure 1.

The next two selection rules are based on the frequency of reporting all five core variables. ${ }^{12}$ One rule is based on the total number of years in which the five variables are reported: a bank is included when it reports all five variables simultaneously in at least eight years (not necessarily consecutive years). The other rule is based on the proportion of years for which the bank is known to exist, existence being indicated by the fact that the bank currently reports data, or that it has previously reported data and its name still appears in Bankscope. A bank is included when it reports all five variables simultaneously in at least $66 \%$ of the years for which it is known to exist. The $66 \%$ rule is less restrictive than the eight-year rule, as it entails the inclusion of banks that disappear early in the sample period.

The final two selection rules are variable-specific. For each of the five variables individually, the first of these rules includes a bank if it reports that variable in at least eight years. The other rule includes a bank if it reports that variable in at least $66 \%$ of the years for which it is known to exist.

Table 6 summarises all of the selection rules and Figure 1 illustrates their taxonomy. Letting Var stand for one of the five core variables, "Var" in Table 6 indicates a national aggregate constructed using the base sample rule, "VarR" indicates an aggregate based on the variable-specific eight-year rule, "VarR5" indicates an aggregate based on the eight-year rule applied to all five variables simultaneously, "VarH" indicates an aggregate based on the variable-specific $66 \%$ rule, and "VarH5" indicates an aggregate based on the $66 \%$ rule applied to all five variables simultaneously.

[^5]

For each of the five variables, Table 7 indicates the total number of observations entailed by the five different selection rules. For four of the five variables, the base sample corresponds to over 100,000 observations. However, for impaired loans there are slightly fewer than 50,000 observations in the base sample, and consequently the selection rules which require all five variables to be reported simultaneously (VarR5 and VarH5) entail fewer than 50,000 observations in all cases. By contrast, the variable-specific rules ( $\operatorname{Var} R$ and $\operatorname{Var} H$ ) entail relatively moderate reductions in the number of observations relative to the base sample. With the $66 \%$ variable-specific rule $(\operatorname{VarH})$, the reduction represents less than $5 \%$ of the base sample (except in the case of impaired loans, for which the reduction is around $20 \%$ ).

Table 6: Alternative selection rule criteria

| Variable | Is it required that <br> all five variables re- <br> ported simultane- <br> ously in a given <br> year? | Is it required that a <br> bank reports for at <br> least eight years? | Is required that <br> a bank reports at <br> least $\mathbf{6 6 \%}$ of its <br> time in the panel? |
| :--- | :--- | :--- | :--- |
|  | NO | NO |  |
| Var | NO | YES | NO |
| VarR | NO | YES | YES |
| VarH | YES | NO | NO |
| VarR5 | YES | YES |  |
| VarH5 | Notes: Var stands for the variable name. The figures following "Var" and their |  |  |
| definitions are available in the country code book in the Appendix. |  |  |  |

## 4 Construction of the Aggregate National Data

### 4.1 Cross-country variation in bank prevalence

Before discussing the construction of national aggregates, it is informative to examine the variation in the prevalence of banks across countries, as sum-

Table 7: The total number of observations based on selection rules

| Selection Rule | Equity | Impaired <br> Loans | Cost To <br> Income | Return On <br> Assets | Net Loans |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Var | 102195 | 49504 | 125730 | 127365 | 126061 |
| VarR5 | 32519 | 29519 | 32327 | 32498 | 32515 |
| VarH5 | 42491 | 39578 | 42244 | 42423 | 42463 |
| VarR | 76256 | 29830 | 98331 | 99925 | 99168 |
| VarH | 98290 | 39953 | 121284 | 123301 | 121896 |

Notes: Var stands for the variable name. The figures following "Var" and their definitions are available in the country code book in the appendix.
marised in Table 8. ${ }^{13}$ As noted above, there is a strong correlation between the number of banks and the overall level of economic development: in Africa, the number of banks that report data rarely exceeds 10 . But also, the number of banks recorded in Bankscope increases over time. For example, Austria has only 180 banks before 2002, but this number increases to well over 200 in subsequent years. Until 2004 there are only 58 Chinese banks, but by 2010 there are over 150. This trend is much stronger in some countries than others. Moreover, in some countries there is a sharp increase in the number of banks recorded in a particular year, as for example in Italy in 2004. The reasons for this are unknown. These variations should be borne in mind when using the national aggregate data.

### 4.2 Aggregation from the bank to the country level

Aggregate national data are constructed as weighted averages, using weights based on individual banks' total asset values. Let $X_{i j t}$ stand for measure $X$ for bank $i$ in country $j$ in year $t$. Then the national aggregate is constructed as follows:

$$
\begin{equation*}
X_{j t}=\Sigma_{i \in j} W_{i j t} * X_{i j t} \tag{1}
\end{equation*}
$$

Here, $W_{i j t}$ is a weight constructed as follows:

$$
\begin{equation*}
W_{i j t}=\frac{A_{i j t}}{\sum_{i=1}^{i=N_{x j t}} A_{i j t}} \tag{2}
\end{equation*}
$$

where $A_{i j t}$ is the value of bank $i$ 's assets in country $j$ in year $t$. Note that the number of banks $\left(N_{x j t}\right)$ can vary across countries, across time, and also across variables. It also varies according to the selection rule used.

[^6]Our dataset also includes a Z-score measure. The Z-score is expected to be inversely related to financial fragility; it is constructed as follows:

$$
\begin{equation*}
Z_{j t}=\frac{\operatorname{ROAA}_{j t}+\frac{\operatorname{equity~}_{j t}}{\operatorname{assets}_{j t}}}{\sigma_{\mathrm{ROAA}_{j}}} \tag{3}
\end{equation*}
$$

Here, $\sigma_{\mathrm{ROAA}_{j}}$ is a country-specific standard deviation of the national average value of ROAA $\left(R O A A_{j t}\right)$ over time. ${ }^{14}$

This approach is similar to that of Cihak and Hesse (2007) and differs from that of Cihak et al. (2013). Cihak et al. (2013) construct the standard deviation using five-year moving averages, which entails the loss of data for the first four years of the sample.

## 5 Summary Statistics

### 5.1 Summary statistics across all countries under different selection rules

Tables 9-13 present summary statistics for the national aggregate measure of the five core variables, with one table for each variable. Although stricter selection rules entail some reduction in sample size, with some country-year observations disappearing from the sample, the total number of observations is always well above 1,000 (out of a theoretical maximum of 1,860 from 124 countries over 15 years).

Table 14 shows the country level data for the three additional variables: liquid assets divided by total assets, net charge-offs divided by average gross loans, and the Z-score.

Tables 9-14 show a large degree of variability in the fragility measures as most standard deviations are large in comparison to their corresponding means. Note the substantial variation in national aggregate liquidity. Most of the country-year observations in the upper end of this distribution are in Africa, which is consistent with the finding of Demetriades and James (2011) that African banks are typically unable to extend credit to individuals and firms due to the dysfunctional nature of African credit markets.

Overall, the tables do not show a great deal of variation in summary

[^7]Table 8: Equity divided by total assets

| Variable | Obs | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Equity | 1782 | 9.76 | 6.19 | -41.58 | 85.37 |
| EquityR5 | 1338 | 8.82 | 4.78 | -52.04 | 26.86 |
| EquityH5 | 1397 | 9.16 | 5.88 | -45.27 | 97.52 |
| EquityR | 1669 | 9.79 | 6.74 | -44.57 | 74.76 |
| EquityH | 1744 | 9.80 | 5.99 | -42.47 | 85.37 |

Table 9: Impaired loans divided by gross loans

| Variable | Obs | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| ImpLoans | 1493 | 7.52 | 8.43 | 0.03 | 103.29 |
| ImpLoansR5 | 1262 | 6.84 | 7.05 | 0.02 | 63.52 |
| ImpLoansH5 | 1342 | 6.84 | 7.51 | 0.04 | 91.70 |
| ImpLoansR | 1281 | 6.98 | 7.22 | 0.02 | 63.52 |
| ImpLoansH | 1364 | 6.95 | 7.57 | 0.04 | 91.70 |

Table 10: The cost to income ratio

| Variable | Obs | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Costs | 1764 | 60.72 | 21.47 | 3.81 | 382.17 |
| CostsR5 | 1332 | 59.65 | 19.77 | 7.31 | 240.18 |
| CostsH5 | 1391 | 60.26 | 18.91 | 11.32 | 267.35 |
| CostsR | 1630 | 60.25 | 22.01 | 1.94 | 374.52 |
| CostsH | 1716 | 60.69 | 21.45 | 6.80 | 382.17 |

Table 11: Returns on average assets

| Variable | Obs | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Returns | 1779 | 1.34 | 2.57 | -47.43 | 21.79 |
| ReturnsR5 | 1334 | 1.30 | 2.87 | -51.59 | 12.28 |
| ReturnsH5 | 1391 | 1.26 | 2.67 | -50.60 | 8.64 |
| ReturnsR | 1666 | 1.36 | 2.77 | -50.22 | 12.47 |
| ReturnH | 1743 | 1.35 | 2.75 | -48.16 | 45.92 |

Table 12: Net loans divided by total assets

| Variable | Obs | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| NetLoans | 1781 | 49.65 | 15.11 | 2.36 | 92.40 |
| NetLoansR5 | 1338 | 52.49 | 14.29 | 0.50 | 96.52 |
| NetLoansH5 | 1397 | 52.82 | 13.89 | 0.01 | 94.81 |
| NetLoansR | 1664 | 49.60 | 15.49 | 0.75 | 96.50 |
| NetLoansH | 1743 | 49.95 | 15.04 | 0.01 | 92.40 |

Table 13: Additional fragility measures

| Variable | Obs | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Liquid Assets | 1781 | 28.38 | 15.54 | 0.49 | 96.39 |
| Net Charge Offs | 1212 | 1.07 | 2.64 | -16.36 | 31.48 |
| Z-Score | 1779 | 14.97 | 11.13 | -14.33 | 94.16 |


| 98 | 98 | 67 | 67 | 67 | g | g | g | 9 | 4 | 4 | 8 | 4 | 8 | 8 | әए！ЧО |
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| $\zeta$ | $\zeta$ | $\checkmark$ | $\zeta$ | $\checkmark$ | $\zeta$ | $\zeta$ | $\zeta$ | $\checkmark$ | $\zeta$ | $\zeta$ | $\zeta$ | $\zeta$ | I | I |  |
| も | $\pm$ | モ | I | $\varepsilon$ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | әріәл әбел |
| 69 | 02 | 99 | 97 | $0 \varepsilon$ | ¢¢ | 理 | 理 | Z¢ | $0 \varepsilon$ | モ¢ | 67 | $\angle 7$ | 27 | I¢ | ереиел |
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| 9 | 9 | G | 9 | 8 | $\angle$ | 8 | $\llcorner$ | 8 | $L$ | $L$ | 1 | $L$ | $L$ | ஏ | osey reutyng |
| LZ | LZ | L\％ | 6I | 6I | LI | 6I | 07 | LI | 91 | GI | も | \＆I | ¢L | ZI | ет．ıes［ng |
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| II | II | 0I | 0I | 0I | 6 | $\angle$ | $\angle$ | g | $\dagger$ | $\checkmark$ | $\checkmark$ | $\checkmark$ | $\checkmark$ | I | гиемsqоg |
| II | II | II | 8 | 8 | $L$ | 1 | 8 | 8 | 8 | 8 | 8 | 8 | 8 | 6 |  |
| $\varepsilon$ | $\varepsilon$ | ¢ | I | 9 | $\angle$ | 4 | 9 | 9 | 9 | 9 | 9 | Ø | † | も | u！uәg |
| 28 | 07 | It | ET | 87 | 87 | 67 | 67 | $\angle D$ | IG |  | IS | 09 | \％9 | ¢9 | untiopg |
| $\checkmark 7$ | 7\％ | $7 \%$ | 8I | 9 I | も | ZI | II | II | 0I | 8 | 8 | II | 8 | ஏ | sniepeg |
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| 07 | LV | 8 I | 6I | 6I | LI |  | ¢I | ¢I | ¢I | II | 7I | II | 8 | $L$ | ue！！eqıəz\％ |
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| 89 | 89 | 69 | ¢9 | $\angle 9$ | 99 | 02 | 89 | 69 | 89 | IL | 92 | 92 | 92 | 62 |  |
| 9 I | ¢I | GI |  | ZI | II | II | 6 | 8 | 8 | 9 | 9 | \＆ | $\varepsilon$ | $\checkmark$ | ${ }_{\text {elosun }}$ |
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statistics according to the selection rule used. This suggests that in some applications of our data the results may not be that sensitive to the choice of selection rule. However, the figures for some low-income countries with small numbers of banks may be more sensitive. The next sub-section provides some illustrative examples of the effect of the choice of selection rule on the distribution of the variables over time for selected countries.

### 5.2 Summary statistics for selected countries

The results in this sub-section pertain to eight different countries in different parts of the world and at different levels of financial and economic development: Argentina (with around 70 banks), Indonesia (with around 60 banks), Ukraine (with around 30 banks for most years ${ }^{15}$ ), Libya (with 8 banks), Nigeria (with around 50 banks at the beginning of the sample but only around 20 at the end ${ }^{16}$ ), South Africa (with around 20 banks), the Netherlands (with around 30 banks), and the United States (with around 1,000 banks).

Table 15 shows summary statistics for Argentina, which endured a financial crisis early on in the sample period. Since Argentina has a relatively large number of banks, all selection rules entail a full 15 observations for each variable. The choice of selection rule does not have a large impact on the sample mean. Note, however, that in the case of ROAA the less restrictive rules entail a negative mean and the more restrictive rules entail a positive mean. This is because the less restrictive rules lead to the inclusion of a relatively large number of bank-year observations from the crisis period. With the ROAA measure, therefore, stricter selection rules create an impression of less fragility. Nevertheless, this difference in mean values across selection rules is small relative to the standard deviations.

Table 16 shows that in Indonesia, which also has a relatively large number of banks and also endured a financial crisis early on in the sample period, the choice of selection rule also makes little difference to the mean values of the variables. The mean impaired loan ratio is slightly lower under stricter rules, again creating an impression of less fragility, but again this difference is small relative to the standard deviations.

In Libya (Table 17), there are very few banks. Even under the least restrictive rule, there are no observations for impaired loans, and for the other variables the choice of selection rule makes a large difference to the number

[^8]Table 15: Summary statistics for Argentina

| Variable | Obs. | Mean | S. Dev. | Minimum | Maximum |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |
| Equity | 15 | 10.41 | 1.44 | 7.17 | 13.32 |
| EquityR5 | 15 | 10.46 | 1.21 | 8.46 | 12.83 |
| EquityH5 | 15 | 11.33 | 1.33 | 8.88 | 13.90 |
| EquityR | 15 | 10.67 | 1.31 | 8.29 | 13.34 |
| EquityH | 15 | 10.39 | 1.47 | 7.09 | 13.33 |
|  |  |  |  |  |  |
| ImpLoans | 15 | 7.25 | 6.04 | 1.13 | 20.75 |
| ImpLoansR5 | 15 | 6.77 | 5.45 | 1.09 | 18.00 |
| ImpLoansH5 | 15 | 6.75 | 5.72 | 1.11 | 20.92 |
| ImpLoansR | 15 | 6.80 | 5.54 | 1.09 | 18.35 |
| ImpLoansH | 15 | 7.20 | 5.85 | 1.14 | 19.07 |
|  |  |  |  |  |  |
| Costs | 15 | 75.52 | 23.76 | 56.89 | 142.64 |
| CostsR5 | 15 | 73.42 | 24.48 | 54.83 | 151.10 |
| CostsH5 | 15 | 71.90 | 28.59 | 52.27 | 161.83 |
| CostsR | 15 | 75.53 | 25.06 | 56.18 | 147.54 |
| CostsH | 15 | 75.39 | 23.74 | 56.89 | 142.57 |
|  |  |  |  |  |  |
| Returns | 15 | -0.06 | 4.15 | -14.12 | 2.64 |
| ReturnsR5 | 15 | 0.28 | 2.87 | -8.59 | 2.69 |
| ReturnsH5 | 15 | 0.30 | 3.35 | -10.79 | 2.85 |
| ReturnsR | 15 | 0.23 | 3.02 | -9.29 | 2.63 |
| ReturnsH | 15 | -0.07 | 4.16 | -14.16 | 2.64 |
|  |  |  |  |  |  |
| NetLoans | 15 | 43.39 | 7.55 | 30.82 | 56.24 |
| NetLoansR5 | 15 | 44.46 | 8.18 | 31.18 | 59.33 |
| NetLoansH5 | 15 | 44.38 | 6.82 | 32.47 | 57.46 |
| NetLoansR | 15 | 43.21 | 7.49 | 30.61 | 56.19 |
| NetLoansH | 15 | 43.54 | 7.62 | 30.82 | 56.30 |
|  |  |  |  |  |  |

Table 16: Summary statistics for Indonesia

| Variable | Obs. | Mean | S. Dev. | Minimum | Maximum |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  | 15 | 5.05 | 13.57 | -41.58 | 11.85 |
| Equity | 15 | 4.69 | 14.76 | -46.18 | 11.94 |
| EquityR5 | 15 | 4.76 | 14.52 | -45.27 | 11.77 |
| EquityH5 | 15 | 4.90 | 14.37 | -44.57 | 12.02 |
| EquityR | 15 | 4.01 | 13.81 | -42.47 | 11.85 |
| EquityH | 15 | 5.01 |  |  |  |
|  |  | 10.08 | 11.95 | 2.12 | 47.84 |
| ImpLoans | 15 | 15 | 2.68 | 11.09 | 2.06 |
| ImpLoansR5 | 15 | 9.93 |  |  |  |
| ImpLoansH5 | 15 | 9.63 | 10.95 | 2.11 | 44.05 |
| ImpLoansR | 15 | 9.68 | 11.09 | 2.06 | 44.93 |
| ImpLoansH | 15 | 9.78 | 11.23 | 2.13 | 45.29 |
|  |  |  |  |  |  |
| Costs | 15 | 58.71 | 9.88 | 49.30 | 78.65 |
| CostsR5 | 15 | 55.45 | 10.57 | 33.79 | 78.78 |
| CostsH5 | 15 | 54.41 | 10.25 | 31.84 | 78.59 |
| CostsR | 15 | 58.48 | 9.78 | 48.72 | 78.86 |
| CostsH | 15 | 55.75 | 11.27 | 34.04 | 78.61 |
|  |  |  |  |  |  |
| Returns | 15 | -2.74 | 13.14 | -47.43 | 2.46 |
| ReturnsR5 | 15 | -3.09 | 14.21 | -51.59 | 2.51 |
| ReturnsH5 | 15 | -2.99 | 13.97 | -50.60 | 2.46 |
| ReturnsR | 15 | -2.98 | 13.86 | -50.22 | 2.51 |
| ReturnsH | 15 | -2.82 | 13.32 | -48.16 | 2.46 |
| NetLoans | 15 | 43.36 | 14.72 | 19.90 | 63.09 |
| NetLoansR5 | 15 | 43.00 | 15.08 | 18.70 | 63.11 |
| NetLoansH5 | 15 | 43.24 | 14.99 | 18.97 | 62.97 |
| NetLoansR | 15 | 43.24 | 15.00 | 19.02 | 63.24 |
| NetLoansH | 15 | 43.33 | 14.80 | 19.62 | 63.09 |
|  |  |  |  |  |  |

Table 17: Summary statistics for Libya

| Variable | Obs. | Mean | S. Dev. | Minimum | Maximum |
| :--- | ---: | :---: | :---: | :---: | ---: |
|  | 15 | 10.95 | 6.49 | 5.28 | 27.01 |
| Equity | 15 |  |  |  |  |
| EquityR5 | 0 |  |  |  |  |
| EquityH5 | 0 |  |  |  |  |
| EquityR | 15 | 12.27 | 10.78 | 3.75 | 35.83 |
| EquityH | 15 | 11.59 | 7.11 | 5.28 | 27.01 |
|  |  |  |  |  |  |
| ImpLoans | 0 |  |  |  |  |
| ImpLoansR5 | 0 |  |  |  |  |
| ImpLoansH5 | 0 |  |  |  |  |
| ImpLoansR | 0 |  |  |  |  |
| ImpLoansH | 0 |  |  |  |  |
|  |  |  |  |  |  |
| Costs | 15 | 49.16 | 26.07 | 18.49 | 95.12 |
| CostsR5 | 0 |  |  |  |  |
| CostsH5 | 0 |  |  |  |  |
| CostsR | 12 | 57.22 | 10.38 | 44.17 | 79.07 |
| CostsH | 15 | 48.52 | 31.06 | 15.37 | 110.10 |
|  |  |  |  |  |  |
| Returns | 15 | 0.48 | 0.39 | -0.61 | 0.96 |
| ReturnsR5 | 0 |  |  |  |  |
| ReturnsH5 | 0 |  |  |  |  |
| ReturnsR | 15 | 0.44 | 0.39 | -0.58 | 0.83 |
| ReturnsH | 15 | 0.52 | 0.40 | -0.61 | 0.96 |
|  |  |  |  |  |  |
| NetLoans | 15 | 27.56 | 13.03 | 12.79 | 47.86 |
| NetLoansR5 | 0 |  |  |  |  |
| NetLoansH5 | 0 |  |  | 14.29 | 49.78 |
| NetLoansR | 12 | 32.78 | 11.94 | 14.29 | 49.78 |
| NetLoansH | 15 | 24.45 | 16.64 | 0.64 |  |

of observations. Under stricter rules there are no observations for any variable, and in a cross-country analysis Libya would drop out of the sample. However, there is little variation in the sample means across those selection rules that do allow a positive number of observations.

Although the Netherlands (Table 18) is a high-income country, banks do not report consistently to Bankscope across all years, and stricter selection rules do reduce the number of annual observations. Moreover, for some of the measures (in particular bank capitalisation) the variation in mean across selection rules is somewhat larger than in Argentina and Indonesia, both in absolute terms and relative to the corresponding standard deviations. Moreover, unlike in Argentina and Indonesia, there is some variation in the standard deviations across selection rules.

Table 19 shows that in Nigeria, which has a relatively large number of banks, the choice of selection rule makes only a small difference to the total number of annual observations. However, for all five variables, the choice of selection rule makes a very substantial difference to either the mean or the standard deviation (or both). This reflects a substantial variation in the number of banks reporting data in any one year. The total number of banks reporting declines from around 50 at the beginning of the sample ${ }^{17}$ to around 20 at the end; moreover, the total number of banks reporting data in at least one year is 94 , of which 66 cease to report at some point (see Tables V-VI in the Appendix). Whilst, regulatory reforms over the sample period may further explain the variation in bank numbers, ${ }^{18}$ it may well be the case that the decision about whether to report data is influenced by a bank's financial health, in which case the inter-temporal variation in fragility is likely to be captured better by the stricter rules. Note that the strictest rule ( 25 ) is associated with the largest inter-temporal standard deviations. Less restrictive rules may under-estimate inter-temporal changes because a nationwide worsening of fragility causes the most fragile banks to disappear from the sample.

Table 20 shows that the patterns in the South African data resemble those of the Netherlands much more closely than those of Nigeria. One pos-

[^9]Table 18: Summary statistics for Netherlands

| Variable | Obs. | Mean | S. Dev. | Minimum | Maximum |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |
| Equity | 15 | 4.64 | 1.02 | 3.22 | 6.10 |
| EquityR5 | 9 | 4.34 | 0.65 | 3.56 | 5.20 |
| EquityH5 | 9 | 4.18 | 0.50 | 3.47 | 5.00 |
| EquityR | 15 | 6.35 | 2.59 | 3.64 | 10.94 |
| EquityH | 15 | 4.99 | 1.58 | 3.22 | 8.86 |
|  |  |  |  |  |  |
| ImpLoans | 9 | 2.00 | 0.56 | 1.18 | 2.74 |
| ImpLoansR5 | 9 | 2.00 | 0.69 | 1.08 | 3.36 |
| ImpLoansH5 | 9 | 2.01 | 0.57 | 1.19 | 2.76 |
| ImpLoansR | 9 | 2.00 | 0.69 | 1.08 | 3.36 |
| ImpLoansH | 9 | 2.00 | 0.56 | 1.18 | 2.73 |
|  |  |  |  |  |  |
| Costs | 15 | 62.32 | 6.22 | 51.21 | 72.94 |
| CostsR5 | 9 | 65.14 | 2.42 | 61.14 | 69.00 |
| CostsH5 | 9 | 65.37 | 2.05 | 62.79 | 69.42 |
| CostsR | 15 | 60.51 | 7.45 | 49.16 | 70.10 |
| CostsH | 15 | 59.45 | 8.12 | 46.10 | 68.78 |
| Returns |  |  |  |  |  |
| ReturnsR5 | 15 | 0.40 | 0.38 | -0.56 | 1.03 |
| ReturnsH5 | 9 | 0.37 | 0.11 | 0.18 | 0.47 |
| ReturnsR | 9 | 0.27 | 0.32 | -0.53 | 0.49 |
| ReturnsH | 15 | 0.60 | 0.71 | -1.03 | 2.30 |
| NetLoans | 15 | 0.41 | 0.42 | -0.57 | 1.20 |
| NetLoansR5 | 15 | 60.82 | 6.40 | 49.99 | 70.80 |
| NetLoansH5 | 9 | 60.15 | 61.74 | 4.97 | 52.69 |
| NetLoansR | 15 | 58.85 | 4.83 | 54.35 | 62.19 |
| NetLoansH | 15 | 57.59 | 3.50 | 49.91 | 63.82 |
|  |  |  |  |  | 66.84 |

Table 19: Summary statistics for Nigeria

| Variable | Obs. | Mean | S. Dev. | Minimum | Maximum |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |
| Equity | 15 | 11.63 | 5.42 | -3.41 | 19.18 |
| EquityR5 | 14 | 3.51 | 22.97 | -52.04 | 22.12 |
| EquityH5 | 15 | 11.74 | 5.59 | -3.84 | 19.59 |
| EquityR | 14 | 5.63 | 18.07 | -37.66 | 21.83 |
| EquityH | 15 | 11.72 | 5.38 | -2.81 | 20.00 |
|  |  |  |  |  |  |
| ImpLoans | 15 | 15.55 | 7.68 | 4.04 | 32.10 |
| ImpLoansR5 | 13 | 23.59 | 16.27 | 5.92 | 63.52 |
| ImpLoansH5 | 15 | 14.84 | 7.53 | 3.58 | 31.16 |
| ImpLoansR | 13 | 23.59 | 16.27 | 5.92 | 63.52 |
| ImpLoansH | 15 | 14.84 | 7.53 | 3.58 | 31.16 |
|  |  |  |  |  |  |
| Costs | 15 | 69.83 | 14.34 | 56.68 | 101.94 |
| CostsR5 | 13 | 78.74 | 39.52 | 49.99 | 164.49 |
| CostsH5 | 15 | 69.62 | 15.37 | 56.33 | 107.72 |
| CostsR | 14 | 73.03 | 31.89 | 43.18 | 145.64 |
| CostsH | 15 | 69.42 | 14.58 | 55.83 | 105.38 |
| Returns |  |  |  |  |  |
| ReturnsR5 | 15 | 1.47 | 3.91 | -10.57 | 4.15 |
| ReturnsH5 | 14 | -0.61 | 12.26 | -40.06 | 4.88 |
| ReturnsR | 14 | 1.45 | 4.22 | -11.32 | 4.27 |
| ReturnsH | 15 | 1.46 | 9.71 | -31.27 | 4.82 |
| NetLoans |  |  | 3.98 | -10.66 | 4.09 |
| NetLoansR5 | 14 | 35.79 | 31.79 | 3.32 | 31.10 |
| NetLoansH5 | 15 | 36.00 | 3.51 | 15.82 | 41.49 |
| NetLoansR | 14 | 32.88 | 5.01 | 22.28 | 44.43 |
| NetLoansH | 15 | 35.83 | 3.49 | 31.25 | 43.36 |

Table 20: Summary statistics for South Africa

| Variable | Obs. | Mean | S. Dev. | Minimum | Maximum |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |
| Equity | 15 | 9.50 | 4.10 | 6.00 | 18.47 |
| EquityR5 | 9 | 6.45 | 0.62 | 5.82 | 7.51 |
| EquityH5 | 13 | 7.34 | 1.61 | 5.94 | 11.57 |
| EquityR | 9 | 6.65 | 0.69 | 5.98 | 7.85 |
| EquityH | 15 | 8.11 | 2.21 | 6.00 | 14.30 |
|  |  |  |  |  |  |
| ImpLoans | 14 | 5.24 | 4.82 | 1.71 | 20.84 |
| ImpLoansR5 | 9 | 3.74 | 2.08 | 1.59 | 6.72 |
| ImpLoansH5 | 13 | 3.42 | 1.92 | 0.73 | 6.69 |
| ImpLoansR | 9 | 3.74 | 2.08 | 1.59 | 6.72 |
| ImpLoansH | 13 | 3.42 | 1.92 | 0.73 | 6.69 |
|  |  |  |  |  |  |
| Costs | 15 | 59.86 | 13.92 | 46.49 | 102.05 |
| CostsR5 | 9 | 56.31 | 4.70 | 49.84 | 64.24 |
| CostsH5 | 13 | 53.51 | 8.10 | 34.31 | 63.93 |
| CostsR | 9 | 56.34 | 4.69 | 49.96 | 64.32 |
| CostsH | 15 | 60.21 | 13.68 | 47.84 | 102.05 |
|  |  |  |  |  |  |
| Returns | 15 | 1.33 | 0.46 | 0.34 | 2.13 |
| ReturnsR5 | 9 | 1.17 | 0.22 | 0.88 | 1.45 |
| ReturnsH5 | 13 | 1.44 | 0.51 | 0.92 | 2.91 |
| ReturnsR | 9 | 1.20 | 0.23 | 0.89 | 1.50 |
| ReturnsH | 15 | 1.27 | 0.41 | 0.34 | 1.86 |
| NetLoans | 15 | 64.62 | 7.97 | 44.63 | 75.35 |
| NetLoansR5 | 9 | 67.61 | 2.13 | 64.85 | 70.59 |
| NetLoansH5 | 13 | 67.31 | 3.27 | 59.69 | 72.45 |
| NetLoansR | 9 | 67.27 | 2.20 | 64.29 | 70.44 |
| NetLoansH | 15 | 64.15 | 8.72 | 44.63 | 75.34 |
|  |  |  |  |  |  |

Table 21: Summary statistics for the United States

| Variable | Obs. | Mean | S. Dev. | Minimum | Maximum |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |
| Equity | 15 | 7.64 | 0.72 | 6.84 | 9.12 |
| EquityR5 | 15 | 8.62 | 0.84 | 7.34 | 9.86 |
| EquityH5 | 15 | 8.89 | 0.86 | 7.57 | 10.15 |
| EquityR | 15 | 7.48 | 0.84 | 6.39 | 9.08 |
| EquityH | 15 | 7.65 | 0.73 | 6.84 | 9.12 |
|  |  |  |  |  |  |
| ImpLoans | 15 | 2.47 | 1.39 | 0.86 | 5.08 |
| ImpLoansR5 | 15 | 2.57 | 1.37 | 0.96 | 4.86 |
| ImpLoansH5 | 15 | 2.56 | 1.47 | 0.86 | 5.27 |
| ImpLoansR | 15 | 2.56 | 1.36 | 0.96 | 4.82 |
| ImpLoansH | 15 | 2.56 | 1.47 | 0.86 | 5.27 |
|  |  |  |  |  |  |
| Costs | 15 | 58.90 | 7.84 | 53.04 | 85.64 |
| CostsR5 | 15 | 55.76 | 5.19 | 49.41 | 70.06 |
| CostsH5 | 15 | 56.96 | 5.38 | 50.19 | 71.73 |
| CostsR | 15 | 58.42 | 8.11 | 52.56 | 86.01 |
| CostsH | 15 | 58.77 | 7.98 | 52.49 | 85.88 |
|  |  |  |  |  |  |
| Returns | 15 | 0.80 | 0.49 | -0.54 | 1.20 |
| ReturnsR5 | 15 | 0.84 | 0.48 | -0.31 | 1.26 |
| ReturnsH5 | 15 | 0.87 | 0.49 | -0.33 | 1.30 |
| ReturnsR | 15 | 0.79 | 0.49 | -0.54 | 1.19 |
| ReturnsH | 15 | 0.80 | 0.49 | -0.54 | 1.20 |
| NetLoans | 15 | 50.99 | 1.98 | 45.89 | 53.38 |
| NetLoansR5 | 15 | 56.32 | 2.12 | 51.61 | 59.26 |
| NetLoansH5 | 15 | 55.98 | 2.14 | 50.61 | 58.82 |
| NetLoansR | 15 | 51.35 | 2.08 | 47.41 | 54.95 |
| NetLoansH | 15 | 51.11 | 1.93 | 45.90 | 53.43 |
|  |  |  |  |  |  |

Table 22: Summary statistics for Ukraine

| Variable | Obs. | Mean | S. Dev. | Minimum | Maximum |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  | 15 | 12.23 | 2.16 | 10.00 | 17.33 |
| Equity | 15 | 10.19 | 2.23 | 3.88 | 14.15 |
| EquityR5 | 15 | 10.82 | 2.69 | 3.88 | 14.88 |
| EquityH5 | 15 | 11.19 | 2.86 | 4.56 | 17.76 |
| EquityR | 15 | 11.76 | 3.00 | 4.56 | 17.76 |
| EquityH |  |  |  |  |  |
|  | 15 | 11.46 | 8.49 | 1.84 | 26.44 |
| ImpLoans | 15 | 9.94 | 7.08 | 2.03 | 23.00 |
| ImpLoansR5 | 15 | 10.90 | 8.16 | 1.91 | 25.63 |
| ImpLoansH5 | 15 | 9.94 | 7.08 | 2.03 | 23.00 |
| ImpLoansR | 15 | 10.90 | 8.16 | 1.91 | 25.63 |
| ImpLoansH | 15 |  |  |  |  |
|  |  |  |  |  |  |
| Costs | 15 | 63.88 | 9.24 | 44.94 | 85.58 |
| CostsR5 | 15 | 59.63 | 8.76 | 43.43 | 72.97 |
| CostsH5 | 15 | 58.26 | 8.39 | 42.59 | 72.91 |
| CostsR | 15 | 65.65 | 13.37 | 44.64 | 107.19 |
| CostsH | 15 | 65.51 | 13.50 | 44.94 | 107.19 |
|  |  |  |  |  |  |
| Returns | 15 | 1.07 | 2.02 | -3.94 | 6.22 |
| ReturnsR5 | 15 | 0.74 | 1.60 | -3.06 | 4.12 |
| ReturnsH5 | 15 | 0.77 | 1.61 | -3.29 | 4.12 |
| ReturnsR | 15 | 0.77 | 2.20 | -4.03 | 6.06 |
| ReturnsH | 15 | 0.82 | 2.15 | -3.94 | 6.06 |
|  |  |  |  |  |  |
| NetLoans | 15 | 64.95 | 12.17 | 39.41 | 82.52 |
| NetLoansR5 | 15 | 66.95 | 11.53 | 43.72 | 83.23 |
| NetLoansH5 | 15 | 66.48 | 11.31 | 43.72 | 83.39 |
| NetLoansR | 15 | 65.83 | 11.74 | 44.16 | 82.63 |
| NetLoansH | 15 | 65.35 | 11.53 | 44.16 | 82.52 |
|  |  |  |  |  |  |

sible explanation for this is that accounting standards in South Africa are much higher than those in Nigeria. There is no compulsion to report data to Bankscope, and the propensity to report data consistently, even when it reveals an increase in bank fragility, may be a function of local culture associated with expectations around accounting standards.

The United States (Table 21) is the nation with the largest number of banks in the dataset, has a high level of economic development and a well supervised banking system. All selection rules entail the full 15 annual observations of each variable, and alternative rules produce similar sample statistics. For some variables (for example ROAA) the consistency is even greater than in Argentina and Indonesia.

In Ukraine (Table 22) there are also 15 observations for all variables, regardless of the selection rule. The variability in sample statistics is of a similar order of magnitude to that in Argentina and Indonesia, and greater than in the United States. However, unlike in Argentina and Indonesia, there is no obvious connection between the strictness of the rule and the implied level of fragility.

### 5.3 Principal component analysis of the variation due to selection rules

Table 23 provides further information on the relative importance of the selection rule for the sample statistics of each variable, focussing on the eight countries discussed in the previous section. For each variable and for each country, the table reports the first principal component of the five different measures of the variable.

Table 23 shows that the highest level of consistency is in South Africa,
Table 23: Explained variation of the first principal component

|  | Equity | Impaired Loans | Cost to income | ROAA | Net Loans |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Whole Sample | 0.82 | 0.95 | 0.87 | 0.92 | 0.91 |
| Argentina | 0.65 | 0.99 | 0.99 | 0.99 | 0.99 |
| Indonesia | 0.99 | 0.99 | 0.77 | 0.99 | 0.99 |
| Libya | 0.96 | $\mathrm{~N} / \mathrm{A}$ | 0.82 | 0.92 | 0.98 |
| Netherlands | 0.88 | 0.83 | 0.68 | 0.89 | 0.86 |
| Nigeria | 0.94 | 0.89 | 0.91 | 0.98 | 0.60 |
| South Africa | 0.99 | 0.99 | 0.99 | 0.98 | 0.98 |
| United States | 0.90 | 0.99 | 0.96 | 0.99 | 0.91 |
| Ukraine | 0.88 | 0.99 | 0.65 | 0.96 | 0.99 |

with the first principal component explaining at least $98 \%$ of the variation
in the measures in all cases. The figures for the United States are almost as high. While the figures for other countries are generally in excess of $90 \%$, each of the other countries has at least one variable with a much lower figure, around $60-70 \%$. For most countries the outlier is managerial efficiency, but for Argentina it is bank capitalisation and for Nigeria it is net loans divided by total assets.

### 5.4 Correlations across the indicators of financial fragility

The eight different variables in our dataset capture different dimensions of financial fragility. Table 24 reports coefficients of correlation across the eight variables, providing information about how closely related these different dimensions are. The correlations are based on the whole dataset.
The table shows that in no case is the correlation very high. The coef-

Table 24: Correlations between the financial fragility indicators

|  | Equity | Impaired Loans | Cost to Income | Return on Av. Assets | Net Loans | Liquid Assets | $\begin{gathered} \hline \hline \text { Net Charge } \\ \text { Offs } \end{gathered}$ | Z-Score |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Equity | 1 |  |  |  |  |  |  |  |
| Impaired Loans | -0.07 | 1 |  |  |  |  |  |  |
| Cost to Income | -0.04 | 0.16 | 1 |  |  |  |  |  |
| Return on Av. Assts | 0.32 | -0.18 | -0.28 | 1 |  |  |  |  |
| Net Loans | -0.01 | -0.18 | -0.10 | -0.08 | 1 |  |  |  |
| Liquid Assets | 0.02 | 0.17 | 0.01 | 0.11 | -0.71 | 1 |  |  |
| Net Charge Offs | 0.01 | 0.12 | 0.19 | -0.09 | -0.17 | 0.10 | 1 |  |
| Z-Score | 0.24 | -0.13 | -0.17 | 0.10 | 0.16 | -0.15 | -0.12 | 1 |

ficient with the largest absolute value (0.71) is for the correlation between two variables based directly on elements of the banks' balance sheet: net loans divided by total assets and liquid assets divided by total assets. Other correlations are very much lower, the next largest coefficient (0.32) being for bank capitalisation and ROAA. Each of the variables appears to capture a different element of financial fragility. Therefore, whether a banking system is considered fragile depends very much upon which dimensions of fragility are given the most weight.

### 5.5 Advantages of including additional bank-types

Whereas existing financial sector datasets focus exclusively on commercial banks, this database includes five additional types of institutions, comprising cooperative banks, investment banks, Islamic banks, real estate and mortgage banks and savings banks. Neglecting these additional types of banks can lead
to misleading conclusions about the level of fragility in the financial system. The pivotal role played by investment banks and real estate and mortgage banks in the latest global financial crisis is well known, Bordo (2009). It is, therefore, likely that their omission may lead to under-measurement of financial fragility. Alternatively, in countries such as Germany, where savings banks are more prevalent (and are generally more risk-averse than commercial banks), their omission may result in over-measurement of financial fragility.

To illustrate some of the advantages of including additional bank types, we also constructed the financial fragility indicators using just commercial banks and made comparisons with our measures for a range of representative countries. What we found was that unless the financial system is almost completely dominated by commercial banks, there were very large differences between the indicators. For countries such as Argentina, Brazil and Indonesia, where commercial banks represent over $95 \%$ of total banking assets, the omission of other bank types does not matter. However, the differences are quite considerable even for countries in which commercial banking assets represent $80 \%$ of total assets. The case of the Netherlands provides a good illustration of this point. In the Netherlands, commercial banking assets represent $79.8 \%$ of total banking assets, while cooperative banks represent $17.5 \%$; real estate and mortgage banks have a $1.7 \%$ share and investment banks $0.9 \%$. While most of the fragility indicators do not change much when all bank types are included, the Z-score for the Netherlands jumps from 9.2 for commercial banks to 13.4, indicating a much less fragile banking system than is suggested by looking at commercial banks alone. ${ }^{19}$

Not surprisingly, in countries where commercial banks have even lower shares in total banking assets, the comparison between our measures and this obtained by focussing on commercial banks alone reveals even greater differences. We use the examples of the United States and Germany to illustrate this point but also to show that focussing on commercial banks alone can lead to both under-measurement or over-measurement of financial fragility.

In the United States, investment banks and real estate and mortgage banks hold $22.0 \%$ and $10.7 \%$, respectively, of total banking assets while commercial banks account for $58.4 \%$ and savings banks for $7.8 \%$. Focussing on commercial banks alone, leads to under-measurement of financial fragility. Specifically, average capitalisation is 2.0 percentage points higher for commercial banks than for all banks, impaired loans lower by 0.6 percentage points and the Z-score climbs from 17.2 to 26.5 when other bank types are

[^10]excluded.
In Germany commercial banks hold $42.4 \%$ of total banking assets, while savings banks hold $25.6 \%$, cooperative banks $14.9 \%$, real estate and mortgage banks $14.0 \%$ and investment banks $3.1 \%$. Germany provides an example in which financial fragility would be over-estimated if additional bank types are excluded. Bank capitalisation increases by 0.6 percentage points on average while the Z-score more than doubles, jumping from 14.1 to 29.8 when other bank types are included.

Overall, these comparisons show that financial fragility can be substantially mis-measured if one utilises indicators that focus exclusively on commercial banks.

## 6 Conclusion

This paper provides a description of a new country-level dataset on financial fragility based on data for individual banks reported in Bankscope. The dataset includes eight different country-level indicators of financial fragility, including a bank capitalisation measure, an asset quality indicator, a managerial efficiency measure, an asset return measure, two different liquidity indicators, a measure of risk exposure and an overall indicator of financial stability.

Particular attention is given to the issues arising from the fact that banks are under no compulsion to report data, and that the propensity to report may be correlated with bank characteristics, or changes in bank characteristics. Although there is no one straightforward solution to these problems, they can be mitigated by using alternative rules for the inclusion of individual banks in the sample, and comparing results using different rules. This facility for robustness checks is an original and distinctive characteristic of the dataset. Furthermore, by covering a wide range of financial institutions (not just commercial banks), the dataset provides an overview of financial development and financial fragility that is broad in scope.

It is intended that this dataset will be used both for academic research and to inform policy-making. In the future, analysis of the dataset might inform questions such as: how financial fragility influences economic growth, whether countries that liberalise their financial systems too quickly become more vulnerable to financial fragility, and whether there are indicators of fragility that can be used for predicting financial crises.

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## Appendix

Table I: Country list

| Albania | Algeria | Angola |
| :---: | :---: | :---: |
| Argentina | Australia | Austria |
| Azerbaijan | Bangladesh | Belarus |
| Belgium | Benin | Bolivia |
| Botswana | Brazil | Bulgaria |
| Burkina Faso | Burundi | Cameroon |
| Canada | Cape Verde | Central African Republic |
| Chad | Chile | China |
| Colombia | Congo | Dem. Rep Congo |
| Costa Rica | Cote d'Ivoire | Czech Republic |
| Denmark | Djibouti | Dominican Republic |
| Ecuador | Egypt | El Salvador |
| Equatorial Guinea | Eritrea | Estonia |
| Ethiopia | Finland | France |
| Gabon | Gambia | Georgia |
| Germany | Ghana | Greece |
| Guatemala | Guinea | Guinea-Bissau |
| Hong Kong | Hungary | India |
| Indonesia | Ireland | Israel |
| Italy | Jamaica | Japan |
| Jordan | Kazakhstan | Kenya |
| Korea (Rep) | Kyrgyzstan | Latvia |
| Lesotho | Liberia | Libya |
| Lithuania | Madagascar | Malawi |
| Malaysia | Mali | Mauritania |
| Mauritius | Mexico | Morocco |
| Mozambique | Namibia | Nepal |
| Netherlands | New Zealand | Nicaragua |
| Niger | Nigeria | Norway |
| Pakistan | Paraguay | Peru |
| Philippines | Poland | Portugal |
| Romania | Russia | Rwanda |
| Sao Tome \& Principe | Senegal | Seychelles |
| Sierra Leone | Singapore | South Africa |
| Spain | Sri Lanka | Sudan |
| Swaziland | Sweden | Switzerland |
| Taiwan | Tanzania | Thailand |
| Togo | Tunisia | Turkey |
| Uganda | Ukraine | United Kingdom |
| United States | Uruguay | Uzbekistan |
| Venezuela | Vietnam | Zambia |
| Zimbabwe |  |  |

Table II: Full variable definitions in Bankscope

| Variable | Components |
| :--- | :--- |
| Loans (2000) | Residential mortgage loans that are secured by <br> property, non-residential loans or total mortgage <br> loans with no breakdown, other consumer and re- <br> tail loans/leases to individuals either secured or un- <br> secured by assets other than residential property (in- <br> cluding personal loans and credit cards), corporate <br> and commercial loans to enterprises (business loans), <br> all other loans and leases that do not fall into any <br> other category (including leased assets, bills of ex- <br> change) minus reserves for possible losses on im- <br> paired loans. |

Total Assets (2025)
Total loans, securities lent out or used as collateral for funding proposes, all securities and assets classified as "held for trading" excluding derivatives, all in the money trading derivatives recognised for hedging less the value of netting arrangements, investment securities designated as available for sale recorded at fair value, investment securities held to maturity at cost value, stakes in associated companies and subsidiaries, all other securities not classified above, investments in property, cash and non-interest earning balances with central banks, real estate acquired as a result of foreclosure on a loan secured by property, fixed assets (property etc.), goodwill net of impairment, intangible assets excluding goodwill (patents and copyrights etc.), tax assets to be refunded for the current year, deferred tax assets, assets of a business which has been sold or written off, any other assets not categorised (e.g. prepayments).

Common shares and premium, retained earnings, statuary reserves and reserves held for general banking risks, loss absorbing minority interests, reserves available for sale, foreign exchange reserves, other revaluation reserves, and preference shares and hybrid capital accounted for as capital, redeemable capital in cooperative banks.

Total liabilities and equity (2060)

Loan loss reserves (2070) Reserves against possible losses on impaired loans, accumulated credit provisions for off-balance sheet items such as guarantees (securities reserves).

Net Interest Revenue (2080) Interest and commission received on loans, advances and leasing, interest income from the trading book, short-term funds and investment securities excluding insurance related interest, dividend income from trading and available for sale and held to maturity investments minus interest paid to customers' deposits including commission fees and transaction costs, interest paid on debt securities and other borrowed funds excluding insurance related interest expenses if separately identified, preferred dividends paid and declared.

Other Operating Income Net gains (losses) on items disclosed as trading as(2085)

Common shares and premium, retained earnings, statuary reserves and reserves held for general banking risks, loss absorbing minority interests, reserves available for sale, foreign exchange reserves, other revaluation reserves, hybrid capital (no breakdown provided), and the sum of all financial liabilities.
sets/portfolio in the accounts, gains (losses) on trad- ing derivatives, net gains (losses) on items input into "available for sale securities" "held to maturity securities" or "other securities", net gains (losses) on assets disclosed as "assets at fair value through income", share of profit from associates and others under equity accounting.

Overhead costs (2090) Includes wages, salaries, social security costs, pension costs, and other staff costs including expensing of staff stock options, depreciation, amortisation, administrative expenses, occupancy costs, software costs, operating lease rentals, audit and professional fees, other administrative operating expenses.

Net Income (2115)

Impaired Loans (2170) Includes impaired loans stated in the bank's accounts. According to the Fitch Universal Model used in Bankscope, this includes the total value of the loans that have a specific impairment against them. This includes, non-accrual loans, restructured loans, watchlist loans and any loans 90 days overdue. Some banks may not include restructured or watchlist loans. Some loans may be classed in more than one category. Moreover, non-performing loans and charge offs are driven by national regulations and may not be directly comparable across countries. In fact, even within Europe there are huge differences, something the European Banking Authority (EBA) is currently addressing.

Table III: The number of banks in Africa

| Country Name | Frequency | Country Name | Frequency |
| :--- | :--- | :--- | :--- |
| Nigeria | 96 | Burkina Faso | 10 |
| South Africa | 59 | Mali | 10 |
| Kenya | 56 | Benin | 9 |
| Egypt | 39 | Gambia | 9 |
| Tanzania | 34 | Namibia | 8 |
| Ghana | 33 | Rwanda | 8 |
| Zimbabwe | 29 | Burundi | 7 |
| Sudan | 25 | Niger | 7 |
| Mauritius | 22 | Togo | 7 |
| Uganda | 22 | Gabon | 6 |
| Tunisia | 21 | Madagascar | 6 |
| Zambia | 21 | Chad | 5 |
| Morocco | 20 | Guinea | 5 |
| Algeria | 19 | Liberia | 5 |
| Cote d'Ivoire | 18 | Seychelles | 5 |
| Angola | 17 | Swaziland | 5 |
| Malawi | 15 | Cape Verde | 4 |
| Mozambique | 15 | Congo | 4 |
| Democratic Republic of Congo | 14 | Lesotho | 4 |
| Cameroon | 13 | Eritrea | 3 |
| Senegal | 13 | Sao Tome and Principe | 3 |
| Sierra Leone | 13 | Central African Republic | 2 |
| Botswana | 12 | Djibouti | 2 |
| Ethiopia | 11 | Equatorial Guinea | 2 |
| Libya | 11 | Guinea Bissau | 1 |
| Mauritania | 11 | Total | 796 |

Table IV: Variable name codes

| Variable <br> Code Name | Definition |
| :--- | :--- |
| Var | This is the base case where the variable uses all the informa- <br> tion available. Therefore no rules are used in the construction <br> of the country level data under this scenario. |
| VarR5 | This variable requires that all five CAMELS financial variable <br> ratios are reported simultaneously for a minimum of eight <br> years (hence, a bank reports all five ratios simultaneously at <br> least eight times) during its existence in the data set. If a <br> bank reports other financial variables more frequently (e.g <br> one variable is reported for the duration of the panel) then <br> this data will be used in the construction of the country level <br> variables for that particular variable. |

VarH5 This variable requires that all five CAMELS financial variable ratios are reported simultaneously for a minimum of $66 \%$ of a banks duration during its presence in the dataset. A bank is assumed to enter the dataset the first year it reports any financial ratio and is assumed to remain in the dataset unless it provides a specified reason for leaving (for example becoming inactive or remaining active and leaving the Bankscope database). Hence, a bank that enters the database in 2004 (the first year it reports a financial variable), remains in the panel until the end and subsequently reports all five CAMELS variables simultaneously for six years, then its observations will be used in the construction of the country level data (as it reports in six of the nine years possible). Moreover, as before if it reports one of its financial ratios for more than the minimum six years (say annually) these observations will be used to construct the country level data, and not just purely the times all five CAMELS ratios are reported simultaneously.

| VarR | For a banks data to be used in the construction of the country <br> level data, this rule requires that a bank reports this variable <br> at least eight times during the time series, regardless of the <br> time the bank enters the database or whether or not the bank <br> exits the database. |
| :--- | :--- |
| $\operatorname{VarH} \quad$ | This rule requires that the selected variable is reported by the <br> bank for at least $66 \%$ of the time during a banks duration in <br> the panel. A bank is assumed to enter the panel the first year <br> it reports a value, and exits the panel if it becomes inactive <br> (specified in the Bankscope data) or is active but leaves the <br> Bankscope database, where the year it leaves is assumed as <br> the last year the bank reports. |
| Notes: Var may represent any variable used in the study e.g. impaired loans divided |  |
| by total assets, cost to income ratio or net loans divided by total assets etc. |  |


| $\tau$ | $\tau$ | $\checkmark$ | $\checkmark$ | $\tau$ | $\tau$ | $\checkmark$ | $\tau$ | $\tau$ | $\tau$ | $\checkmark$ | ${ }^{2}$ | $\tau$ | $\checkmark$ | I | I | ！！qndəy ueว！up［exquә， |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\pm$ | $\pm$ | $\dagger$ | ■ | $\dagger$ | $\varepsilon$ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| ¡てI | LII | LII | ¢01 | L9 | 09 | 09 | 69 | 99 | ¢¢ | LG | gt | 珧 | 98 | 78 | I¢ | ереие， |
| ¢I | ¢I | ¢L | ¢L | ¢1 | ¢1 | 2 L | 2 L | ${ }^{2}$ | 01 | 01 | 0I | 8 | 8 | 8 | 2 | иоо．әшия， |
| $\llcorner$ | 2 | $\llcorner$ | 2 | $\llcorner$ | $\llcorner$ | 2 | 2 | 2 | $\angle$ | 2 | 2 | 9 | g | g | g | ！pun．rng |
| 01 | 01 | 01 | 01 | 01 | 01 | 6 | 6 | 8 | 8 | 4 | 4 | 4 | $L$ | 4 | $\pm$ | osear bupurng |
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## A note on Islamic banking

One category of bank in Bankscope is Islamic banks, this category appearing in 15 of the 124 countries. Table VII shows the prevalence of Islamic banks relative to other banks. In only the Gambia, Jordan, Pakistan, Malaysia, Mauritania and Sudan does the number of Islamic banks exceed $10 \%$ of the total number of banking institutions in the country, and in no country does it exceed $50 \%$ (Sudan is the closest, with a figure of $48 \%$ ). The relative size of the Islamic banking sector in each country is shown in the final column of the table. Here, size is measured by mean total assets over the sample period. In only three cases does the share of total assets exceed $10 \%$.

Table VII: Islamic bank prevalence

| Country <br> Name | Islamic Bank <br> Frequency | Bank <br> Numbers | Bank <br> Perrent | Share of Total <br> Assets Percent |
| :--- | ---: | ---: | ---: | ---: |
| Bangladesh | 2 | 39 | 5.13 | 1.88 |
| Egypt | 3 | 39 | 7.69 | 4.01 |
| Gambia | 1 | 9 | 11.11 | 3.19 |
| Indonesia | 3 | 132 | 2.27 | 2.24 |
| Jordan | 3 | 19 | 15.79 | 3.83 |
| Malaysia | 18 | 105 | 17.14 | 11.42 |
| Mauritania | 2 | 11 | 18.18 | 11.63 |
| Pakistan | 9 | 57 | 15.79 | 3.68 |
| Philippines | 1 | 66 | 1.52 | 0.01 |
| Russia | 1 | 1,186 | 0.08 | 0.00 |
| Singapore | 1 | 67 | 1.49 | 0.10 |
| Sudan | 12 | 25 | 48.00 | 43.05 |
| Tunisia | 1 | 21 | 4.76 | 1.06 |
| Turkey | 5 | 69 | 7.25 | 4.15 |
| United Kingdom | 5 | 377 | 1.33 | 0.01 |

Notes: Certain institutions failed to report total asset data which included:
Indonesia (35), Malaysia (6), Pakistan (1), Philippines (5), Russia (27), Singapore (8), Turkey (1), United Kingdom (43).


[^0]:    *We acknowledge the support of ESRC-DFID grant number ES/J009067/1

[^1]:    ${ }^{1}$ The bank-level data downloaded is version 275.1 and available along with the newly constructed macro data at http://www2.le.ac.uk/departments/economics/research/esrc-dfid-project.
    ${ }^{2}$ These types are pre-defined in Bankscope. However, it is important to note that the activities of an individual bank may span several categories. For example, in some counties a commercial bank may also be involved in investment banking activities.
    ${ }^{3}$ In the UK these are known as building societies.

[^2]:    ${ }^{4}$ The Appendix offers a detailed note on Islamic banking prevalence in the database.
    ${ }^{5}$ These findings conform to what we would expect as these bank types are generally more risk-averse than commercial banks, Beck et al. (2009).

[^3]:    ${ }^{6}$ We considered trimming/winsorising the bank-level data to remove possible outliers but decided against that following assurances from Bankscope that the data is triplechecked before it appears online. Specifically, the procedure for uploading and checking the data is as follows (i) banks update directly to Fitch using some form of automatic procedure (e.g. Oracle or some database) and the data entries are checked by Fitch (ii) Bankscope gets the data from Fitch and checks it (iii) Bankscope uploads and checks the data again to verify it has been uploaded correctly.
    ${ }^{7}$ Net charge-offs are generally used to measure risk exposure, but also may measure asset quality. Banks charge off bad debt (i.e. remove it from their loan portfolios) when all other methods to reclaim the loan have failed, so charge-offs may be a lagging indicator

[^4]:    ${ }^{10}$ One alternative measure is the number of employees in a bank; however, few banks report this figure.
    ${ }^{11}$ The appendix provides a complete breakdown of the number of banks in each African country.

[^5]:    ${ }^{12}$ The additional three variables (the Z-score, liquid assets divided by total assets and net charge-offs divided by gross loans) are constructed using the base sample selection rule only.

[^6]:    ${ }^{13}$ The Appendix provides more detailed information about the appearance of individual banks in the database.

[^7]:    ${ }^{14}$ In the computation of this standard deviation, the total number of years varies across countries: in some countries there are years with no data at all.

[^8]:    ${ }^{15}$ However, there was a large increase in the number of Ukrainian banks in Bankscope in 2012.
    ${ }^{16}$ This is partly attributable to the vast amount of regulatory reforms throughout the sample period.

[^9]:    ${ }^{17}$ In 1998 the start date of the panel, due to reforms, 26 banking licenses were revoked in Nigeria leaving the total number of banks at 89 . This already begins to question Bankscope's coverage, although 46 banks are reporting in this year, which is over $50 \%$.
    ${ }^{18}$ In July 2004, a regulatory decree was passed that banks had to increase their minimal capital requirements. The intention was to increase bank size and create a more stable banking system. From the 89 banks in 1998, 14 failed to raise capital requirements or merge with another bank, and the number of banks fell to a total of 25 . For further details see Hesse (2007). Our database shows that 31 banks exit the database in this year and Bankscope's coverage of 15 banks is $60 \%$.

[^10]:    ${ }^{19}$ This finding raises the interesting research hypothesis, whether the presence of cooperative banks in a banking system contributes to financial stability.

[^11]:    леәК Кq pue Кıұunoo
    Table V：The cumulative number of banks entering by

