

NBER WORKING PAPER SERIES

DOES INEQUALITY LEAD TO A FINANCIAL CRISIS?

Michael D. Bordo
Christopher M. Meissner

Working Paper 17896
<http://www.nber.org/papers/w17896>

NATIONAL BUREAU OF ECONOMIC RESEARCH
1050 Massachusetts Avenue
Cambridge, MA 02138
March 2012

Michael Bordo is a visiting scholar at the Federal Reserve Bank of Dallas and the Federal Reserve Bank of Cleveland, though the research in this paper was not funded by these institutions. We thank Ju Hyun Pyun, Peter Lindert, James Lothian, Daniel Waldenström, and participants at the 2011 JIMF/Santa Cruz Center for International Economics conference for many helpful comments and suggestions on an early draft of this paper. The aforementioned are of course not responsible for remaining errors. The views expressed herein are those of the authors and do not necessarily reflect the views of the National Bureau of Economic Research.

NBER working papers are circulated for discussion and comment purposes. They have not been peer-reviewed or been subject to the review by the NBER Board of Directors that accompanies official NBER publications.

© 2012 by Michael D. Bordo and Christopher M. Meissner. All rights reserved. Short sections of text, not to exceed two paragraphs, may be quoted without explicit permission provided that full credit, including © notice, is given to the source.

Does Inequality Lead to a Financial Crisis?
Michael D. Bordo and Christopher M. Meissner
NBER Working Paper No. 17896
March 2012
JEL No. E51,N1

ABSTRACT

The recent global crisis has sparked interest in the relationship between income inequality, credit booms, and financial crises. Rajan (2010) and Kumhof and Rancière (2011) propose that rising inequality led to a credit boom and eventually to a financial crisis in the US in the first decade of the 21st century as it did in the 1920s. Data from 14 advanced countries between 1920 and 2000 suggest these are not general relationships. Credit booms heighten the probability of a banking crisis, but we find no evidence that a rise in top income shares leads to credit booms. Instead, low interest rates and economic expansions are the only two robust determinants of credit booms in our data set. Anecdotal evidence from US experience in the 1920s and in the years up to 2007 and from other countries does not support the inequality, credit, crisis nexus. Rather, it points back to a familiar boom-bust pattern of declines in interest rates, strong growth, rising credit, asset price booms and crises.

Michael D. Bordo
Department of Economics
Rutgers University
New Jersey Hall
75 Hamilton Street
New Brunswick, NJ 08901
and NBER
bordo@econ.rutgers.edu

Christopher M. Meissner
Department of Economics
University of California, Davis
One Shields Avenue
Davis, CA 95616
and NBER
cmmeissner@ucdavis.edu

1. Introduction

The recent financial crisis in the U.S. has been attributed to a rise in inequality by several authors. In his 2010 book, *Fault Lines*, Raghuram Rajan argued that rising inequality in the past three decades led to political pressure for redistribution that eventually came in the form of subsidized housing finance. Political pressure was exerted so that low income households who otherwise would not have qualified received improved access to mortgage finance. The resulting lending boom created a massive run-up in housing prices which reversed in 2007 and led to the banking crisis of 2008.

Along these lines, Kumhof and Rancière (2011) study the links between inequality, credit and crises complementing the Rajan hypothesis with a DSGE model. In this model, rising inequality and stagnant incomes in the lower deciles lead workers to borrow to maintain their consumption growth. As these households become increasingly indebted, they continue to borrow more to maintain their consumption. This increases leverage, and eventually a shock to the economy leads to a financial crisis. They posit that their story holds both for the 1920s stock market boom in the US and the run up to the 2008 crisis. The focus on income inequality by Kumhof and Rancière and Rajan is a novel approach to understanding macroeconomic outcomes prior to the recent financial crisis, and to the Great Depression. The theme deserves further empirical scrutiny from other time periods and countries.

There is reason to wonder about the generality of this new view since income inequality rarely plays a significant role in the large literature on financial instability and credit booms. Mendoza and Terrones (2008) study the experience of a large number of advanced and emerging economies since the 1960s finding that current account deficits, strong economic growth and fixed exchange rates accompanied credit booms. Borio and White (2003) have also elaborated a view of pro-cyclical financial systems. Periods of expected low and stable inflation, strong economic growth and liberalized finance can give rise to complacency amongst borrowers, lenders and regulators. Endogenous market forces that might normally “rein in” these imbalances seem to be absent.

Massive buildups in credit lead to financial instability in this case. Income inequality plays no active role in generating the boom-bust outcome in these contributions.

In this paper, we present new empirical evidence on whether rising inequality has any explanatory power in accounting for credit booms and financial crises. Rather than limiting the focus to inequality as the Rajan/Kumhof/Rancière (RKR) frameworks do, we control for more traditional determinants of the credit cycle. Different from these authors, we also bring evidence from a much larger sample than the two unique periods in US economic history that are the focus of RKR. Our sample is a panel of 14 mainly advanced countries from 1920 to 2008 covering a wide variety of boom-bust episodes and financial crises.

We find very little evidence linking credit booms and financial crises to rising inequality. Instead, the two key determinants of credit booms are the upswing of the business cycle or economic expansion and low interest rates. This is very much consistent with a broader literature on credit cycles. While inequality often ticks upwards in the expansionary phase of the business cycle, this factor does not appear to be a significant determinant of credit growth once we condition on other macroeconomic aggregates. Neither is income concentration a good predictor of the financial crises that often follow above average growth in credit. The anecdotal evidence from several historical credit booms finds little support for the inequality/crisis hypothesis.

Section 2 reviews the literature on the link between inequality, credit booms and financial crises. Section 3 presents evidence on the credit boom/financial crisis linkage. Section 4 takes one step back to explore the determinants of credit booms. Here we examine inequality and other determinants of credit growth. Section 5 discusses our findings within the context of historical narratives. Section 6 concludes.

2. Income Inequality, Credit Booms, and Financial Crises

Since the 1970s income concentration has increased dramatically in the US and several other nations. Atkinson, Piketty and Saez (2011) focus on the rapidly rising share of pre-tax income amongst the top 1% of tax units since the 1980s that is marked in several nations such as the US and the UK. Many other studies have confirmed the stagnating wages and incomes for the lower deciles

in the US and the fact that the median wage for US male workers has not risen since 1973 (e.g., Goldin and Katz, 2008). Rajan (2010) attributes rising income inequality in the U.S. since the 1970s largely to problems in the education system.¹

Rajan's argument builds on the evidence in Goldin and Katz (2008) which shows that public education has failed to provide the type of training skilled workers required to keep up with advances in technology. This deficiency is reflected in the incomes of the highly educated that are rising relative to the incomes of the majority of the less educated whose incomes have barely risen since the 1980s in real terms—a rising college premium over secondary school. Goldin and Katz document declines in graduation rates in high school and in college. They argue that this reflects the lack of access by those of low to moderate income to quality education.

According to Rajan, the trend rise in inequality in the U.S. in recent decades, as in earlier episodes such as the social turmoil at the beginning of the twentieth century, (e.g., strikes, anarchism, the IWW, the progressive movement) and then in the 1930s (e.g., the rise of American communism, Coxe's army of the unemployed and major strikes) led to a demand for political intervention. Ostensibly, the goal was to maintain the growth of consumption for the middle class. But unlike in the 1930s, the polarized political system today (e.g., McCarthy et al., 2006) has not been able to use the tax system to redistribute income or to fix the education system. Instead, it is alleged the US political system found it much easier to provide cheap credit for housing. The process relied on government intervention in housing markets as well as financial innovation and deregulation.

In the US, government involvement in the housing markets is largely via the Federal Housing Administration (FHA) and the GSEs -- Fannie Mae and Freddie Mac. The FHA and Fannie Mae have their roots in the 1930s (Freddie Mac in the 1970s), and were set up to encourage the development of the mortgage market and to provide housing finance to much of the population. Rajan argues that as rising inequality and stagnant incomes became apparent in the early 1990s, successive administrations and Congress pushed for affordable housing for low income families using Fannie Mae and Freddie Mac as instruments for redistribution. In 1992 the Federal Housing Enterprise Safety and Soundness Act encouraged HUD to develop affordable housing and allowed

¹ Other factors such as the decline of unions, immigration, globalization, reductions in the top marginal tax rates, and deregulation are given less weight.

the GSEs to reduce their capital requirements. This according to Rajan (2010, p. 35), led the agencies to take on more risk. Lending was encouraged, and rising prices raised the GSE's profits leading to more lending. The Clinton administration encouraged the GSEs to increase their allocations to low income households and urged the private sector "to find creative ways to get low income people into houses" (Rajan, 2010 p. 36). The FHA in the 1990s also took on riskier mortgages, reduced the minimum down payment to 3%, and increased the size of mortgages that would be guaranteed. The housing boom came to fruition in the George W. Bush administration which urged the GSEs to increase their holdings of mortgages to low income households (Rajan, 2010 p. 37). Between 1999 and 2007 national home prices doubled according to the widely followed S&P/Case Shiller repeat sales index. The FHFA house price index shows a 75 percent rise between 1999 and 2007. Either index demonstrates that house prices grew much faster than their long run trend during these years.

The private sector also joined the party as they recognized that the GSEs would backstop their lending (Rajan, 2010 page 38). During this period, lending standards were relaxed and practices like NINJA and NODOC loans were condoned. These developments led to the growth of subprime and Alt A mortgages which were securitized and bundled into mortgage backed securities and then given triple A ratings which contributed to the financial fragility. Mortgage backed securities (MBS) were further repacked into collateralized debt obligations (CDOs). Credit default swaps (CDS) provided insurance on many of these new products. Financial firms ramped up leverage and avoided regulatory oversight and statutory capital requirements with special purpose vehicles (SPVs) and special investment vehicles (SIVs).

In this view, the financial crisis is directly linked to the explosion in credit and the inequality that drove it. Extending loans to low income households worked to increase home ownership in the short-run. At the same time, capital gains on real estate enhanced the amount of home equity and turned houses into ATMs via home equity credit lines. Ultimately the lending was deemed to be fundamentally unsound. Once housing prices began to decline in 2006 and 2007, the stage was set for the subprime mortgage crisis. SIVs faced liquidity runs due to their maturity mismatch problems. Correlated defaults overwhelmed the quality of the lower tranches of MBSs and CDOs and the leverage that begat high returns led to fires sales of all assets. CDS insurance exposure rendered AIG insolvent in this market. Further, the lack of transparency for investment bank balance sheets and

uncertainty about Federal Reserve and government liquidity support led to a total collapse in the intermediation system.

Rajan concludes that the government had good incentives to deal with rising inequality “but the gap between government intent and outcomes can be very wide indeed, especially when action is mediated through the private sector... On net, easy credit, ... proved an extremely costly way to redistribute income” (Rajan, 2010 p.44). Rajan also does not provide extensive direct evidence that the government deliberately chose cheap credit for housing to deal with rising inequality, but he states that:

“... it is easy to be cynical about political motives but hard to establish intent, especially when the intent is something the actors would want to deny—in this case using easy housing credit as a palliative... it may well be that many of the parts played by the key actors was guided by the preferences and applause of the audience, rather than by well thought out intent. Even if the politicians dreamed up a Machiavellian plan to assuage voters with easy loans, these actions—and there is plenty of evidence that politicians pushed for easy housing credit—could have been guided by the voters they cared about... whether the action was driven by conscious intent or unintentional guidance is immaterial to its broader consequences.” (Rajan, 2010 p. 39).²

Kumhof and Rancière (2011) develop a DSGE framework to complement the Rajan hypothesis. They take as stylized facts the correlation between rising inequality and credit growth in the U.S. both in the 1920s before the 1929 stock market crash and in the recent period. In both periods, there was a sharp increase in the incomes commanded by the top 1 per cent of the income distribution and a rapid increase in the household sector’s debt to income ratio. In both instances, they argue that higher credit was an equilibrium outcome. Lower deciles use debt in a bid to maintain real consumption growth that is temporarily below trend while the richest households accumulate claims on these low income households. Meanwhile, the growth in credit and higher leverage in the financial and household sector heightens the possibility of a systemic financial crisis.

This model which has a marked similarity to the one pioneered by Michal Kalecki (1990) has two classes; the rich who own the capital who save invest and consume, and the rest (labor) who earn wages, do not own any capital and consume their wages plus the proceed from any borrowing.

² In a similar vein to Rajan, Stiglitz (2009) hypothesized that in the face of stagnating real incomes, households in the lower part of the distribution borrowed to maintain their living standards leading to an unsustainable credit boom.

Kumhof and Ranci ere assume habit formation so that agents have a minimum level of consumption they need to attain.

Workers and the rich bargain for the surplus created by production, and so consumption for workers depends on bargaining. When a bargaining shock hits, real wage growth declines, real wages fall below their steady state level and workers borrow to maintain their consumption in the face of their falling incomes. The rise in income inequality is driven by a shock that reduces the bargaining power of the workers relative to capitalists. In this case, the rich lend some of their income which they save to the workers via financial intermediaries which they own. Even as the poor become increasingly indebted, they continue to borrow to maintain their standard of living. It is assumed that the probability of a crisis depends positively on leverage. When a crisis occurs the lending boom stops. In this model, the authors present sufficient conditions within a micro-founded equilibrium model linking rising inequality, credit booms and financial crises.

At least two alternative views of credit booms and financial crises exist. In contrast to the ideas surveyed above, income inequality has not typically been seen as a standard driver of credit growth in either of these. The first of these covers a range of macroeconomic indicators. Mendoza and Terrones (2008) illustrate how these forces matter in a data set covering 49 credit booms in 48 advanced and emerging countries between 1960 and 2006. During such booms, countries are likely to experience economic expansions, real exchange rate appreciation, capital inflows/current account deficits, and asset price booms.³ In industrial economies, financial reform has also been a strong correlate of credit booms. Mendoza and Terrones also confirm that credit booms are a strong predictor of financial crises.

In a prescient essay, Borio and White (2003) argued that financial instability could be an outcome of strong credit growth and large financial imbalances. This would be particularly so when expectations for inflation were low and monetary policy sought to fight inflation but accommodated (i.e., did not react to) asset price booms. The drivers of these imbalances differ radically from the

³ Menodza and Terrones note that not all business cycle expansion phases are accompanied by credit booms but that when credit booms occur the economy is likely to be growing above trend. Credit also appears to have its own cycle. This is in contrast to RKR which seem to focus on longer term trends. It would be difficult to reconcile trend inequality with cyclical credit. On the other hand inequality does seem to have a cyclical component according to Barlevy and Tsiddon (2006). Roine, Vlachos and Waldenstr om (2009) also show that top incomes are correlated with periods of economic expansion.

RKR frameworks. In fact, Borio and White tie into an older view of business cycles originating with A.C. Pigou, Irving Fisher, and Hyman Minsky amongst others.

In Borio and White's synthesis, financial liberalization since the 1970s has interacted with the monetary policy environment and low expected inflation since the mid-1980s to create boom-bust cycles. Cross country data reveal a rising frequency of banking crises compared to the 1950-1972 period and a large number of boom-bust episodes for equity prices and credit growth since the 1980s in the leading (G10) countries. They observe that the financial system in these countries is increasingly pro-cyclical owing to the fact that equilibrating mechanisms fail to ignite when asset prices are rising rapidly. Mainly they lay the blame on perceptions of risk and value which tend to be overly optimistic in boom episodes. Strong competition in financial markets only enhances these pressures to reach for yield. Increasing leverage, when feasible with regulatory standards also is a product of the process, and increases the fragility of the system in the event of a systemic shock.

Borio and White note that credit booms can occur in periods of low and stable inflation. Often these periods arise when monetary policy credibility is strong as it has been in advanced and some emerging countries in recent decades. An additional feature is that monetary policy that is aimed exclusively at control of inflation may not be responsive to financial imbalances in the absence of inflation.⁴ This is not an intuitive process for those most familiar with standard equilibrium models of the business cycle. In these mainstream models, demand pressures should lead to inflation and a monetary reaction to such inflation. The economy returns quickly to equilibrium after the monetary shock. Contractionary monetary policy in such models would also limit asset price growth and financial excess (Bernanke and Gertler, 2000).

Borio and White's view is one of a cumulative process that is out of equilibrium. They cite a number of factors that allow inflationary pressures to be limited during credit booms. In such a case monetary intervention is not forthcoming from a central bank focused on short-run price pressures. Expectations of monetary policy credibility and stable inflation can lead actors to believe that recessions induced by monetary tightening will be less likely since limiting rising inflation is the monetary objective. Workers and consumers have balance sheets that improve in the boom and they may be able to easily sustain consumption growth with credit rather than demanding wage

⁴ The corollary is that monetary policy that leans against the financial wind can stop a credit boom. For a related argument see Bordo and Jeanne (2002).

increases.⁵ Firms' accounting profits can improve in times of high asset prices leading to "more aggressive pricing strategies". The prediction from Borio and White is that economic booms and accommodating monetary policy can give rise to large run-ups in credit and asset prices.

The recent literature on credit booms thus posits several hypotheses. One view argues that rising inequality fosters credit booms. Alternative views suggest that credit booms are: pro-cyclical; more likely when monetary policy reacts largely to short-term inflationary pressures; associated with financial sector liberalization and competition. All of these views share the conclusion that credit booms can heighten the probability of a financial crisis. In what follows, we ascertain whether the evidence from a panel of 14 countries over the last 100 years is consistent with either of these explanations for credit booms.

We are not the first to make an effort in this direction. Atkinson and Morelli (2010) use a data set ranging from 1911 to 2010 and look for patterns of rising inequality prior to financial crises similar to that observed in the U.S. before 2007. Like Kumhof and Rancière they find evidence for a strong inverted V in the 1920s and 1930s. They also find a weaker pattern for the decade leading to the Savings and Loan Crisis in the late 1980s. Another case which they argue looks like the U.S. in the 1920s is Iceland preceding its 2007 crash. Still, their conclusion is that the overall evidence is not conclusive that financial crises are preceded by a run-up in income inequality. Moreover, Atkinson and Morelli do not explore the channel we are interested in which is that inequality drives a credit boom. Within their framework, they also do not allow for other determinants of credit booms and crises besides inequality.

2. Empirical Evidence: Credit as a Driver of Banking Crises

To investigate the relationships outlined above, we employ a data set for 14 countries between 1880 and 2008. The first relationship we explore is that between banking crises and credit expansion. The theoretical frameworks outlined above all agree there is a positive relationship between these two outcomes, and a large amount of recent research confirms this relationship empirically. We

⁵ This points out a possible mechanism for inequality to be a by-product of a credit boom rather than a cause. If workers are induced to limit wage demands, but creditors, capitalists, and rentiers take capital gains and benefit from the financial and economic expansion inequality might rise during the boom.

investigate it and confirm it in our data set for illustrative purposes. We then move on to examine the determinants of credit growth including income inequality as well as other variables discussed above.

Borio and Lowe (2002) found evidence that credit booms since the 1970s were associated with financial instability. Menodza and Terrones (2008) note that many, but not all credit booms are followed by banking crises. Schularick and Taylor (forthcoming) go beyond the last few decades and study the long-run evidence. Their data on credit growth, which we rely on here, span the years 1889 to the present covering the most economically advanced countries. They also find a similar positive empirical relationship. In their work, Schularick and Taylor estimated the probability of a systemic banking crisis as a function of credit growth using an estimating equation similar to the following:

$$\text{Pr}(\text{Banking Crisis}) = f\left\{\sum_1^P \beta_p \Delta \ln(\text{Credit}_p) + \mu_i + \delta_t + \varepsilon_{it}\right\}$$

The Δ symbol denotes the annual change, credit is proxied by the ratio of Schularick and Taylor's bank loans variable deflated by the local consumer price index, μ_i is a set of country fixed effects which are included in the linear models, δ_t is a set of time period indicator dummies, and ε_{it} is an idiosyncratic error term for each country in each period. We estimate this model with a linear probability model and alternatively with a logit model without country fixed effects. The lag length, P , is set to 5 as in Schularick and Taylor.⁶

Schularick and Taylor's credit variable is defined as the amount of outstanding domestic currency loans made by domestic banks to domestic households and non-financial corporations. They assume that bank loans is a reasonable proxy for total credit, but this might not be true if substantial amounts of financial liabilities are financed by non-banks. Many types of obligations generated in the equity, bond or derivative markets are not included in this data set. What constitutes a bank and which types of banks reported their lending depends on the country and the time period. In the absence of more comprehensive data for the long run, we follow them in using bank loans as

⁶ The sample for banking crises covers the following countries and years: Australia (1880-2008), Canada (1880-2008), Denmark (1891-2008), France (1886-2008), Germany (1889-2008), Italy (1880-2008), Japan (1894-2008), Netherlands (1906-2008), Norway (1880-2008), Spain (1906-2008), Sweden (1880-2008), Switzerland (1912-2008), United Kingdom (1886-2007), and the United States (1902-2008).

a good proxy for credit. Schularick and Taylor also argue that despite changes in data coverage over time and across space, that time series variation is meaningful, so we follow their approach and always use country fixed effects and/or focus on changes in levels.

A banking crisis is coded as a binary indicator variable. These data are from Bordo, Eichengreen, Klingebiel and Martinez-Peria (2001) and were updated for 2001-2008 using the data underlying Schularick and Taylor (forthcoming). Banking crises involve systemic panics, widespread failures in the banking industry, and large losses to the capital base of the domestic banking system. Banking crises can last multiple years if, for instance, restructuring in the banking sector continues over time and losses last for several years, or re-capitalization does not occur immediately. We include subsequent years after the first year of a crisis but our results are qualitatively robust to using only the first year of a crisis as our dependent variable.

Table 1 presents results for the models and sub-samples. Overall there is a strong positive relationship between real credit growth and the probability of having a banking crisis. Although real credit growth lagged one year is associated with a lower probability of a crisis, credit growth from two to five years earlier is strongly positively related to a crisis. The sum of lag coefficients in column 1 is 0.487. In the linear probability model, the sum of coefficients implies that a sustained five year period rise of one standard deviation or 0.10 log points in real bank loans would be associated with a rise in the probability of a banking crisis of 0.049. The results are somewhat larger if we use the logit specification from column 2. Here a rise in the growth of real credit from 0.05 to 0.15 (roughly one standard deviation above the mean) would raise the predicted probability by 0.15. For the sample that is restricted to the post-World War II period, the impact is slightly smaller. Here there is a rise in the probability of 0.06 when the mean growth of credit rises from its mean of 0.04 by one standard deviation to 0.10. These results are in line with Schularick and Taylor and the literature on credit booms surveyed above. They pave the way to thinking about the fundamental drivers of credit growth.

4. Credit Growth Determinants

In this section we explore the cross-country relationship between the growth in credit and other macroeconomic aggregates. Our dependent variable is credit growth, and the set of independent variables includes income concentration as well as some of the more traditional variables included in Mendoza and Terrones and Borio and White.

The goal here is to provide a simple econometric test of the hypothesis that credit growth has no relationship to changes in income concentration after conditioning on other factors. The alternative hypothesis, discussed above, is that income inequality or top income concentration drives credit growth even after controlling for other macroeconomic variables. Again, there is strong evidence, based on our priors and previous research, that there is a relationship between credit growth and the business cycle and other macroeconomic aggregates. There is also some evidence that changes in inequality are related to these cyclical variables. Whether changes in inequality matter once we control for these variables is then a question for the data.

4.1 Time Series Plots of Credit and Top Incomes

Time series plots of the growth in credit (i.e., the change in the log ratio of bank loans to the price index) and of changes in the share of income for the top 1% for seven different country/time periods are presented in Figures 1 and 2. For the US, we look at the following periods: 1924-1929, 1993-2000 and 2004-2007. All three seem to be suggestive of a positive relationship between credit growth and rises in top income shares. Of course, the economy was in full expansion during the bulk of these periods, so that an alternative view might simply be that general economic expansion generates these co-movements. There are only a limited number of examples from other countries that support the idea that income concentration drives credit. Sweden from 2003-2008 might also be consistent with the US case between 2004 and 2007.

Despite these examples, there are many cases that do not fit the RKR frameworks. In Japan in the 1980s, credit growth clearly rises in advance of top income shares. On the other hand, top income shares started rising in 1995 in Japan while credit growth languished. In Sweden, sharp rises in top incomes followed, rather than led credit growth in the 1980s. Again, in Sweden, top income shares continued to grow in the aftermath of the banking and real estate bust in 1991 while credit fell. In Australia, credit growth was unrelated to top income shares in the 1970s. Top income shares follow rather than lead credit growth in the late 1980s in Australia. The UK shows similar patterns that do not fit the story that rising income concentration endogenously generates a credit boom.

4.2 Econometric Models of Credit Growth

We next investigate the determinants of changes in credit in a single equation regression framework. Formally we run regressions of the following form:

$$\Delta \ln(\text{Credit}_{it}) = \Delta X'_{it-1} \beta + \mu_i + \delta_t + \varepsilon_{it}$$

The dependent variable in our models is the growth in the (log) level of real credit. The $k \times 1$ vector X includes the share of income of the top 1%, the log of Real GDP, an index of investment relative to the price level, the log of the ratio of M2 relative to the price level, and a short term nominal interest rate. We also include the lagged value of the change in log real credit growth. In annual regressions, we lag all of our explanatory variables by one year, as the equation indicates, to deal with simultaneity issues. The time period for the estimation sample runs from 1920 up to 2008.⁷ Some data are available for credit for the period prior to World War I but only for a very limited set of countries and years. We begin with 1920 and omit the years of World War II. We are highly constrained by the availability of information on top incomes. These data become more widely available over time which makes the sample bulge in terms of “ N ” as time proceeds. To deal with this potential compositional issue, we present results from different sub-periods that have greater balance in terms of country coverage.

Our proxy for income concentration and income inequality is the share of total income earned by the top 1% of individuals or households or tax units. This variable has been collected from personal tax returns for each country by various teams of researchers following the methodology of Piketty (2001). The data set we use for the post-World War II period was downloaded from “World Top Incomes Database” (<http://mond.parisschoolofeconomics.eu/topincomes/>). The income data themselves are constructed as the ratio of the top income earners’ incomes divided by the total amount of national income recorded. Income here includes labor, business and capital income before taxes and transfers. In a limited number of cases, realized capital gains are included in the income concept. There are some issues of comparability over time and across space in these top incomes data. In a few cases, as

⁷ The sample in column 1 of Table 3 includes 14 countries for the following years: Australia (1931-1939 and 1972-2008), Canada (1937-1939 and 1958-2001), Denmark (1982-2006), France (1924-1938 and 1961-2007), Germany (1928-1937 and 1962-1999), Italy (1976-2005), Japan (1920-1939 and 1959-2006), Netherlands (1923-1939 and 1962-2000), Norway (1974-2008), Spain (1983-2008), Sweden (1921, 1936 and 1952-2008), Switzerland (1950-1996), United Kingdom (1960-2008), United States (1921-1939 and 1950-2008).

mentioned, capital gains are included in the income concept. The definition of the tax unit also varies by country and over time. Some issues in comparability over time due to changes in tax laws have also been highlighted by Piketty and Saez who provided the web-based interface for the World Top Incomes Database. Despite these caveats, we follow Roine, Vlachos and Waldenström (2009) and pool these data as a reasonable first pass at the key relationships.⁸

4.3 Credit Growth Determinants: Evidence from Five Year Periods

Our first results rely on information from five-year periods as in Roine et. al. Annual volatility is significant in the data especially in the top income shares. Also this is an attempt to look at medium term trend relationships which is the horizon on which the RKR frameworks seem to be focused. An observation in this sample is thus the cumulative change over the previous five years for each variable. We restrict the sample to the endpoint of each quinquennium, and period dummies are included for each five year period to control for shocks. In these models, variables in the vector X are for the same five-year period.

Table 2 presents four OLS regressions where the dependent variable is the cumulative growth over the previous five-years in the log of real credit. Column 1 includes only the cumulative percentage point change in the top income share as well as country fixed effects, time dummies, and growth in credit from the previous five-year period. While top income growth has a positive marginal effect, it is only significant at the 80 percent level of confidence. Column 2 adds the cumulative change in the log of real GDP over the previous five years to the model in column 1. Income growth is strongly related to credit growth. Again, we fail to reject the null hypothesis that top incomes have no association with the growth in credit. In column 3, we interact these two key variables. The reason is that the RKR hypothesis suggests that when total incomes and inequality are rising a credit boom arises in the bid to smooth consumption by the lower deciles. We find no evidence of such an impact with the interaction effect. Finally, column 4 controls for other macroeconomic fundamentals including the short-term nominal interest rate, the ratio of investment

⁸ We note that income inequality measures based on the Gini Coefficient or comparisons between the 90th and the 10th or the 50th percentiles are different from the top incomes variable which captures only one part of the distribution. For checking the robustness of our results, we also investigated, but do not report results using the top 5% and the top 10% of earners.

to GDP and the (log) real money supply (M2). The only variable that has a statistically significant relationship after inclusion of these controls is the growth of GDP.

From Table 2, the evidence is that the primary determinant of credit growth is real incomes. When GDP grows at a pace above its average, credit also rises above its average growth rate. We decisively reject any relationship between credit growth and changes in the share of income accruing to the top 1%. In unreported regressions, we found that no relationship existed between credit growth and the share of the top 0.01%, the top 5%, or the top 10%.

4.4 Credit Growth Determinants: Annual Evidence

Table 3 examines the relationship between credit growth and these determinants using annual data. Year-to-year variation provides another perspective and a robustness check on our results from Table 2. Once again, the annual data provide no evidence that growth in top income shares are a significant determinant of credit growth. In none of the specifications presented, or in the many which we leave unreported, do top incomes have any statistically significant relationship with credit growth. On the other hand, all specifications show a strong relationship between annual changes in GDP and credit growth.

Table 3 also presents evidence from three different subsamples. The first three columns are for the 1920-2008 period, the fourth column is limited to the years between 1950 and 2008, and the fifth column is for 1972-2008 when international capital flows began to rise rapidly. The first three columns add progressively more explanatory variables to the long-run sample. Echoing results in Table 2, GDP growth is highly significant and positively related to credit growth. Figure 3 presents this relationship in a scatter plot. This figure plots the relationship between residuals from regressions of credit growth and income growth that condition on country and time fixed effects and all variables included in column 4 of Table 3. Here one gets a sense of how tight the conditional relationship is between changes in top incomes and changes in income growth. Given the standard error of the estimated relationship, there is less than a 1 in 1000 chance that this relationship is statistically indistinguishable from zero, while the 95% confidence band is 0.25 to 1.03.

Table 3 shows that changes in the short-term nominal interest rate are also a significant determinant of credit growth. When short-term interest rates fall, credit growth rises. This result is

also robust to using ex post real interest rates. The relation between interest rates and credit seems to be consistent with the Borio and White story that low interest rates reflecting benign inflationary expectations can provide an environment favorable to creating a credit boom. It is also consistent with a simpler story emphasizing the role of loose monetary policy in fueling a credit boom. This result, together with the relationship between credit and income growth resoundingly rejects any role for income concentration.

The negative relationship between interest rates and credit might also be consistent with a “savings glut” type of argument. An increase in the supply of savings drives interest rates down perhaps due to rising savings abroad or in the upper deciles. If the upper deciles were responsible for this rise, then it might be consistent with RKR. If so, then the top income share and interest rates should be strongly negatively correlated. There is no strong evidence for this notion. When we omit interest rates from column 1, the coefficient on top incomes rises to 0.0046 from 0.0042. The standard error hardly changes either.

As a robustness check, three other variables were included in Table 3: money growth, changes in the rate of investment relative to GDP and changes in the current account to GDP ratio. Mendoza and Terrones (2008) found that a rise in the current account deficit accompanied credit booms, but their sample included many emerging markets as well as leading countries. Our sample is limited to a subsample of the most developed countries. Here current account deficits have no significant relationship with credit growth.

A long literature on credit booms argues that technological breakthroughs and displacements drive investment and these need to be financed with credit (Fisher 1933, Kindleberger 1978, Minsky 1986). After controlling for the business cycle and the interest rate, we find no convincing evidence that higher investment is associated with credit growth. Money supply growth is also not associated with credit growth. This result is consistent with Schularick and Taylor (forthcoming) who observe a dwindling correlation between growth in the money supply and bank lending since World War II. According to them, this reflected financial innovation allowing banks to increase leverage and rely less on deposits for their funding. Overall, low interest rates and strong economic growth seem to be the most robust determinants of credit growth.

None of the econometric models we have deployed can reject the null hypothesis that top income shares have no relationship with changes in credit. Our cross-country evidence is also

inconsistent with Kumhof and Rancière who argued that rises in inequality could give rise to credit booms and financial crises. The results in Table 1 show a high probability of a banking crisis after credit growth rises, but since top income growth is not a determinant of credit growth, income concentration is not associated with banking crises. Indeed, unreported regressions that include growth in top incomes in regressions like those of Table 1 show that income inequality is not a significant determinant of banking crises in our sample.

5. Historical Evidence & Discussion

Recent literature has made the novel claim that income inequality in the US played a big role in driving the credit boom from 2002 and in the 1920s. Our cross country empirical evidence shows that there is no statistical relationship between income concentration and credit booms. In fact, historical evidence from the episodes in the US which are used to motivate RKR shows that income concentration may be only coincidental with these credit boom and bust episodes. Here we provide some anecdotal evidence in support of our findings above.

For the 1920s, time series for the US show that the share of income earned by the top 1% increased from 15% in 1922 to 18.42% in 1929. Research based on top-incomes, as well as early work by Williamson and Lindert (1980), identify this as a period of rising income inequality. However, this rise in income inequality does not seem to be associated with any stagnation in real wages for the working class. Indeed annual income of nonfarm employees rose grew at an average of 1.89 percent or a total of 23 percent between 1919 and 1929. In addition, rises in the standard of living must have been even greater. Leisure increased in this period as the standard work week fell to 48 hours by 1929. The introduction of electrification, better indoor plumbing and a host of new consumer durables including automobiles, radio, washing machines and refrigerators, made home production more efficient and leisure more enjoyable.

At the same time, as credit allegedly “boomed” in the 1920s, the economy grew largely above trend from 1923 up to 1929. Olney (1999) reports that consumer non-mortgage debt to income rose from 5.6% to 9.3% between 1923 and 1929. The ratio of individual and non-corporate debt to nominal income (this includes farm production credit, farm mortgages, non-farm mortgages, commercial, financial and consumer credit) increased between 1923 and 1929 from 63% (\$53.7

billion in credit and \$85.1 billion in gross output) to 70.7% (\$72.9 billion in credit versus \$103.1 billion in gross output).⁹ The bulk of the rise in credit was attributable to non-farm mortgage lending which rose by \$13.3 billion. Consumer credit doubled from \$3.2 billion to \$6.4 billion in nominal terms between 1923 and 1930 but the amounts were relatively small relative to total income. Eichengreen and Mitchener (2004) cite competition amongst lenders, monetary stability and improved housing quality as drivers of the housing boom.

Eugene N. White (2009) delves further into the US housing boom of 1920-1926 investigating demand, monetary policy, lending standards, mortgage securitization, risk-taking and supervision. Out of sample predictions of a demand for housing model account for a large portion of the rise in housing starts since demand had been repressed during the war, GDP was growing fast and interest rates were low. Still, the portion of housing starts left unexplained by this model is equal to 1/2 to 2/3 of annual average housing starts from 1900-1917. White concludes that monetary policy was somewhat too loose, but other important features of the boom were supply-side financial innovations including mortgage securitization, weakened supervision of lenders, lower lending standards and even corruption. White recounts that Florida politicians were “bought” by developers in order to achieve low supervision and easy charters. This version of the story is driven by the supply side rather than the demand side as in RKR. White is also emphatic that the housing bust did not impair the financial system or generate a banking crisis in 1926 nor was the housing boom and bust directly responsible for the banking crises of 1930, 1931-32 and 1933. There is little evidence that housing, which accounts for the bulk of the rise in household borrowing in the 1920s, was affected by the level of earnings inequality in the period.

Consumer credit was closely related to the rise of new big-ticket consumer durables in the 1920s. However, the rise in consumer credit also arguably came from supply side innovations rather than from a household demand to maintain consumption in the face of stagnant incomes. The historical record is not consistent with the RKR hypothesis in this regard. Olney (1989) demonstrates that a major driver of consumer credit for automobiles, an important component of overall consumer credit, was pressure from automobile manufacturers. Producers wanted retailers to carry the costs of the winter inventory buildup so that production could be smoothed and average costs could be lower. The solution was the advent of finance companies including GMAC that

⁹ While nominal credit aggregates peaked in 1929, the ratio of debt to GDP continued to grow up to 1933 due to the fact that nominal GDP declined by almost 50% between 1929 and 1933.

innovated methods of extending credit to consumers. Often they did so by purchasing installment contracts from dealers. The original intent was not to “bolster retail sales but to finance dealer’s wholesale inventory” (Olney, 1989).

On this basis it is very hard to generalize the RKR view even to the 1920s in the United States. There is simply no evidence of a political conspiracy to increase home-owning in the 1920s in the USA in order to win votes. Nor is there any evidence that the demand for credit rose in order to make up for lost income and lagging consumption. Eichengreen and Mitchener (2004) dissect the 1920s credit boom and suggest that the view of Borio and White is largely consistent with the cross-country experience. In the US, but also in other nations, competition in the financial sector and accommodative monetary policy in the 1920s drove credit up during the boom and likely fed back into the economic cycle. The stock market crash in 1929 and the subsequent economic slump can be attributed to many causes but amongst them are the large financial imbalances that built up over the 1920s which were cataclysmic for household balance sheets and aggregate demand due to the subsequent deflation (Mishkin, 1978 Olney, 1999).

The experience of other countries that experienced major credit booms and subsequent financial crises is also not obviously consistent with RKR. Sweden had a credit boom following the liberalization of their financial sector in the 1980s. The other key factors leading to its systemic banking crisis in 1991 were capital inflows and an overvalued exchange rate (Jonung et. al. 2009). Australia had a massive housing and credit boom in the late 1980s as well. Factors cited there were rapid financial de-regulation, excessive risk taking in the corporate sector and irrational speculation in the housing market (Macfarlane, 2006). Japan’s real estate boom in the 1980s was also fueled by expansionary monetary policy under pressure from the U.S. to weaken the Yen at the Plaza Accord of 1985 (Funibashi, 1989). Mullbauer and Murphy (1990) argued that the UK credit and consumption boom of the late 1980s was enabled by financial liberalization that allowed consumers to cash in on housing equity. An alternative story for the UK in the 1980s is that households anticipated an increase in lifetime labor incomes (Attanasio and Weber, 1994). None of these cases or points of view associates the credit and asset price booms with rising income inequality. Most of them witnessed declining interest rates or accommodative monetary policy, financial liberalization and economic expansion.

6. Conclusions

Our paper looks for empirical evidence that might corroborate Rajan (2010) and Kumhof and Rancière (2011). Both attributed the US subprime crisis to rising inequality, redistributive government housing policy and a credit boom. Using data from a panel of 14 countries for over 120 years, we find strong evidence linking credit booms to banking crises, but no evidence that rising income concentration was a significant determinant of credit booms. Narrative evidence on the US experience in the 1920s, and that of other countries in more recent decades, casts further doubt on the role of rising inequality.

We do find significant evidence that rising real income and falling interest rates are important determinants of credit booms. This evidence is more consistent with the alternative story of Borio and White (2003) attributing credit booms and crises in the past three decades to the Great Moderation which created a benign environment conducive to rising credit. It is also consistent with other empirical work that covers the period 1960-2002 (Mendoza and Terrones, 2008). The negative and significant relationship of short-term interest rates and credit growth may also be consistent with the story of for example Taylor (2009) or Meltzer (2010) who attribute the U.S. housing boom to expansionary policy by the Federal Reserve in the early 2000s in an attempt to prevent perceived deflation. Moreover, housing booms and busts in other countries did not reflect redistributive housing policy. In the period before the Great Moderation they occurred during episodes of expansionary monetary policy. Regardless of whether the Borio and White story or a simpler monetary policy story is the true explanation for credit booms that lead to financial crises it now seems fairly clear from our examination of the data that neither have much to do with rising income inequality.

References

- Attanasio, O. Weber, G. 1994. UK Consumption Boom of the Late 1980s: Aggregate Implications of Microeconomic Evidence. *Economic Journal* 104 (427) pp. 1269-1302.
- Atkinson, A.B., Piketty, T., Saez, E. 2011 Top Incomes in the Long Run of History. *Journal of Economic Literature*. 49 (1) 3-71.
- Atkinson, T., Morelli, S. 2010. Inequality and Banking Crises: A First Look. mimeo, Oxford University.
- Barlevy, G. Tsiddon, D. 2006 Earnings inequality and the business cycle. *European Economic Review*. 50 pp. 55-89.
- Bernanke, B and Gertler M, 2000. Monetary Policy and Asset Price Volatility. *American Economic Review*. May
- Bordo, M. and Jeanne, O. 2002. Monetary Policy and Asset prices: Does Benign Neglect Make Sense? *International Finance*.
- Bordo, M. Eichengreen, B, Klingebiel, D., Martinez-Peria, M.S. 2001. "Is the Crisis Problem Growing More Severe?" *Economic Policy*, 16 (32), 53-82.
- Borio, C., Lowe, P. 2002. Asset prices, financial and monetary stability: exploring the nexus. BIS Working Papers. no 114.
- Borio, C. and White, W. W. 2003. Whither monetary and financial stability? The implications of evolving policy regimes. in *Monetary Policy and Uncertainty: Adapting to a Changing Economy: A Symposium sponsored by the Federal Reserve Bank of Kansas City, Jackson Hole, Wyoming, 28-30 August*, pp 131-211.
- Eichengreen, B., Mitchener, K. 2004. The Great Depression as a Credit Boom Gone Wrong." *Research in Economic History*. 22.
- Fisher, I. 1933 "The Debt Deflation Theory of Great Depressions." *Econometrica* 1 (4). Pp. 337-357.
- Funibashi, Y. 1988. *Managing the Dollar: From the Plaza to the Louvre*. Institute for International Economics. Washington DC

- Goldin, C. Katz, L. 2008. *The Race Between Education and Technology*. Belknap Press.
- Jonung, L. Kiander, J and Vartia, P. 2009. *The Great Financial Crisis in Finland and Sweden: The Nordic Experience of Financial Liberalization*. London. Edward Elgar
- Kalecki, M. 1990-1997. *Collected Works of Michal Kalecki*, 7 vols, ed. J. Osiatynski. Oxford: Clarendon.
- Kindleberger, C. 1996. *Manias , Panics and Crashes: A History of Financial Crises*. New York: John Wiley
- Kumhof, M., Rancière, R. 2011. *Inequality, Leverage and Crises* IMF working Paper 10/268.
- McCarthy, N, Poole, K and Rosenthal, H. 2006. *Polarized America: The Dance of Ideology and Unequal Risks*. Cambridge MA ; MIT Press
- Macfarlane, I. 2006. *The Search for Stability*. Australian Broadcasting Company Boyer Lectures.
- Mendoza, E.G. Terrones, M. 2008. *An Anatomy of Credit Booms: Evidence from the Macro Aggregates and Micro Data*. NBER Working Paper 14049.
- Minsky, H. 1986. *Stabilizing an Unstable Economy*. New Haven: Yale University Press
- Mishkin, F., 1978. The household balance sheet and the Great Depression. *Journal of Economic History* 38, 918--937.
- Mullbauer, J., Murphy, A. 1990. Is the UK balance of payments sustainable?. *Economic Policy*. 11 pp. 345-83.
- Olney, M. 1999. Avoiding Default: The Role of Credit in the Consumption Collapse Of 1930. *Quarterly Journal of Economics*, 114 (1), 319-335.
- Olney, M. 1989. Credit as a Production-Smoothing Device: The Case of Automobiles, 1913-1938 *Journal of Economic History*. 49 (2). Pp. 377-391.
- Piketty, Thomas, 2001a. *Les hauts revenus en France au 20ème siècle*. Grasset, Paris.
- Rajan, R. 2010. *“Fault Lines”* Princeton University Press, Princeton, NJ.

Roine, J. Vlachos, J., Waldenström, D. 2009. The long-run determinants of inequality: What can we learn from top income data? *Journal of Public Economics* 93. 974-988.

Schularick, M. and Taylor A.M. forthcoming. Credit Booms Gone Bust Monetary Policy, Leverage Cycles and Financial Crises, 1870–2008. *American Economic Review*.

Stiglitz, J 2009. “Joseph Stiglitz and Why Inequality is at the Root of the Recession” Next Left website. January 9 2009.

Taylor, J.B., 2009. *Getting Off Track: How Government Actions and Intervention Caused, Prolonged and Worsened the Financial Crisis*. Stanford. Hoover Institution Press.

White, Eugene N. 2009. Lessons from the Great American Real Estate Boom and Bust of the 1920s. NBER Working Paper No. 15573.

Williamson, J.G., Lindert, P.H. 1980. *American Inequality: A Macroeconomic History*. Academic Press Inc, New York.

Table 1 Relationship Between Banking Crises and the Growth in Credit, 1889-2008

	OLS, 1889-2008	Logit 1889-2008	Logit 1889-1913	Logit 1920-1940	Logit 1950-2008
$\Delta \ln (\text{Loans/CPI})_{t-1}$	-0.145** [0.060]	-4.087** [1.861]	-6.211 [5.533]	-1.164 [2.718]	-5.833* [3.410]
$\Delta \ln (\text{Loans/CPI})_{t-2}$	0.167** [0.076]	3.710** [1.445]	0.399 [3.828]	5.329** [2.197]	2.98 [2.918]
$\Delta \ln (\text{Loans/CPI})_{t-3}$	0.146** [0.053]	3.316** [1.546]	5.344 [6.613]	3.12 [2.521]	2.99 [2.691]
$\Delta \ln (\text{Loans/CPI})_{t-4}$	0.12 [0.096]	2.854** [1.432]	-4.719 [3.836]	3.828* [1.980]	4.685* [2.511]
$\Delta \ln (\text{Loans/CPI})_{t-5}$	0.199*** [0.062]	3.448** [1.468]	7.851* [4.077]	-0.115 [2.241]	3.729 [2.306]
Year fixed effects	yes	yes	yes	yes	yes
Country fixed effects	yes	no	no	no	no
Observations	1,490	609	67	167	375

Notes: Dependent variable is a dummy equal to 1 when a banking crisis occurred. The symbol Δ signifies an annual difference. Year fixed effects are included in all specifications. Robust standard errors are reported in parentheses. A number of years are omitted in the logit specifications since no countries experienced a banking crisis in some years.

Table 2 Credit Growth and the Share of Income of the Top 1 Percent, Five year periods between 1920 and 2008

	(1)	(2)	(3)	(4)
Δ Top 1% Share	0.024 [0.019]	-0.003 [0.018]	0.011 [0.024]	0.015 [0.027]
$\Delta \ln(\text{GDP/Capita})$	---	1.248*** [0.392]	1.204** [0.442]	1.215** [0.437]
Δ Top 1% Share x $\Delta \ln(\text{GDP/Capita})$	---	---	-0.294 [0.178]	-0.294 [0.193]
Δ Short term nominal interest rate	---	---	---	-0.238 [0.487]
Δ (investment/GDP)	---	---	---	-0.216 [0.563]
$\Delta \ln$ (Money/Price Level)	---	---	---	-0.07 [0.313]
$\Delta \ln(\text{Credit/Price Level})_{t-1}$	0.152 [0.144]	0.139 [0.101]	0.142 [0.102]	0.133 [0.097]
Total Observations	115	115	115	115
R-Squared	0.507	0.583	0.595	0.596
Number of Countries	14	14	14	14

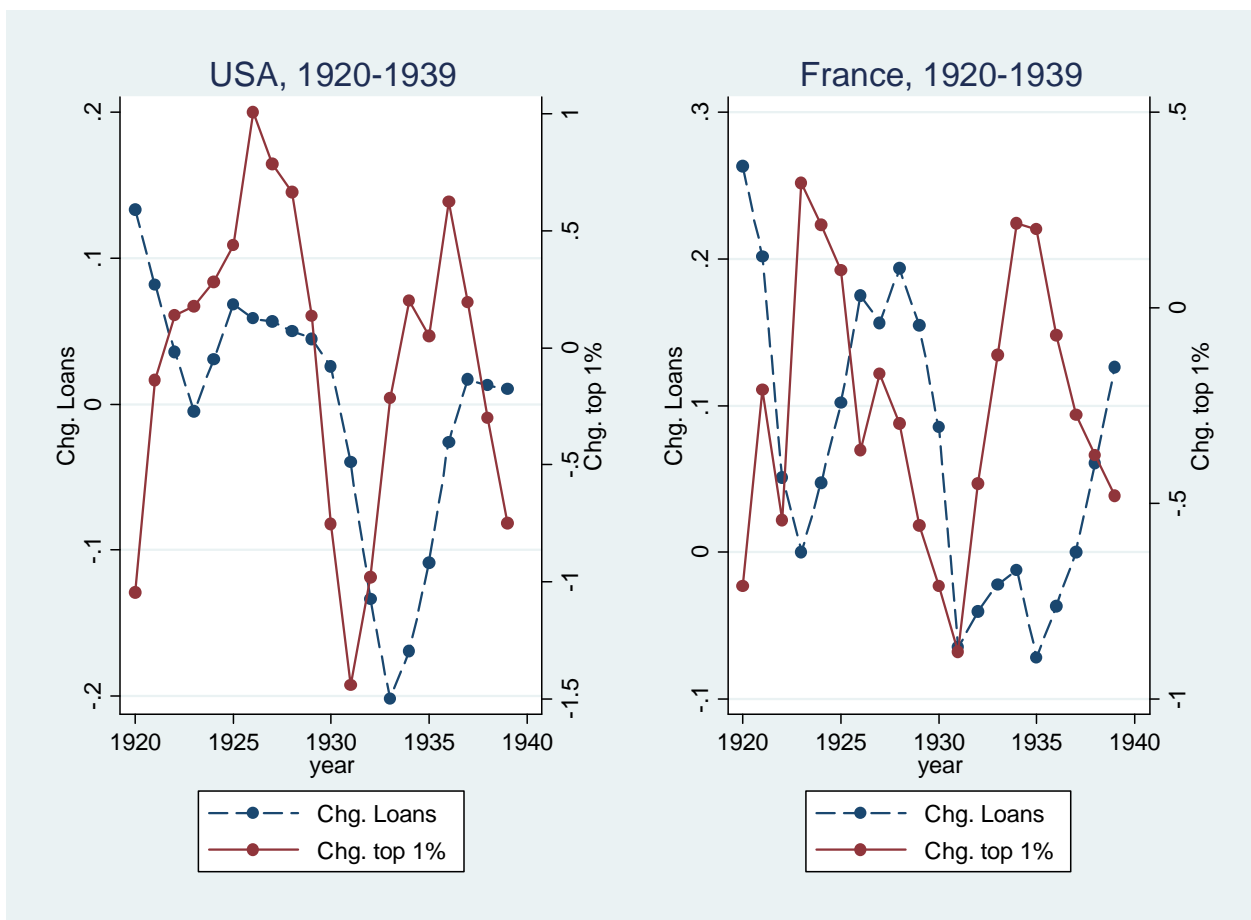
Notes: Dependent variable is the cumulative change in the share of total income earned by the top percentile over the five years leading up to and including 1920, 1925, 1930, 1935, 1940, 1955, 1960, 1965, ..., 2005. Estimation is by OLS. Quinquennial period dummies are included in all regressions as are country fixed effects and period fixed effects. Robust standard errors are reported throughout. *** p<0.01, ** p<0.05, * p<0.1

Table 3 Explaining Changes in the level of ln(Credit/Price Level), Annual Data, 1920-2008

	(1) 1920-2008	(2) 1920-2008	(3) 1920-2008	(4) 1950-2008	(5) 1972-2008
Δ Top 1% Share $t-1$	0.004 [0.004]	0.004 [0.004]	0.005 [0.004]	0.0002 [0.003]	-0.001 [0.002]
Δ ln(GDP) $t-1$	0.353*** [0.102]	0.330*** [0.100]	0.450*** [0.145]	0.628*** [0.169]	0.643*** [0.164]
Δ Short term nominal interest rate $t-1$	-0.460*** [0.131]	-0.427*** [0.108]	-0.374*** [0.123]	-0.477*** [0.142]	-0.447*** [0.133]
Δ ln (Money/Price Level) $t-1$	---	0.064 [0.105]	0.056 [0.104]	0.115 [0.104]	0.074 [0.093]
Δ (investment/GDP) $t-1$	---	---	-0.475 [0.274]	-0.066 [0.224]	0.043 [0.355]
Δ (Current Account/GDP) $t-1$	---	---	---	---	0.001 [0.002]
Δ ln(Credit/Price Level) $t-1$	0.300*** [0.049]	0.284*** [0.069]	0.284*** [0.068]	0.255*** [0.059]	0.293*** [0.091]
Total Observations	672	672	672	576	421
R-Squared	0.394	0.396	0.401	0.372	0.363
Number of Countries	14	14	14	14	14

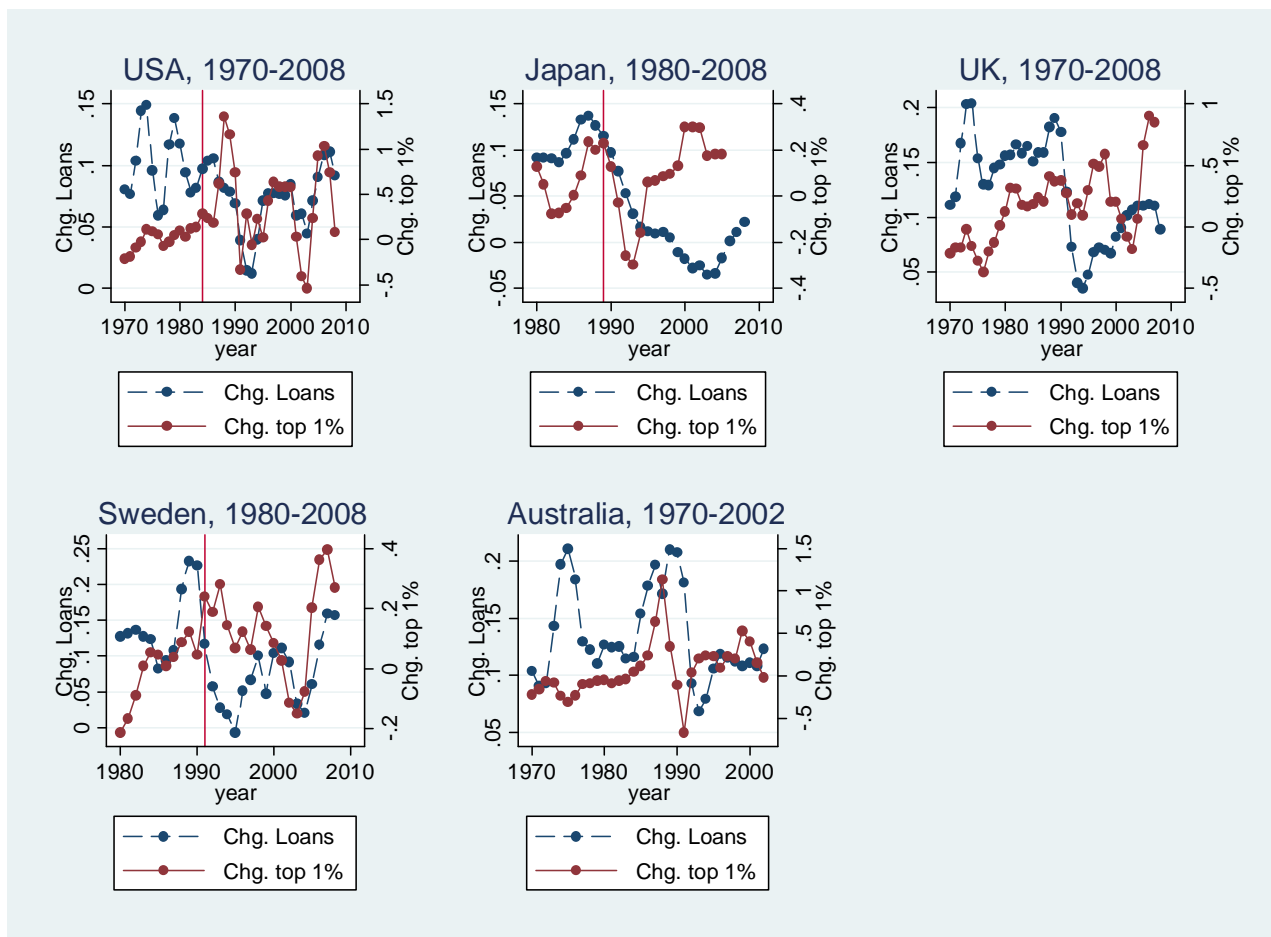
Notes: Dependent variable is the annual change in ln(Credit/Price Level). Estimation is by OLS. Year dummies are included in all regressions as are country fixed effects. Robust standard errors are reported throughout. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Figure 1 Changes in Bank Loans and Top Income Shares for France and the USA in the Interwar Period



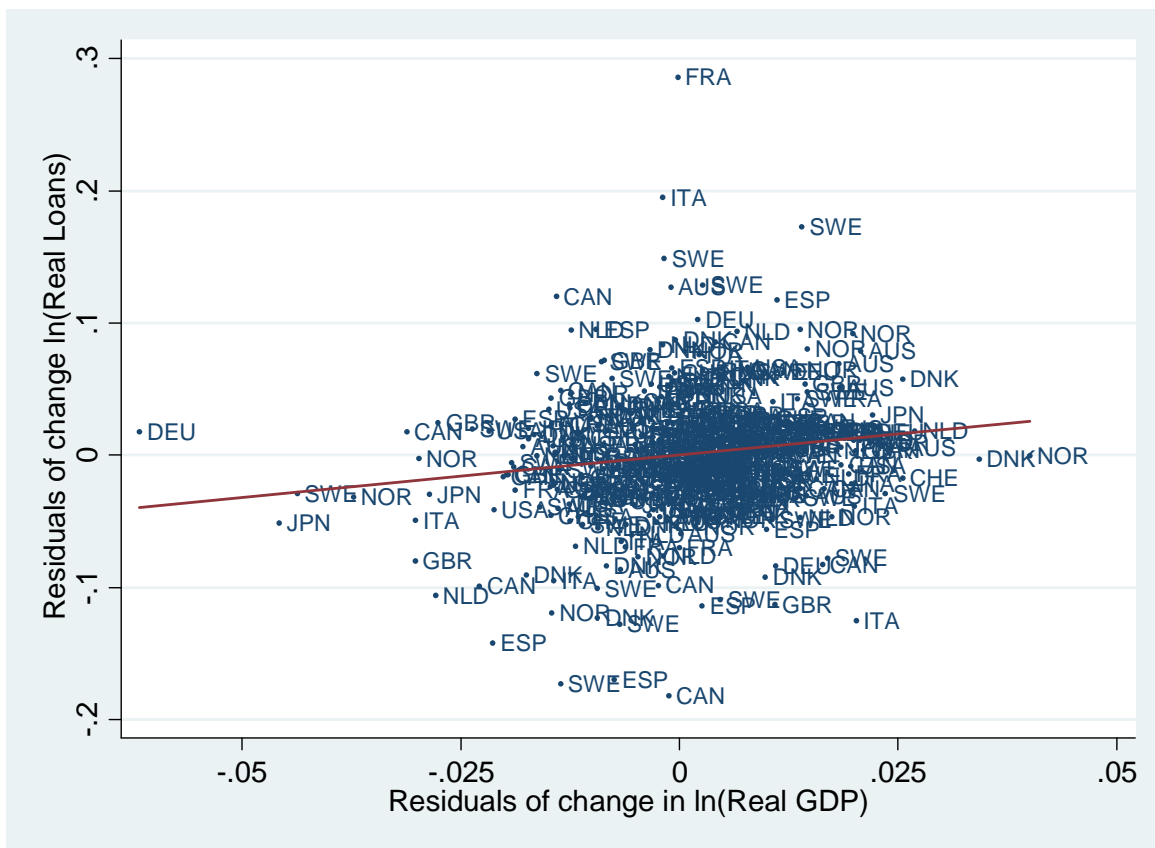
Notes: Chg. Loans is the annual change in the logarithm of the ratio of total bank loans to the domestic consumer price index. Chg. Top 1% is the annual change in the percentage share of the top 1% of income earners.

Figure 2 Changes in Bank Loans and Top Income Shares for Five Countries over Time



Notes: Chg. Loans is the annual change in the logarithm of the ratio of total bank loans to the domestic consumer price index. Chg. Top 1% is the annual change in the percentage share of the top 1% of income earners.

Figure 3 Change in Loans versus the Change in GDP, 14 countries, 1972-2008



Notes: Residuals of change in ln(Real Loans) on the y-axis is the residual from a panel OLS regression of the annual change of the log of real loans on country fixed effects, year fixed effects, and lagged changes in the share of income of the top 1 percent of tax units, short-term interest rates, log of the real money supply, log of the ratio of investment to GDP, and the ratio of the current account to GDP. The change in ln(Real GDP) is residual from a similar regression using contemporaneous changes in the log of real GDP and contemporaneous explanatory variables. This scatter plot therefore corresponds to the regression in column 5 of Table 3.