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Real Output Costs of Financial Crises: A Loss Distribution Approach*

Daniel Kapp[†] Marco Vega[‡]

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Abstract

We study cross-country GDP losses due to financial crises in terms of frequency (number of loss events per period) and severity (loss per occurrence). We perform the Loss Distribution Approach (LDA) to estimate a multi-country aggregate GDP loss probability density function and the percentiles associated to extreme events due to financial crises.

We find that output losses arising from financial crises are strongly heterogeneous and that currency crises lead to smaller output losses than debt and banking crises.

Extreme global financial crises episodes, occurring with a one percent probability every five years, lead to losses between 2.95% and 4.54% of world GDP.

JEL classification: C15, G01, G17, G22, G32.

Keywords: Financial Crisis, Severity, Frequency, Loss Distribution Approach.

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1 Introduction

Financial crises have played a quintessential role after the collapse of the Bretton Woods system of fixed exchange rates. Episodes like the Latin American debt crises in the 80's, the 1987 Black Monday, the 1992-1993 ERM crisis, the 1994-1995 Tequila crisis, the 1997-1998 South East Asian meltdown, the 1998-1999 Brazilian and Russian crisis, the 2000-2001 Turkish crisis, the 2001 Argentine crisis and the 2007-2009 global financial crisis all resemble disaster events, just like hurricanes or earthquakes.

Like catastrophic events, financial crises can be characterized by frequency and severity. In fact, analysis made in the financial crises literature often refers to terms such as frequency and severity (see Bordo et.al., 2001). The insurance and operational risk theory and practice offer toolkits to analyze frequency and severity of losses as well as aggregate losses due to catastrophic or operational risk events. Given that the losses to country economies from financial crises (in terms of GDP drop or forgone GDP) are similar to catastrophic losses, we can apply the loss distribution approach (LDA) familiar in the actuarial literature to analyze frequency and severity of losses and thus study rare events and their probabilities (see Panjer, 2006; Shevchenko, 2011). To the best of our knowledge, there is no attempt to quantify the frequency and severity of financial crises using the LDA.

The advantage of the LDA is that we can analyze frequency of events and severity in a separate fashion and then combine them to obtain a global loss probability density function. In the study of GDP losses arising from financial crisis, it is of interest to study frequency and severity on their own as shown in Bordo et.al. (2001).

The LDA allows us to estimate a multi-country aggregate GDP loss distribution and thus estimate conditional losses in the event of a financial crisis occurring in the near future. We can also determine the probability of rare economic disasters as defined for example in Barro (2006). In contrast to Barro (2006) however, we do not make any assumptions

on the channel through which crises occur (e.g. catastrophic events like earthquakes or regular/cyclical disaster events).

In terms of methodology, we use the financial crises database of Laeven and Valencia (2008) to date financial crises across 170 countries from 1970 onwards. The number of such events over a predetermined period is called the frequency of events. We then estimate output losses per financial crises event with a number of methods. Afterwards, we aggregate country output losses across events over the period. Since a particular crisis event can generate output losses over various years, we set the span of analysis to be five years. Finally, we compound frequency and severity to generate a loss probability density of aggregate losses that allows us to report standard risk measures.

We find that output losses after financial crises are strongly heterogeneous and a large number of countries never recovered their pre-crisis growth rates or trends. Also, we show that currency crises lead to smaller output losses than debt and banking crises, while the largest losses are found after debt crises. The presence of a debt crisis also exacerbates any of the other two forms of crises, while the presence of a currency crisis in the wake of a debt or banking crisis diminishes output losses through faster recovery. Banking and debt crises alone are found to be more severe than twin crises consisting of banking and currency crises or debt and currency crises.

The LDA approach leads us to conclude that mean worldwide costs of financial crises within periods of 5 years are in the range of 0.52% to 0.81% of 2005 world GDP. Extreme crises episodes, occurring with a one percent probability, can lead to losses between 2.95% and 4.45% of world GDP.

The analysis of losses produced in the paper can be a useful tool in discussions about the existence of insurance against the risk of financial crises at the aggregate level. For example, Caballero (2003) proposes such an arrangement for emerging market economies.

In what follows, we will provide a short literature review and discuss the possibilities at hand to calculate output costs of financial crises. In section 3, we introduce the method-

ology of crisis identification and loss calculation, while the Loss Distribution Approach is explained in Section 4. Results are presented in Section 5. Section 6 discusses a potential form of international insurance and Section 7 concludes.

2 The costs of financial crises

In order to quantify the costs of financial crises, we need to define a metric. Studies on this subject use a varied set of cost measures. Costs have been estimated as fiscal costs, costs to the stock market, and output costs. Among studies analyzing the same kinds of losses, methodologies also differ. In addition, one of the main obstacles to measuring losses caused by financial crises is how to identify losses only due to financial crisis and not arising from other contemporaneous factors.

2.1 Fiscal costs, costs to the stock market and output losses

In an attempt to quantify the costs of banking crises to the economy, Hoggarth et.al. (2002) consider direct resolution costs as well as broader welfare costs to the economy, approximated by output losses. They argue that resolution costs are a rather limited proxy for costs incurred through banking crises, as they may reflect a transfer of income from taxpayers to banks rather than costs imposed to the economy as a whole. The authors reason that there could be a positive correlation between fiscal costs and output losses if crises are systemic. On the other hand, if fiscal costs are a good proxy for effective crisis resolution, higher spending on crisis resolution should lead to lower output losses during a crisis period. No clear statistical relationship between fiscal costs and length of crises is found, while output losses and the length of crises do depict a clear positive correlation.

Frydl (1999) presents a comparative analysis of prior banking crises studies. As one of the reasons for the non-significant statistical relation between resolution costs and crises

length, he claims that resolution costs usually measure fiscal costs of banking crises, which are often subject to various errors and do not incorporate many indirect costs to the government or the economy. Having dismissed fiscal costs as a reliable indicator for crisis severity, Boyd et.al. (2000), use the discounted value of corporate returns to measure the impact of crises. Under the condition that corporate profits represent a relatively constant fraction of total output, a decline in the real values of stock prices at the onset of a crisis in percentage terms is approximately equal to the decline in the present discounted value of total output.

With regards to the impact and depth of currency crises, possible measures to be considered (in addition to output losses) are the loss of international reserves and the depreciation of the real exchange rate (Kaminsky and Reinhart, 1999). However, the most popular method is to proxy costs to the economy with GDP losses, given that economic growth is a natural final performance indicator. This is the approach we follow in this paper.

We identify two strains of literature approximating real GDP losses due to financial crises. The first uses a dummy variable approach to estimate growth losses over samples of countries, studies like Demirguc-Kunt et.al. (2006), Gupta et.al. (2007), Hanna and Huang (2002) and Barro (2001) follow this path. The second approach proxies welfare losses by comparing GDP during a crisis period with some estimate of potential output. Hoggarth et.al. (2002) is one representative study of this latter approach. It estimates potential output assuming that output would have grown at the same constant rate based on its past performance. Various studies, such as Bordo et.al. (2001), Aziz et.al. (2000), Frydl (1999), Boyd et.al. (2000, 2002) and Cecchetti et.al. (2009) calculate output losses from banking crises in a similar fashion, even though their trend estimates are based on differing pre-crisis windows, methods, definitions about onsets, ends and durations of crisis episodes.

A main criticism of the dummy variable approach is that it can only identify average

magnitudes of growth contractions associated with crises for all countries. It therefore does not seem to be well suited for highly heterogeneous losses. Output costs calculated through cross-section or panel data regressions are usually found to be lower than losses calculated based on output gap estimations.

In a comparative analysis, Angkinand (2008) suggests that the output gap approach is more appropriate than the dummy variable approach in capturing the output costs of crises. This is so because the individual output costs across crises vary substantially. Our loss estimation methodology therefore follows the output gap approach.

2.2 Output gap calculations

Determining the length of a crisis

The literature does not offer a unanimously agreed method to date the beginning of financial crises. Banking crisis start dates are usually defined through a mixture of quantitative and qualitative criteria. Caprio et.al. (1996) rely on the assessment of finance professionals. Including and expanding on this approach, Demirguc-Kunt and Detragiache (1998) compile five studies of banking crises' starting dates. Other studies identifying banking crises' dates are Dziobek et.al. (2008), Kaminsky and Reinhart (1999), and Lindgren et.al. (1996).

The onset of a currency crisis is generally defined as a situation whereby a sufficiently large depreciation of the domestic currency occurs¹, often accompanied by a loss in international reserves, while a debt crisis takes place in the event of a country defaulting or renegotiating all or parts of its private debt. The most recent financial crises compilation, widely used in previous empirical studies, stems from Laeven and Valencia (2008) and comprises banking, currency, and debt crises over the period 1970 to 2008.

¹Sometimes an additional criterion of increased speed of depreciation as compared to some prior time window is introduced.

To date the end of a crisis episode, one possibility is to define the end date based on “expert” opinions or on the “consensus” view from various studies. An alternative is to define the end of a crisis endogenously once a country returns to a certain pre-crisis *growth rate* or recovers its potential output *growth path*.

Studies determining the end of a crisis based on the recovery of the average *growth rate* of a pre-crisis window are, among others, Bordo et.al. (2001) and Aziz et.al. (2000). Authors such as Boyd et.al. (2002) argue that summing up deviations from an estimated trend up to the point at which the observed *growth rate* returns to its pre-crisis average is problematic since output typically remains well below its pre-crisis absolute output trend once the growth rate has recovered.

Cecchetti et.al. (2009) avoid calculating a counterfactual and define the end of a crisis as the point in time when real GDP has reached its absolute pre-crisis level. This method is problematic in at least two ways. First, the method does not take opportunity costs of foregone output growth into account. Second, the method implies that a crisis is only counted as such if output growth actually turns negative during the crisis year. It can be argued, however, that a financial crisis has negative effects without having caused an actual recession, e.g. through a transitory or permanent slowdown in growth. Moreover, since potential growth rates vary across countries, dating the end of a recession by reaching its pre-crisis level of real GDP can lead to an underestimation of total losses incurred.

Estimation of a counterfactual

To be able to measure output losses during crisis periods according to the methods described above, it is necessary to compare actual output with its trend level. There is a handful of approaches to estimate trend GDP levels. They differ mainly by the pre-crisis time window chosen, which in turn depends on the assumption about financial crises either following economic booms or a slowdown in economic activity.

Hoggarth et.al. (2002) assume that output would have grown at a constant rate based on

past growth performance and extrapolate linear three and ten-year trends, while Bordo et.al. (2001) use five-year pre-crisis trends. Instead, Frydl (1999) apply ten year pre-crisis periods and Boyd et.al. (2000) extrapolate linear pre-crisis growth trends.

A large part of the heterogeneity in the magnitude of output losses in the study of crises stems from the calculation of trend output. Studies like Kindleberger (1978), Borio et.al. (1994) and Logan (2001) find that banking crises follow economic booms. In this case, a trend estimated over a short period prior to a crisis would overestimate potential output and lead to an overestimation of crisis length and depth. On the other hand, studies like Kaminsky and Reinhart (1999) or Gorton (1988) find that banking crises are often preceded by a slowdown of economic activity, in which case losses would be understated.

If one assumes that pre-crisis growth deviates (in either direction) from the long-term potential output growth path, one option is to increase the pre-crisis trend calculation period in order to capture mostly “normal” years. An alternative is to exclude a certain period prior to the onset of a crisis. Last, the Hodrick-Prescot (HP) filter can be applied to diminish the influence of booms or recessions on the potential growth path. While Aziz et.al. (2000) base potential output on the average output of the three years prior to crises, Bordo et.al. (2001) use five-year pre-crisis growth rates and Hoggarth et.al. (2002) calculate potential output trends based on ten-, three-, and one-year pre-crises growth rates; for a comparative study see Angkinand (2008).

Once we establish the counterfactual, we can estimate total output losses by adding up the difference between actual and potential output over the duration of the respective crisis.

Even though the general concept is agreed upon across the studies mentioned above, several methodological issues remain debated. Identification of crises accompanied by output losses varies among studies. While some authors include a crisis if output is below its trend or if output growth is negative during the crisis year, other studies include crises even though output is above its trend in the crisis year, given that output is below trend in

the subsequent year (Angkinand, 2008). Further issues arise in the case of multiple crises per country within short periods of time. In the case that output has not yet recovered from one crisis at the point of outbreak of a following crisis, some studies choose to sum losses of subsequent crises and report a single loss, while others divide losses across crises or simply choose to exclude countries with multiple crises during the sample period.

2.3 Identification of causality

To assess the direction of causality between economic growth and banking crises, Hoggarth et.al. (2002) compare a sample of 29 countries experiencing banking crises with neighboring countries which did not face banking crises at the same time. The hypothesis is that “the movement in output relative to trend during the crisis period would have been, in the absence of a banking crisis, the same or similar to the movement in the pairing country” (Hoggarth et.al., 2002). Their analysis hints at the point that output losses are in most cases caused by banking crises and come as unforeseen events.

Bordo et.al. (2001) find, across all countries and crisis periods considered, that recessions with crises are more severe than recessions without them. These results are in line with various studies such as Frydl (1999).

3 Methodology

3.1 Crisis identification

We use the financial crises dataset of Laeven and Valencia (2008). Currency crises, banking crises, and debt crises are identified over the period 1970 to 2008. Laeven and Valencia (2008) identify banking crises on the basis of a number of quantitative and subjective criteria, such as a large number of defaults and a high quantity of non-performing loans. The starting year of a currency crisis is identified by building on an

approach developed in Frankel and Rose (1996). Sovereign debt crises are reported in the case of sovereign defaults to private lending as well as in a year of debt rescheduling. The number of currency crises peaked during the early 1980's and the early 1990's with around 30 currency crises per year, while banking crises have in general been less frequent and peaked during the early nineties. The number of debt crises per year has been decreasing since the mid-1980's and debt crises have nearly ceased to exist until recently, see Bicaba et.al. (2011).

We identify the starting date of a financial crisis as the year of outbreak of any one of the three types of crises. Real GDP data is taken from the World Economic Outlook database and spans the time period from 1960 to 2010.

To calculate output losses caused by crises, the first step is to define whether a crisis has an impact on the economy. In the case where output is compared to a counterfactual, we identify a crisis if output in the crisis year is below its trend level (trend output estimations are discussed below). In an alternative calculation of crisis losses, no counterfactual is established and a crisis accompanied by output losses is considered as such if output growth is negative during the crisis year.

For countries with multiple crises during the sample period it is possible that a crisis occurs before the economy has recovered from a previous crisis. In this case, we assign subsequent losses to the later crisis date, establishing a new counterfactual. This method is problematic, though other alternatives suffer from larger errors. Allocating output losses from subsequent crises to the first crisis would largely overstate output losses in various cases.

3.2 Output losses

We estimate three kinds of potential output trends and propose several cutoff points to determine the end of a crisis. In short, there is no perfect method to estimate an ob-

jective output trend. Every method presented shows both advantages and disadvantages depending on the assumptions about the mechanics of financial crises.

Following the literature, we estimate potential GDP after the onset of a crisis in three ways. We estimate an HP trend, where potential output during a crisis episode is based on the average HP growth rate of the ten and three year pre-crisis periods. In addition, we use average growth rates from three year and ten year pre-crisis time windows. We compare the losses against trend output levels to absolute losses (for episodes with negative growth) without considering opportunity costs. In total we have thirteen possible ways to measure output losses as well as their severity and frequency.

Output losses are calculated as the difference between actual real GDP and its trend level. A graphical example is given in Figure 1, depicting output during the Ecuadorian banking crisis, which shows that losses depend not only on the definition of the counterfactual but largely on establishing an end-point of a crisis.

According to the time output needs to recover its pre-crisis *level*, the effects of the crisis lasted two years and led to an output loss of 10% of GDP. If crisis length is calculated until real GDP growth reached its pre-crisis *growth rate*, output losses occurred over a period of two and three years and led to output losses of 10.1% and 22.4% of GDP, calculated against a three and ten-year pre-crisis growth trend respectively.

As shown in Figures 1 and 2, this calculation most likely still does not account for the total output loss caused by the 1998 Ecuadorian banking crisis. The period of output loss increases to six and seven years, and output losses accumulate to 38.1% and 51.4% of GDP respectively if losses are estimated until the level of real GDP recovers its three and ten-year pre-crisis *trend*. As illustrated in Figures 1 and 2, output losses seem to be underestimated if crisis recovery is defined as being completed at the point where the pre-crisis *growth rate* or the pre-crisis *level* of real GDP are recovered.

As mentioned above, we estimate losses using various trends and various cut-off points to determine the end of a crisis. In total, 13 loss estimations are presented.

Figure 1: Ecuador 1998

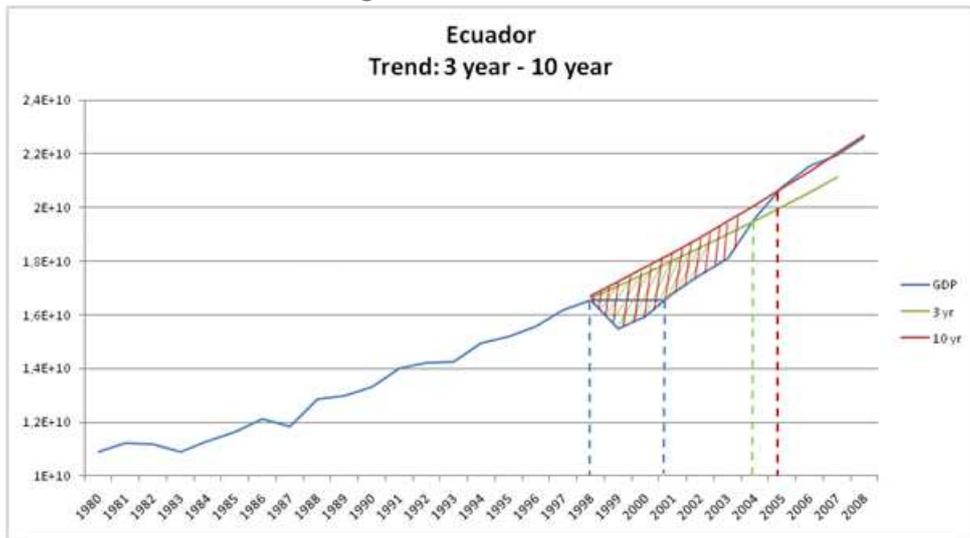


Figure 2: Ecuador 1998



We distinguish three definitions of recovery from a crisis. According to the first definition, a crisis ends once real output has reached the level of its counterfactual. The alternative is that recovery is completed once the average pre-crisis growth rate is resumed. As some countries never recover according to these definitions, accumulated losses against the counterfactuals based on linear three and ten year trends, based on simple averages of pre-crisis growth, are considered over maximum periods of five and ten years, while the losses against trends based on the HP filter are allowed to accumulate over periods of maximum ten years. Note that an alternative potential remedy that we do not pursue

here in cases where we have an infinite sum of losses is to perform present-value GDP losses by discounting future losses appropriately. In the absence of a counterfactual, a crisis is supposed to be ended once output reaches its absolute pre-crisis level of real GDP.

4 The Loss Distribution Approach

The estimated output losses across countries obtained in the previous section allow us to study the frequency and severity of losses. In the analysis of a financial crisis hitting the world economy, two usual questions appear: a) what is the frequency of financial crisis? and b) given a financial crisis, how severe is it? The frequency of financial crises is the number of such events over a specific period of time. Since a financial crisis duration spans generally more than a year, we choose a five-year reference period. The severity of a financial crisis is the amount of output loss incurred in each crisis episode.

Given the frequency and severity of losses occurring within 5-year periods, we can use the LDA common in the insurance and operational risks literature. Chernobai et.al. (2011) and Cruz (2004) provide detailed descriptions of the workings of the LDA. We estimate two possible parametric distribution functions commonly used to describe the frequency of events n_t over a period of time t . We also estimate a set of six severity probability density functions² for events $z_{t,i}$. As opposed to the standard one year period, our t represents periods of 5 years. This is because GDP losses due to financial crises consider losses over more than one year. The index i tracks each event within the period of analysis t .

During a 5-year period, global losses are given by the sum of each loss event i across countries in the sample.

²The set is composed of the gamma, exponential, generalized extreme value, generalized pareto, log normal and weibull density functions.

$$S_t = \sum_{i=1}^{n_t} z_{t,i} \quad (1)$$

The random variable n_t takes discrete values while the variables $z_{t,i}$ are non-negative (positive losses), real valued quantities. The aggregated loss S depends on the realization of the discrete random variable (n) and the continuous random variable (z_i). Therefore, the aggregation S is itself a random variable whose distribution has to be determined by convolution methods (Panjer, 2006; Shevchenko, 2011).

Specifically, the frequency of loss events (n) has a probability distribution denoted by $p_n = Pr(N = n)$ while the loss severity z has a density distribution and cumulative distribution functions denoted by f_z and F_z , respectively. According to Panjer (2006), the cumulative density function of S is defined as:

$$\begin{aligned} F(S) &= Pr(\omega \leq S) \\ &= \sum_{n=0}^{\infty} p_n Pr(\omega | N = n) \\ &= \sum_{n=0}^{\infty} p_n F_Z^n(S) \end{aligned} \quad (2)$$

Where $F_Z^n(S)$ is the n -fold convolution of the cumulative density function of S . A simple way to estimate $F(s)$ is via Monte Carlo simulations. First we draw n from p_n and then we draw z from f_z as many times as indicated by n . Last, we sum up the z draws.

5 Results

In total, we observe 62 debt crises, 122 banking crises, and 196 currency crises. As some of these crises in effect form twin crises, we examine a total of 340 crises episodes. Depending on the method applied, we find between 110 and 219 contractionary crises

episodes. While around 210 contractionary crises episodes are identified against HP trends, only 110 crises were accompanied by negative growth rates during the crisis year.

Average losses are higher compared to previous studies, this is so because econometric studies usually estimate losses over a sample comprising recessionary crises as well as crisis followed by a rebound in GDP, while we only consider crises leading to output losses (against a trend or absolute). From a total of 196 currency crises in our sample, between 90 and 120 crises (depending on the calculation method applied) have led to a loss of output compared to some measure of potential output, while 63 crises were accompanied by negative output growth. This result is in line with previous studies such as Gupta et.al. (2007), who find that about 60 percent of currency crises lead to output contractions while the rest were accompanied by output expansions.

The analysis that follows concentrates on a benchmark loss classification group that relies on the HP filter: losses until recovering average 10 year HP filtered GDP growth rates (HP10perc), the level of a 10 year HP filtered GDP trend (HP10trend), 3 year HP filtered GDP growth rates (HP3perc), and the level of a 3 year HP filtered GDP trend (HP3trend).

5.1 The frequency of financial crises

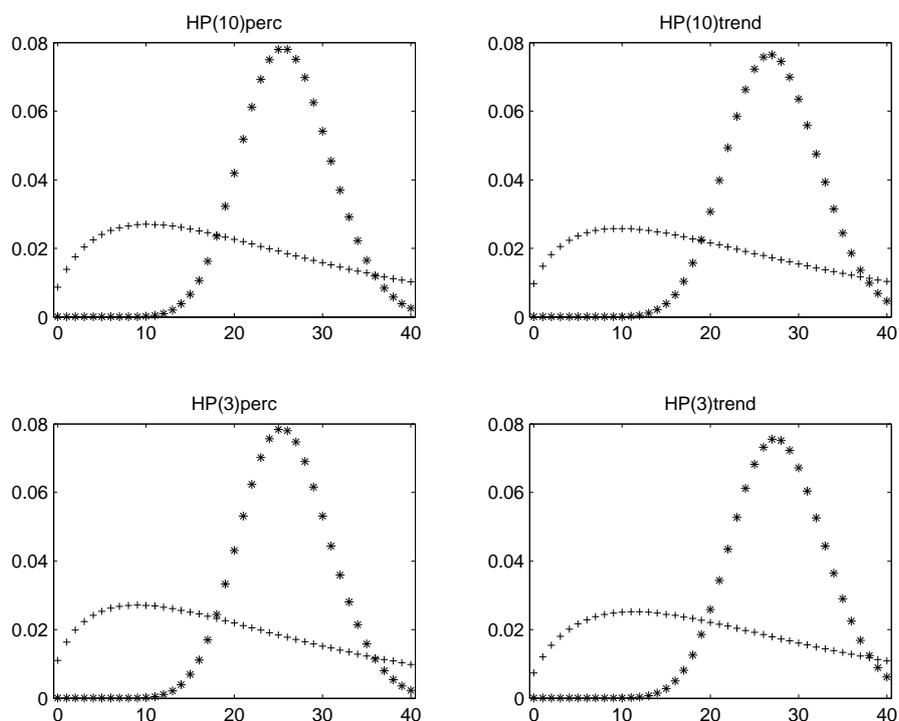
In order to carry out the LDA, we first estimate distribution functions for the frequency of losses from financial crises. We assume two commonly used distributions such as the Poisson and the Negative Binomial Distribution. The key parameter in the Poisson Distribution is λ which is also the mean and variance of the number of losses. This is the drawback of the Poisson fit, the data in our benchmark case (the four HP trend counterfactuals) have numbers of crisis events with variances of about sixteen times the mean.³ Therefore, the two parameter Negative Binomial distribution is a more flexible

³The number of crisis events for the other loss classifications have a variance to mean ratio of more than 13, except for the case where opportunity costs are not considered (ABS).

way to accommodate our data.

In Figure 3 we depict the estimated distributions for our benchmark loss classification. The Negative Binomial distribution has a lower mode but allows a more extreme number of losses relative to the Poisson distribution. The probability that the data come from the Negative Binomial is higher for all cases (the benchmark and all types of loss classifications). In all cases, the mean number of crises occurrence over five-year periods ranges from 20 to 30.

Figure 3: Estimated distributions for the frequency of losses



Note: +++: Negative Binomial, ***: Poisson. Horizontal axis measures the number of financial crisis in the world over a typical five-year period

5.2 The severity of financial crises

The average accumulated loss caused by financial crises varies from 9% of real GDP to 15% of real GDP if output losses are accumulated against trends based on HP filtered data (Table 1). In total, depending on the loss measure applied, 186 to 219 crises episodes

are observed. Average losses are the largest when calculated against a ten year trend and if losses are considered until the level of trend output has been recovered over a maximum time span of 10 years.

Median losses lie between 4.9% and 7.15% of initial GDP and the most severe crises destroy up to three years of economic output.⁴

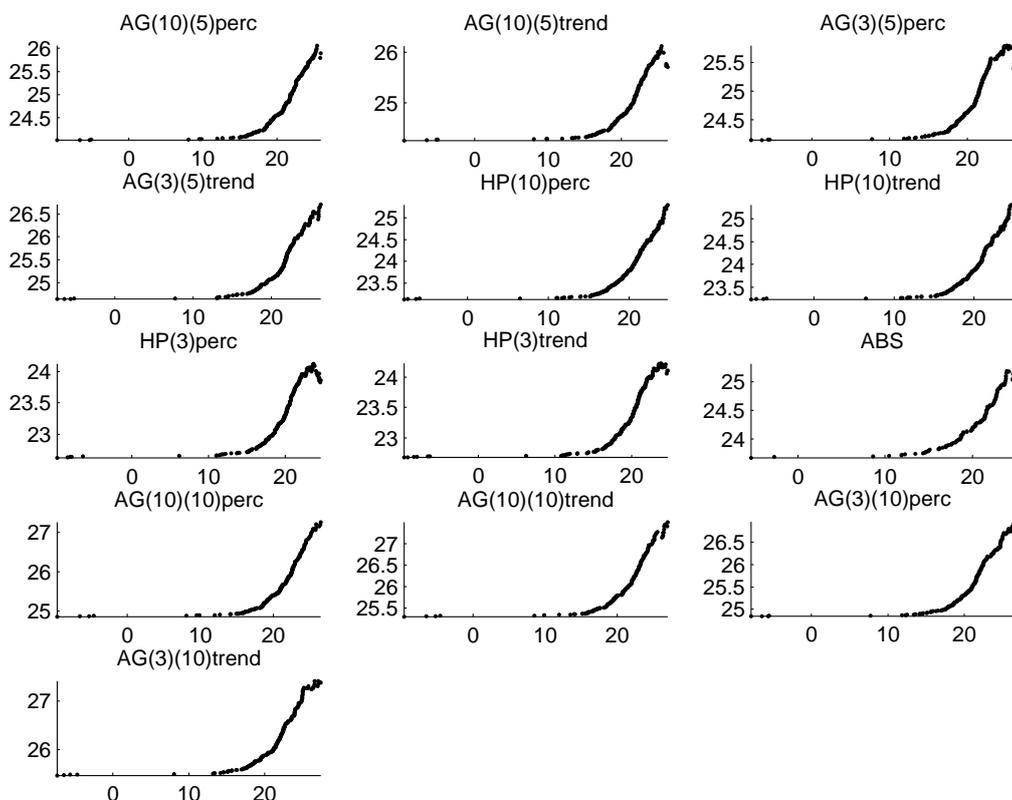
As can be seen in Tables 7 through 14 in Appendix B, alternative output losses calculated are very heterogeneous and large average percentage losses are driven by few especially severe crises events. Of these most severe events, potential output has not been reached again within a period of ten years. This result is in line with Furceri et.al. (2011), who find that the growth rate after debt crises eight years after the onset is still suppressed by nearly 10 percentage points. Amongst the largest losses observed are those of several Asian countries, namely Indonesia in 1998 and Thailand in 1997, both experiencing severe losses in the wake of the Asian crisis.

Severity data due to financial crisis have a key feature, it is extreme valued. In figure 4 we see that the mean-excess-over-threshold plots⁵ have positive slope at the right extreme of losses. Figure 4 shows that the long-tailed nature of severity data is generated by specific types of crises. Currency and twin currency-banking crises produce the more extreme type of losses.

⁴Crises' percentage losses are calculated as the sum of the difference between real output observed and potential output after the onset of a crisis until its end, divided by the real output in the year of crisis onset. Losses are usually larger if calculated on the basis of other trend estimations as presented in the appendix.

⁵The sample mean excess plot is defined by: $mep = \frac{\sum_i^n (X_i - u) I_{(X_i > u)}}{\sum_i^n I_{(X_i > u)}}$, with $u > 0$ and $I_{(\cdot)}$ and indicator function.

Figure 4: Mean excess over threshold for severity data. Both axes are in log scale of losses



We fit the severity data with six possible probability density functions using the maximum likelihood estimator for the corresponding parameters. The six distributions are Gamma, Exponential, Generalized Extreme Value (GEV), Generalized Pareto, Log-normal, and Weibull. Some distributions, like the GEV, fit the right end tail better, while others have a better fit over the entire range of data. Our benchmark choice is the Weibull distribution because it maintains a better fit over the entire range of data for all severity classifications.

Types of financial crises

Consistent with studies like Reinhart and Rogoff (2009), we find that currency crises lead to smaller output losses than debt and banking crises.⁶ About 70 of the 122 banking

⁶Distributions of these three types of crises' losses, calculated as losses accrued over a maximum

crises in our sample lead to output losses. We can not conclude that banking and currency crises are generally preceded by high or low periods of growth as we do not observe a general dominance of losses calculated against three year pre-crisis trends as opposed to losses against ten year pre-crisis trends.

We find that average output losses after debt crises⁷ are 9% higher than losses after banking crises. The median debt crisis is accompanied by output losses of 11.7%, three percentage points larger than the median banking crisis. A large share of debt crises has, however, been accompanied by banking crises. Of the 62 debt crises in our sample, 36 have led to periods of negative growth. Of these 36 episodes, 26 have been accompanied by banking crises with mean losses about 30% higher than if debt crises occur alone. Currency crises⁸ incur smaller losses than banking or debt crises of 15% in the mean and 5% in the median. This result is in line with Furceri et.al. (2011) that concluded that debt crises tend to be more detrimental than banking and currency crises.

Banking and debt crises alone are more severe than twin crises consisting of banking and currency crises or debt and currency crises. While twin crises between banking or debt and currency crises mostly lead to larger growth reductions in the very short run than banking or debt crises alone, long term losses are often found to be smaller. A possible explanation could be that a depreciation in the wake of a twin crisis (including a currency crisis) allows for a competitiveness gain which is not present when a banking crisis occurs alone and allows for a faster recovery. Twin crises consisting of debt and banking crises consequently incur the highest losses in our sample.⁹

Severity by region

The highest losses after financial crises are experienced in Asia with average losses ranging

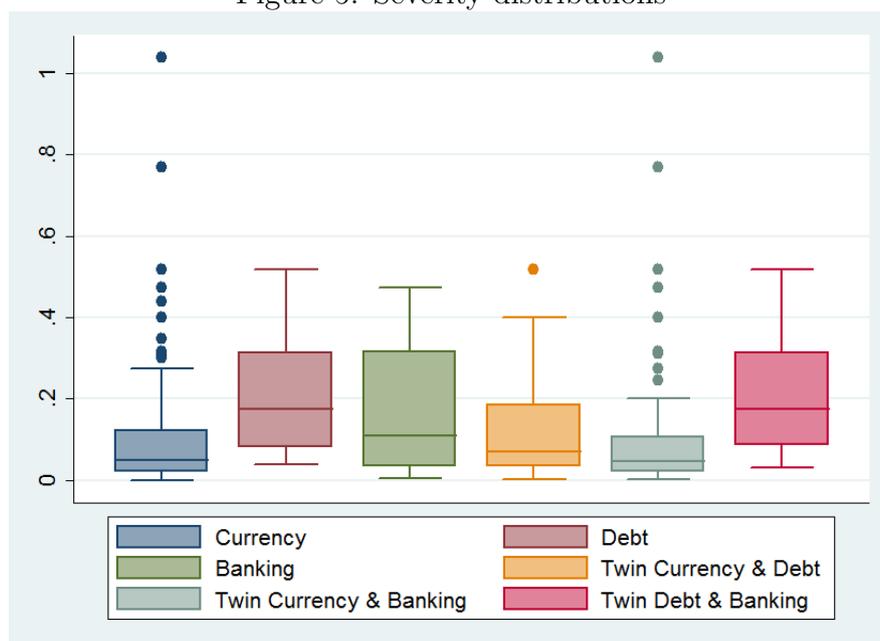
period of five years until the average growth rate of a ten year pre-crisis period is recovered, are depicted in figure 6, while descriptive statistics of all calculation methods are provided in tables 8 to 14.

⁷23% of initial GDP if recovery of the average HP *growth rate* ten years before a crisis is considered (HP10 perc)

⁸According to HP10 perc

⁹For detailed descriptive statistics of other loss measures, please see tables 8 to 14.

Figure 5: Severity distributions



Source: Authors' calculations. For graphical reasons, distributions are cut at 1.5

from 9.8% to 21.8% of initial GDP.¹⁰ The second highest losses are observed in Europe, followed by Latin America and Africa. The highest frequency of financial crises is, however, observed in Africa. Europe and Asia are mostly struck by currency and banking crises, while Africa and Latin America have suffered from all three types of crises to nearly equal degrees. In a joint analysis, debt crises were observed to be more severe than currency or banking crises. This result holds for all regions except for Asia, where banking crises lead to the most severe losses. Since a high share of currency and banking crises reported are in effect twin crises, it is however difficult to disentangle losses from both types separately. Furceri et.al. (2011) confirm our results by controlling for the other types of crises.

Severity by income groups

In order to compare losses across different income groups, we classify countries into low, middle, and high income categories.¹¹ On average, middle income countries experience the highest output losses (15%), followed by high and low income countries (12% and

¹⁰Calculated against counterfactuals based on average HP filtered growth rates

¹¹World Bank classification.

11% of initial GDP respectively, see Table 3). This observation holds for currency and banking crises, while losses from debt crises are almost exclusively observed in low and middle income countries.

The median crisis is more severe in high income than middle income countries. High average losses in middle income countries are driven by some extreme crisis events, such as Thailand in 1998, experiencing losses of 229%.¹² The median loss is 9% for high income countries, 7% for middle income countries, and 4% for low income countries. In total, middle income countries suffer more from financial crises than high income countries as they experience a larger number of crises per country over the period observed.

Severity of financial crises over time

As expected, financial crises were especially harmful during the 1990's and depict the lowest losses during the 1970's and the period after the year 2000. However, this does not necessarily mean that financial crises have become smaller in magnitude.

The severity of the 2007 financial crisis in the USA

Our main analysis considers a sample ending in 2005. However, we have enough information to estimate the severity of the 2007 financial crisis in the USA. We date the US banking crisis to late 2007 even though the main event, Lehman Brothers filing for Chapter 11 bankruptcy protection, did not take place until September 2008. Output trends are calculated as before with 2007 being the first year of output gap calculations. Data are taken from the IMF World Economic Outlook database until 2011. From 2012 until 2016, real GDP forecasts reported in the IMF database are used. Reliable output losses can therefore only be calculated over an interval of 5 years (2007-2011), larger loss windows are however reported for comparative reasons.

Figure A-5 in Appendix A shows real GDP for the USA from 1995 until 2016. Average real GDP growth in the ten years prior the financial crisis was 3.2 percent and 3.1 percent

¹²See Figure A-2 in Appendix A.

in the last three years before 2007. Real GDP growth slowed down to less than 2 percent in 2007 and finally turned negative during 2008 and 2009.

If the costs from the financial crisis are to be considered until the pre-crisis level of real GDP has been reached again and opportunity costs are ignored, the endpoint of the crisis is found to be in 2011 and costs remain modest at 5 percent of 2007 real GDP.

Turning towards our baseline calculations, calculated against a linear trend based on a pre-crisis time window of a HP filtered real GDP series, costs become larger and range between 15 and 48 percent of initial GDP if a 5 year time is considered and between 15 and 107 percent if costs are allowed to accumulate for up to ten years. The windows vary from a three year period, if the crisis is considered to be overcome once the pre-crisis average growth rate is recovered, to ten years in the case of defining the end of a crisis as recovery of the pre-crisis growth trend.

Baseline results over 5 year windows of around 30 percent of GDP are consistent with costs calculated in e.g. Chinn and Frieden (2011), who estimate costs of the 2011 financial crisis at 3.5 trillion USD, which is roughly 26 percent of 2007 real GDP.

Historically, the 2007 financial crisis ranks among the most severe crises episodes. A comparable percentage loss of real GDP during the last 40 years has only been experienced by some Asian countries in the wake of the 1998 Asian crisis.

5.3 The distribution of total losses

Given our choice of the frequency distribution and the severity probability density function, we perform Monte Carlo simulations to obtain a numerical probability density function (PDF) of total losses over five years. These PDFs are markedly skewed to the right. Table 5 summarizes the results. The 99.9 percentile of the total loss distribution ranges from 2.4 to 3.6 trillion of 2005 USD. Estimates of cumulative world GDP loss due to the 2008-2009 world financial crisis are 5 trillion (IMF, 2009). This means that the recent

financial crisis losses occur with very low probability. For the USA, Chinn and Frieden (2011) estimate a cumulative GDP loss for the USA to be about 3.5 trillions in 2005 USD.

6 Risk-sharing against financial crises

Can output losses from financial crises be diminished or crises prevented ex ante through an insurance scheme against rollover risk or capital flight? As financial crises are relatively rare events, one could imagine countries paying a certain amount of premium during financially stable times and in return having access to these funds during times of need.

Papers like Caballero (2003) and Cordella and Levy-Yeyati (2005); Cordella and Yeyati (2006) have proposed insurance schemes with global scope. The estimation of the stochastic properties of financial crisis losses at the global perspective in terms of frequency, severity and their global aggregation is a useful device in thinking about the proposed insurance schemes.

According to our estimates, the potential worldwide costs from financial crises over periods of 5 years in percentage terms of 2005 world GDP are presented in Table 6. Average costs of financial crises during a period a five years are relatively small and amount to less than one percent of 2005 world GDP. A period of extreme crisis events, occurring with a one percent probability, produces output costs of up to 4.54% of World GDP. For example, if there is a will to cover global losses up to the 99 percentile, the amount of insurance coverage is such percentile minus the median value.

Many of the debt crises included in the above calculations are in fact destabilizing confidence crises or liquidity crises, during which rollover costs of debt become excessively high, rendering an illiquid country insolvent. In a similar manner, liquidity risk is often the cause for the occurrence of banking crises. The IMF provides a de facto insurance (Cordella and Yeyati, 2006, see) for these cases in the form of standard IMF programs.

While the IMF does play the role of lender of last resort, it does so ex-post, after the country usually has already entered into financial turmoil.

As a type of national insurance against crises, nearly all emerging economies have accumulated large amounts of international reserves in order to possess a buffer against pro-cyclical international capital flows. Caballero (2003) argues that hedging the financial mechanism behind macroeconomic disasters is a problem of a magnitude larger than a single market can handle.

The moral hazard problem arising through a potential insurance scheme is addressed in various studies. Cordella and Levy-Yeyati (2005) examines to what degree the presence of a country insurance scheme affects the policymakers' incentives to undertake reforms. An important channel through which insurance can foster reforms can be identified: Insurance reduces the probability that deteriorating fundamentals evolve into large crises, which may enhance the expected political reforms and increase reform incentives.

Participation in a potential crisis insurance fund would therefore have to be subject to ex ante compliance with a number of clearly defined eligibility criteria, such as low budget deficits and a clear debt to GDP threshold. The potential insurance coverage must be forfeited as soon as the country does not fulfill all criteria. As stated in Cordella and Yeyati (2006), it would also be crucial to characterize and standardize the procedures followed after funds from the insurance facility have been accessed. In the optimal case, the existence of this insurance would incite fiscal discipline and, at the same time, provide liquidity if needed which in turn would lead to fewer crises.

Attractiveness of interest rate insurance implies that the insurance premium in tranquil times is lower than the costs through high rollover costs in turbulent times. In addition, the countries not experiencing liquidity problems have to be assured a benefit from the other country not entering into a crisis which exceeds the costs incurred through the insurance premium. Contrary to regular insurance, risk sharing depends on the contagion effects of financial crises from the country in financial turmoil to the liquid countries.

Feasibility and benefit analysis is a potentially attractive area for future research.

7 Conclusion

Through the use of the financial crises database of Laeven and Valencia (2008) to date financial crises, we characterize the heterogeneity of aggregate output losses. In our LDA setup, the number of crisis events over a certain period is called the frequency of events, while the different metrics of GDP losses describe the severity of such losses. We fit common parametric frequency and severity distributions to compound global GDP loss densities and to report standard risk measures.

In line with the existing literature, we find that output losses after financial crises are strongly heterogeneous and a large number of countries never recover their pre-crisis growth rates or trends. Loss distributions are skewed to the right, with average losses ranging between 9% and 15% of initial GDP.

Currency crises lead to smaller output losses than debt and banking crises, while the largest losses are found after debt crises. The presence of a debt crisis also exacerbates any of the other two forms of crises, while the presence of a currency crisis in the wake of a debt or banking crisis diminishes output losses through faster recovery. Banking and debt crises alone are found to be more severe than twin crises consisting of banking and currency crises or debt and currency crises.

We compare output costs from financial crises over world regions, country-income groups, and time. We find that Asia has suffered from the most severe financial crises, while Africa experiences the highest frequency of financial crises. Congruently, middle income countries experience the highest output losses, followed by high and low income countries. Financial crises are observed to have been especially harmful during the 1990's, while a global assessment of the severity of the recent 2008 financial crisis cannot yet be undertaken with our approach.

The LDA approach leads us to conclude that mean worldwide costs of financial crises within periods of 5 years are in the range of 0.5% to 0.7% of 2005 world GDP. Extreme crises episodes, occurring with a probability of one percent, can lead to losses in the range of 2.95% to 4.54% of world GDP.

There are some aspects that go beyond the scope of this paper. For example, the treatment of the appropriate time span for the LDA planning. We consider a five-year horizon but this is not necessarily the best from an optimal insurance design perspective. Also, all the calculations are based on aggregating GDP levels in comparable Dollar units. All calculations regarding percentage GDP losses are made after the LDA aggregation and relative to an initial GDP level. There is a possibility of exploring the LDA benchmark by allowing for the aggregation of percentage GDP losses directly. In this case, suitable GDP weights have to be considered. In addition, financial crisis distort long run growth in a number of cases, if so, losses in terms of GDP per capita instead of GDP levels can provide an adequate long run picture. Last, our loss approach is unconditional but the framework can be extended to incorporate conditioning factors in the stochastic processes that drive frequency and severity of GDP losses. We leave this issues open for future research.

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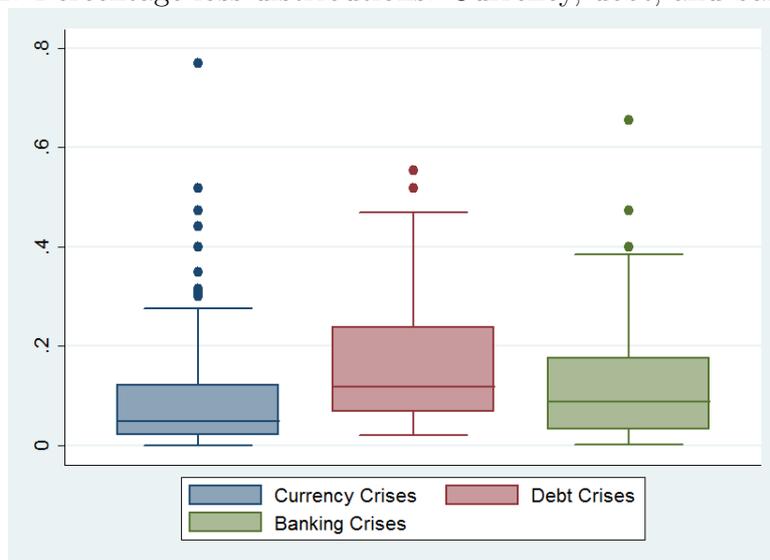
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Appendix A Figures

Figure A-1: Percentage loss distributions. Currency, debt, and banking crises



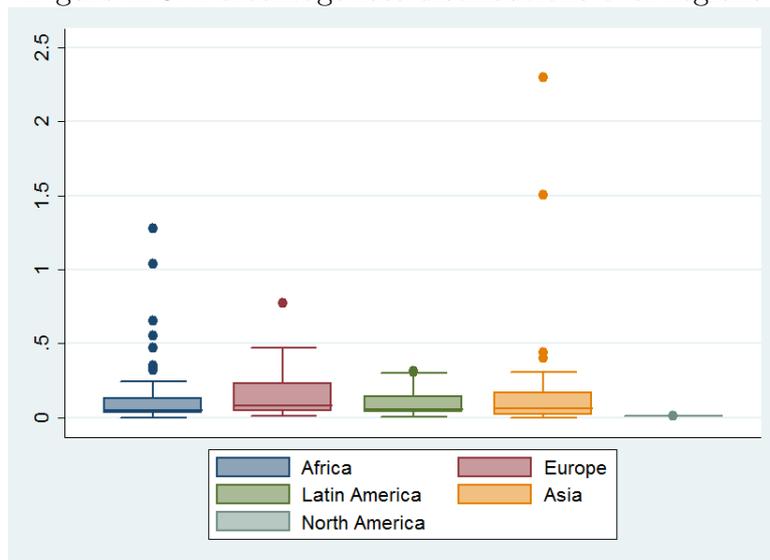
Source: Authors' calculations. Distributions are cut at 100 percent

Figure A-2: Percentage loss distributions over income groups



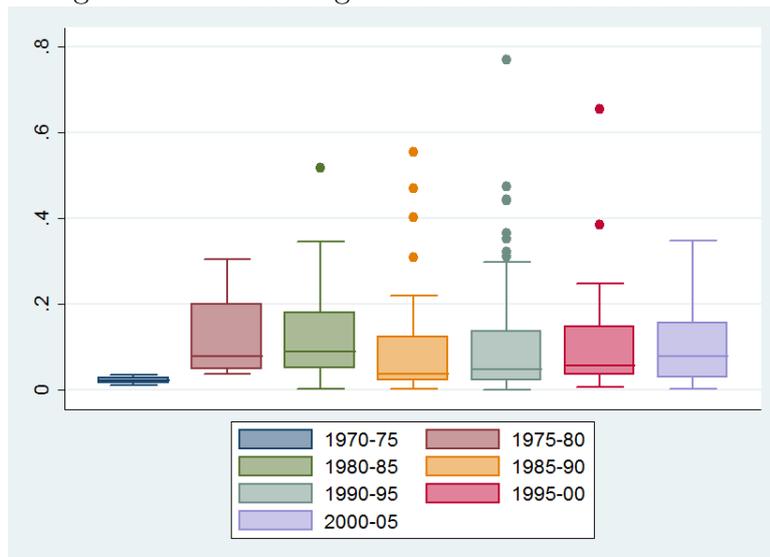
Source: Authors' calculations

Figure A-3: Percentage loss distributions over regions



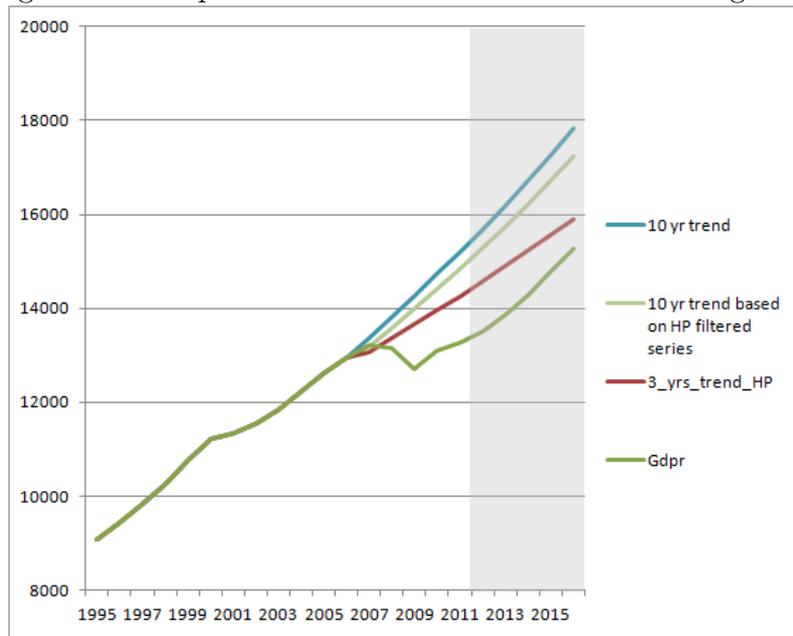
Source: Authors' calculations

Figure A-4: Percentage loss distributions over time



Source: Authors' calculations, Cut at 100 percent

Figure A-5: Impact of the 2007 financial crisis on US growth



Source: Authors' calculations

Appendix B Tables

Table 1: Severity of financial crises (as percentage of initial GDP)

Loss Measure	Obs.	Mean	Median	Std. Dev.	Min	Max
HP10 perc	204	0.14	0.06	0.24	0.0001	2.30
HP10 trend	214	0.15	0.07	0.25	0.0003	2.30
HP3 perc	203	0.09	0.05	0.17	0.0000	2.04
HP3 trend	219	0.10	0.06	0.12	0.0000	1.06

NOTE: Estimation of losses to recover average 10 year HP filtered GDP growth rates (HP10 perc), to recover the level of a 10 year HP filtered GDP trend (HP10 trend), to recover 3 year HP filtered GDP growth rates (HP3 perc), and to recover the level of a 3 year HP filtered GDP trend (HP3 trend)

Table 2: Severity of crises by Regions - All crises (Losses as percentage of initial GDP)

Variable	Nr.	Africa	Nr.	Europe	Nr.	Latin A.	Nr.	Asia	Nr.	North A.
HP10 perc	79	0.13	22	0.17	39	0.10	39	0.20	1	0.013
HP10 trend	85	0.13	24	0.18	40	0.12	40	0.22	1	0.013
HP3 perc	78	0.07	24	0.10	41	0.12	36	0.10	1	0.008
HP3 trend	86	0.08	24	0.14	45	0.08	39	0.12	1	0.008

NOTE: Estimation of losses to recover a 10 year HP filtered GDP growth (HP10 perc), to recover the level of the 10 year HP filtered GDP (HP10 trend), to recover a 3 year HP filtered GDP growth (HP3 perc) and to recover the level of the 3 year HP filtered GDP (HP3 trend)

Table 3: Severity by income groups (as percentage of initial GDP)

Variable	Nr.	High Income	Nr.	Middle Income	Nr.	Low Income
HP10 perc	21	0.13	122	0.18	61	0.11
HP10 trend	22	0.16	128	0.17	64	0.12
HP3 perc	22	0.07	120	0.11	61	0.07
HP3 trend	23	0.09	128	0.11	68	0.07

NOTE: Estimation of losses to recover a 10 year HP filtered GDP growth (HP10 perc), to recover the level of the 10 year HP filtered GDP (HP10 trend), to recover a 3 year HP filtered GDP growth (HP3 perc) and to recover the level of the 3 year HP filtered GDP (HP3 trend)

Table 4: Severity over time - All crises
(as percentage of initial GDP)

Period	HP10 perc	HP10trend	HP3 perc	HP3 trend
1970 -75	0.02	0.05	0.02	0.06
1975 -80	0.11	0.14	0.19	0.09
1980 -85	0.18	0.14	0.09	0.10
1985 -90	0.12	0.10	0.06	0.07
1990 -95	0.15	0.16	0.09	0.12
1995 -00	0.19	0.20	0.06	0.07
2000 -05	0.11	0.12	0.07	0.07

NOTE: Estimation of losses to recover a 10 year HP filtered GDP growth (HP10 perc), to recover the level of the 10 year HP filtered GDP (HP10 trend), to recover a 3 year HP filtered GDP growth (HP3 perc) and to recover the level of the 3 year HP filtered GDP (HP3 trend)

Table 5: Features of the PDF of total losses.

	HP10perc	HP10trend	HP3perc	HP3trend
99.9 percentile	3.0e+012	3.6e+012	2.4e+012	2.7e+012
99 percentile	1.7e+012	2.0e+012	1.3e+012	1.5e+012
median	1.9e+011	2.2e+011	1.4e+011	1.6e+011
mean	3.0e+011	3.6e+011	2.3e+011	2.7e+011
Std deviation	3.7e+011	4.4e+011	2.9e+011	3.3e+011

Notes:

- (1) Number of simulations equal to 500 000. In all cases, the frequency distribution is the Negative Binomial, the severity PDF is a Weibull
- (2) Losses are measured in constant 2005 USD and correspond to five-year periods

Table 6: LDA, Losses in constant 2005 USD, over periods of five years, as percentage of 2005 World GDP

	HP10perc	HP10trend	HP3perc	HP3trend
99.9% perctl	6.81%	8.18%	5.45%	6.13%
99 % perctl	3.86%	4.54%	2.95%	3.40%
Median	0.43%	0.50%	0.31%	0.36%
Mean	0.68%	0.81%	0.52%	0.61%

Notes:

- (1) Number of simulations equal to 500 000. In all cases, the frequency distribution is the Negative Binomial, the severity pdf is a Weibull, 2005 World GDP from WDI2006

Table 7: Losses after financial crises (as percentage of initial GDP)

Loss Measure	Obs.	Mean	Median	Std. Dev.	Min	Max
AG(10)(5)perc	186	0.3241	0.1632	0.4358	0.0005	2.7310
AG(10)(5)trend	186	0.4257	0.2417	0.5217	0.0005	3.6159
AG(3)(5)perc	182	0.3249	0.1696	0.3997	0.0000	2.0615
AG(3)(5)trend	186	0.4567	0.2776	0.5667	0.0000	4.4838
ABS	110	0.8509	0.1111	2.2077	0.0000	16.0377
AG(10)(10)perc	186	0.5441	0.1632	1.1689	0.0005	8.2695
AG(10)(10)trend	186	1.0587	0.2604	1.9302	0.0005	11.1335
AG(3)(10)perc	180	0.5809	0.1732	1.2679	0.0000	9.2046
AG(3)(10)trend	181	0.9878	0.3403	1.6139	0.0000	9.2046

Table 8: Losses after currency crises (as percentage of initial GDP)

Variable	Obs	Mean	Std. Dev.	Min	Max
AG(10)(5)perc	101	0.2794	0.3949	0.0006	1.8448
AG(10)(5)trend	101	0.3702	0.4870	0.0006	2.2125
AG(3)(5)perc	94	0.2861	0.3658	0.0000	1.7087
AG(3)(5)trend	94	0.3874	0.4260	0.0000	1.7598
HP(10)perc	118	0.1352	0.2825	0.0001	2.2993
HP(10)trend	122	0.1563	0.2947	0.0003	2.2993
HP(3)perc	114	0.0918	0.2111	0.0013	2.0370
HP(3)trend	125	0.0904	0.1293	0.0013	1.0627
ABS	61	0.6811	1.4726	0.0001	7.0657
AG(10)(10)perc	101	0.3674	0.7658	0.0006	6.2588
AG(10)(10)trend	101	0.8587	1.8207	0.0006	11.1335
AG(3)(10)perc	94	0.3635	0.7025	0.0000	5.4637
AG(3)(10)trend	94	0.8509	1.4290	0.0000	8.3445

Table 9: Losses after banking crises (as percentage of initial GDP)

Variable	Obs	Mean	Std. Dev.	Min	Max
AG(10)(5)perc	69	0.3650	0.4636	0.0033	2.7310
AG(10)(5)trend	69	0.4834	0.5726	0.0033	3.6159
AG(3)(5)perc	71	0.3321	0.3925	0.0049	2.0615
AG(3)(5)trend	71	0.4793	0.6519	0.0049	4.4838
HP(10)perc	66	0.1424	0.1914	0.0014	1.2766
HP(10)trend	71	0.1547	0.2084	0.0014	1.3075
HP(3)perc	70	0.0800	0.0763	0.0000	0.4705
HP(3)trend	74	0.0958	0.1075	0.0000	0.7574
ABS	35	0.7631	2.1609	0.0000	12.5942
AG(10)(10)perc	69	0.7637	1.5744	0.0033	8.2695
AG(10)(10)trend	69	1.5239	2.4667	0.0033	11.1335
AG(3)(10)perc	69	0.7888	1.7237	0.0049	9.2046
AG(3)(10)trend	70	1.7113	4.7780	0.0049	9.2046

Table 10: Losses after debt crises (as percentage of initial GDP)

Variable	Obs	Mean	Std. Dev.	Min	Max
AG(10)(5)perc	39	0.5002	0.4842	0.0005	1.6946
AG(10)(5)trend	39	0.6244	0.4797	0.0005	1.6946
AG(3)(5)perc	39	0.4282	0.4024	0.0005	1.8020
AG(3)(5)trend	39	0.5248	0.4477	0.0005	1.8020
HP(10)perc	46	0.1723	0.1413	0.0211	0.5531
HP(10)trend	47	0.1966	0.1597	0.0058	0.7095
HP(3)perc	43	0.1224	0.1079	0.0089	0.4846
HP(3)trend	45	0.1395	0.1108	0.0025	0.4846
ABS	36	0.8658	2.7270	0.0026	16.0377
AG(10)(10)perc	39	1.6642	0.8180	0.0005	4.0388
AG(10)(10)trend	39	1.8217	2.8550	0.0005	11.1335
AG(3)(10)perc	39	0.6273	0.9522	0.0005	4.6050
AG(3)(10)trend	39	1.1286	1.4862	0.0005	5.7084

Table 11: Twin crises - Currency and Debt (as percentage of initial GDP)

Variable	Obs	Mean	Std. Dev.	Min	Max
AG(10)(5)perc	56	0.3983	0.4209	0.0005	1.8448
AG(10)(5)trend	56	0.5286	0.5121	0.0005	2.2125
AG(3)(5)perc	52	0.3849	0.4200	0.0005	1.8020
AG(3)(5)trend	52	0.5198	0.4797	0.0005	1.8020
HP(10)perc	64	0.1334	0.1223	0.0018	0.5170
HP(10)trend	67	0.1561	0.1520	0.0018	0.7095
HP(3)perc	66	0.1172	0.2541	0.0023	2.0370
HP(3)trend	71	0.1032	0.0998	0.0023	0.4846
ABS	42	0.5397	1.2529	0.0026	7.0657
AG(10)(10)perc	56	0.5708	0.9698	0.0005	6.2588
AG(10)(10)trend	56	1.4029	2.5465	0.0005	11.1335
AG(3)(10)perc	52	0.5311	0.9354	0.0005	5.4637
AG(3)(10)trend	52	1.0760	1.3452	0.0005	4.7534

Table 12: Twin crises - Currency and Banking (as percentage of initial GDP)

Variable	Obs	Mean	Std. Dev.	Min	Max
AG(10)(5)perc	78	0.2716	0.3726	0.0005	1.6946
AG(10)(5)trend	78	0.3435	0.4417	0.0005	1.6946
AG(3)(5)perc	73	0.2965	0.3360	0.0005	1.8020
AG(3)(5)trend	73	0.3900	0.3884	0.0005	1.8020
HP(10)perc	101	0.1531	0.3014	0.0018	2.2993
HP(10)trend	103	0.1716	0.3123	0.0018	2.2993
HP(3)perc	95	0.0822	0.1108	0.0013	0.8544
HP(3)trend	103	0.0960	0.1342	0.0013	1.0627
ABS	56	0.6501	1.3273	0.0001	6.8260
AG(10)(10)perc	78	0.3051	0.4492	0.0005	2.1887
AG(10)(10)trend	78	0.8651	2.1069	0.0005	11.1335
AG(3)(10)perc	73	0.3492	0.5440	0.0005	3.7145
AG(3)(10)trend	73	0.7216	1.1281	0.0005	5.7125

Table 13: Twin crises - Debt and Banking (as percentage of initial GDP)

Variable	Obs	Mean	Std. Dev.	Min	Max
AG(10)(5)perc	36	0.5213	0.5045	0.0005	1.6946
AG(10)(5)trend	36	0.6631	0.4996	0.0005	1.6946
AG(3)(5)perc	35	0.4247	0.3862	0.0005	1.8020
AG(3)(5)trend	35	0.5553	0.4129	0.0005	1.8020
HP(10)perc	32	0.2396	0.2447	0.0093	1.2766
HP(10)trend	34	0.2586	0.2595	0.0058	1.3075
HP(3)perc	32	0.1510	0.1293	0.0000	0.4846
HP(3)trend	33	0.1782	0.1590	0.0000	0.7574
ABS	26	1.1814	3.1728	0.0059	16.0377
AG(10)(10)perc	36	0.7952	1.1590	0.0005	5.4650
AG(10)(10)trend	36	1.7638	2.6372	0.0005	11.1335
AG(3)(10)perc	34	0.8204	1.7036	0.0005	9.2046
AG(3)(10)trend	35	1.3295	1.9641	0.0005	9.2046

Table 14: All Twin Crises (as percentage of initial GDP)

Variable	Obs	Mean	Std. Dev.	Min	Max
AG(10)(5)perc	114	0.3173	0.4052	0.0005	1.8448
AG(10)(5)trend	114	0.4238	0.4946	0.0005	2.2125
AG(3)(5)perc	105	0.3344	0.3755	0.0005	1.8020
AG(3)(5)trend	105	0.4527	0.4386	0.0005	1.8020
HP(10)perc	130	0.1588	0.2902	0.0018	2.2993
HP(10)trend	136	0.1759	0.2996	0.0018	2.2993
HP(3)perc	126	0.1068	0.2071	0.0000	2.0370
HP(3)trend	136	0.1078	0.1423	0.0000	1.0627
ABS	78	0.9035	2.2318	0.0001	16.0377
AG(10)(10)perc	114	0.4741	0.9509	0.0005	6.2588
AG(10)(10)trend	114	0.9907	2.0026	0.0005	11.1335
AG(3)(10)perc	104	0.5286	1.1851	0.0005	9.2046
AG(3)(10)trend	105	0.9816	1.5407	0.0005	9.2046

Table 15: Severity of crises by regions - All crises (Losses as percentage of initial GDP)

Variable	Nr.	Africa	Nr.	Europe	Nr.	Latin A.	Nr.	Asia	Nr.	North A.
AG(10)(5)perc	71	0.2481	21	0.4674	37	0.2759	32	0.5112	1	0.0091
AG(10)(5)trend	71	0.3253	21	0.6047	37	0.3909	32	0.5751	1	0.0091
AG(3)(5)perc	62	0.2309	26	0.4075	35	0.3083	32	0.5131	1	0.0390
AG(3)(5)trend	62	0.3080	26	0.5250	35	0.4170	32	0.6949	1	0.0390
ABS	43	0.7816	17	1.2325	21	0.3113	14	1.4106	0	0
AG(10)(10)perc	71	0.4279	21	0.5354	37	0.3072	32	1.0558	1	0.0091
AG(10)(10)trend	71	0.9220	21	1.0626	37	0.7603	32	1.8413	1	0.0091
AG(3)(10)perc	60	0.3427	26	0.5793	35	0.3789	32	1.1557	1	0.0390
AG(3)(10)trend	61	0.6376	26	1.2732	35	0.6525	32	1.6113	1	0.0390

Table 16: Average severity of crises by income groups (percentage of initial GDP)

Variable	Nr.	High Income	Nr.	Middle Income	Nr.	Low Income
AG(10)(5)perc	21	0.2844	112	0.3801	53	0.2215
AG(10)(5)trend	21	0.3901	112	0.4945	53	0.2946
AG(3)(5)perc	21	0.2389	113	0.3800	48	0.2327
AG(3)(5)trend	21	0.4923	113	0.4915	48	0.2890
ABS	10	0.4446	71	0.7927	29	1.1332
AG(10)(10)perc	21	0.4032	112	0.6669	53	0.3403
AG(10)(10)trend	21	1.0673	112	1.1918	53	0.7740
AG(3)(10)perc	21	0.4258	112	0.6814	47	0.4106
AG(3)(10)trend	21	1.1015	113	1.0663	47	0.7485

Table 17: Features of the PDF of total losses (more cases)

	ABS	AG10.10perc	AG10.10trend	AG3.10perc	AG3.10trend
99.9 percentile	5.7e+012	1.8e+013	3.7e+013	2.2e+013	4.7e+013
99 percentile	2.7e+012	8.7e+012	1.7e+013	1.1e+013	2.2e+013
median	1.8e+011	6.4e+011	1.2e+012	8.1e+011	1.6e+012
mean	3.8e+011	1.3e+012	2.4e+012	1.6e+012	3.1e+012
Std. deviation	6.0e+011	1.9e+012	3.8e+012	2.3e+012	4.8e+012

Notes:

- (1) Number of simulations equal to 500 000. In all cases, the frequency distribution is the Negative Binomial, the severity pdf is a Weibull
- (2) Losses are measured in constant 2005 USD and correspond to five-year periods

Table 18: Features of the PDF of total losses (more cases)

	AG10.5perc	AG10.5trend	AG3.5perc	AG3.5trend
99.9 percentile	8.8e+012	1.2e+013	1.1e+013	1.5e+013
99 percentile	4.5e+012	6.2e+012	5.4e+012	7.8e+012
median	4.0e+011	5.4e+011	4.9e+011	6.9e+011
mean	7.3e+011	9.8e+011	8.6e+011	1.2e+012
Std. deviation	9.8e+011	1.3e+012	1.2e+012	1.7e+012

Notes:

- (1) Number of simulations equal to 500 000. In all cases, the frequency distribution is the Negative Binomial, the severity pdf is a Weibull
- (2) Losses are measured in constant 2005 USD and correspond to five-year periods

Table 19: USA 2007 banking crisis losses (percentage of 2007 real GDP)

Loss measure	in Billion USD	% of 2007 GDP
HP(10)(10)perc	3009.28	22.78
HP(10)(10)trend	14027.9	106.22
HP(3) (10)perc	2014.14	15.25
HP(3) (10)trend	7415.09	56.14
HP(3) (5) trend	4043.20	30.61
HP(3) (5) perc	2014.14	15.25
HP(10)(5) trend	6329.73	47.92
HP(10)(5) perc	3009.28	22.78
AG(10)(5) perc	5923.15	44.85
AG(10)(5) trend	5923.15	44.85
AG(3) (5) perc	5552.58	42.04
AG(3) (5) trend	5552.58	42.04
AG(10)(10)perc	15334.18	116.11
AG(10)(10)trend	17874.42	135.34
AG(3) (10)perc	9700.10	73.45
AG(3) (10)trend	17851.69	135.17
ABS	666.10	5.04