

**SAFETY-NET BENEFITS CONFERRED ON DIFFICULT-TO-FAIL-AND-UNWIND
BANKS IN THE US AND EU BEFORE AND DURING THE GREAT RECESSION**

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Abstract: This paper investigates the links between regulatory arbitrage, financial instability, and taxpayer loss exposures. We model and estimate *ex ante* safety-net benefits from increased leverage and asset volatility at a sample of large banks in US and Europe during 2003-2008. Hypothesis tests indicate that, in both crisis and precrisis years, difficult-to-fail-and-unwind (DFU) banks enjoyed substantially higher *ex ante* benefits than other institutions. Compared to the US sample, safety-net benefits prove significantly larger for DFU firms in Europe and bailout decisions are less driven by asset size. Introducing a proxy for differences in government susceptibility to regulatory capture helps to explain bailout decisions in Europe. Our findings suggest that authorities in both venues could better contain safety-net benefits if they refocused their information systems on monitoring volatility as well as capital.

Keywords: safety net, too big to fail policy, regulatory arbitrage, difficult-to-fail banks, financial crisis

JEL Classification: G21, G01

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Alan Greenspan has opined that, during the housing and securitization bubbles that precipitated the 2007-2008 financial meltdown, “managers of financial institutions, the Federal Reserve, and other regulators failed to fully comprehend the underlying size, length, and impact of the negative tail of the distribution of risk outcomes” (quoted in Matthews and Zumbrun, 2010). This paper tests the proposition that adopting a contingent-claims perspective on taxpayers’ evolving exposure to tail risk could have improved managerial and regulatory comprehension.

Using Bancscope data for 2003-2008, we show that routing synthetic estimates of bank leverage and volatility through the Duan-Moreau-Sealey model of safety-net benefits could have helped regulators to detect the buildup of crisis pressure in Europe and the US in 2003-2006. We also investigate whether, during precrisis and early crisis years, ex ante safety-net benefits in the US and Europe were larger at firms that might be deemed difficult to fail and unwind (DFU) ex ante or were revealed to be DFU ex post. We find that, during both 2003-2006 and 2007-2008, DFU banks in the US and Europe enjoyed substantially higher ex ante benefits than other institutions in the sample. In Europe, safety-net benefits prove significantly larger for DFU firms and bailout decisions appear to be less driven by asset size and more by regulatory capture. Across Europe, differences in a government’s susceptibility to regulatory capture help to explain differences in the effectiveness of systemic-risk management and why the crisis hit some countries sooner and harder than others.

1. Safety-Net Costs and Benefits

A nation’s financial safety net consists of whatever array of programs it uses to protect bank depositors and to keep systemically important markets and institutions from breaking down

in difficult circumstances. The avowed goal of safety-net management is to monitor, contain, and finance systemic risk. Systemic risk combines two kinds of risk-taking: calculated risk-taking by protected institutions and partially countervailing risk-management programs operated by safety-net managers. Ideally, safety-net managers safeguard taxpayers interests by making private institutions operate more safely than their managers and stockholders might prefer. Both in Europe and the US, safety-net managers seek to contain risk-taking by restricting the activities of protected institutions and by prescribing risk-based capital requirements and deposit-insurance premia. Although different instruments and functions are often located in different agencies, the net can be envisioned as a holding company managed through a decentralized collection of subsidiary agencies. Considered as a consolidated enterprise, each country's net has its own balance sheet, income statement, and governance network across which taxpayers, bank stakeholders, and the net's managers interact. Its governance procedures are complicated by differences in the capacities of different stakeholders to understand and influence safety-net decisions and these differences vary widely across countries.

Definitions of systemic risk used by the Basel Committee and other policymakers focus on contagion: i.e., a concern for avoiding potential spillovers of institutional defaults across important firms in the financial sector and from this sector to employment and asset values in the real economy. This perspective conceives of safety-net costs simply as negative externalities and fails to acknowledge the extent to which safety-net arrangements complete markets by implicitly accepting deep downside risk ex ante and redistributing most of this tail risk to taxpayers.

Research indicates that bond, stock, and swap markets reward DFU institutions for increases in size and tail risk. See, for example, Brewer and Jagliani (2009), Penas and Unal (2005), and Völz and Wedow (2009). Gropp *et al.* (2011) investigate the competitive effects of government

bail-out policies using a measure of bail-out perceptions that is based on credit-ratings information. With few exceptions, empirical research supports the consensus view that government guarantees distort bank risk-taking. This suggests that taxpayer exposure to tail risk at DFU firms is better interpreted as an implicit contract than an external effect.¹ This contract makes taxpayers unacknowledged equity investors in DFU firms. As any other stakeholder, taxpayers deserve to have their stake serviced fairly and explicitly.

An important lesson of the current crisis is that government guarantees are a form of tax expenditure. Although their precise incidence across the citizenry is not determined in advance, private benefits are skewed toward those with large stakes in protected institutions.

This paper frames the conjectural obligations that government safety nets impose on taxpayers in guaranteeing the performance of protected financial-institutions as the short side of a “taxpayer put” that selected institutions can and do exercise in adverse circumstances. The put reduces the degree to which markets for the liabilities of protected firms are incomplete (cf. Kane, 1980). Costs taxpayers incur in underwriting tail risks at elite institutions are not expected to be fully recovered either *ex ante* or *ex post*. This means that national safety nets are programs of redistributive fiscal policy that subsidize risk taking by systemically important firms.

Despite the global extent of the current crisis, observable spillovers of financial-institution defaults have been minimal. Firms that seemed politically or administratively difficult to fail and unwind (DFU firms) have been characterized as “systematically important” and kept afloat by supporting their access to public and private credit without stopping to resolve their underlying shortage of private capital (i.e., their degree of economic insolvency). In effect, governments authorize institutions to exercise a loss-shifting “taxpayer put” that converts most

¹ Of course, some of the knock-on effects on the labor force and real economy may be accurately characterized as externalities engendered by this implicit contract.

of the losses incurred by *creditors* of insolvent DFU firms into government debt (Kane, 1986; Eberlein and Madan, 2010).

Implicit and explicit costs of supporting the capitalized value of the safety-net subsidies that stakeholders in DFU firms capture from taxpayers represents a cogent way to measure what authorities ought to mean at any time by “systemic risk.” This definition of systemic risk presumes that DFU firms’ political clout ensures that, by engaging in regulatory arbitrage, they can extract hard to observe ex ante safety-net subsidies even in good times. Of course, when and as losses accumulate at prominent firms, ex post subsidies become increasingly visible to other citizens.

Ironically, in economic downturns, the transparency of capital injections and sweetheart loans fuels popular resentment and tends to reduce the flow of ex ante subsidies to the many (generally smaller) banks that are not able politically to command substantial open-bank assistance. Our methods indicate that on average, but not at the margin, support for non-DFU banks fell in 2007-2008. An unfortunate consequence of this distribution effect would be to reduce the flow of new credit to small and medium-sized firms when the economy was at its weakest.

Thinking of taxpayers as investors in DFU firms suggests that policymakers and competitors ought to demand information on the ebb and flow of safety-net subsidies at DFU firms. Developing even rough estimates of the value of safety-net subsidies and recording these estimates on the income statements and balance sheets of governments and DFU firms would let taxpayers track their stake in the safety net, too.

2. Modeling the Determinants of Individual-Firm Risk

At an individual institution, shareholder risk arises as a mixture of leverage and the volatility of its return on assets over a specified time horizon. This paper treats leverage as the ratio of the par value of an institution's debt (B) to the estimated market value of its assets (V) and employs a one-year horizon for measuring safety-net benefits. Benefits are estimated via a two-equation model developed by Duan, Moreau, and Sealey (DMS, 1992). In this model, value-maximizing decisions about asset volatility constrain a firm's choice of leverage and, amplified by leverage, volatility extracts nontransparent safety-net benefits from taxpayers.

The DMS model recognizes that market and regulatory discipline force a financial firm to carry an equity position that imperfectly informed creditors regard as large enough to support the risks it takes. However, creditors are assumed to regard the conjectural value of the off-balance-sheet capital that government guarantees supply through the implicit taxpayer put as a close substitute for on-balance-sheet capital that might otherwise be formally supplied by the firm's shareholders.

The DMS model begins with an efficient portfolio-choice locus that serves to balance the value of leverage and volatility to the bank and combines this equation with a per-period safety-net benefits "production function." The production function linearizes and slightly expands Merton's pioneering model of deposit insurance (1977, 1978). Merton portrays safety-net access as an option that allows bank owners to put the assets of the bank to safety-net managers as soon as it falls below the face value of the bank's debt. Kane (1986) argues that the taxpayers' side of the takeover option tends to be exercised suboptimally. Due to examination lags and political pressure, authorities do not enforce their takeover rights until the assets of the firm are deeply under water. To allow for this, Ronn and Verma (1986) assume that assets must fall fully 3

percent below debt value before the option can be exercised. Hovakimian and Kane (2000) show that allowing a bank to make interim dividend payments lets firms extract additional benefits.²

Firms engage in risk-shifting whenever they expose creditors, derivatives counterparties or guarantors to loss without compensating them adequately. The DMS model measures the per-period return on the current value of safety-net benefits by a variable designated as IPP. IPP is the annual "fair insurance premium percentage" per dollar, per Euro, or per pound of debt that would let taxpayers break even if they were to hedge the amount of the bank's tail risk that is shifted to them. The linear production function makes IPP an increasing function of a bank's asset volatility (σ_V) and leverage. Ignoring the error term, the structural equation for IPP would be:

$$IPP = \gamma_1 \sigma_V + \gamma_2 (B/V) \quad (1)$$

Pennachi (1987a and b) shows that IPP (i.e., each of γ_1 and γ_2) would be larger if safety-net coverage were modeled more realistically as extending far beyond a single year.

Duan, Moreau, and Sealey (1992) stress that market and regulatory disciplines prevent value-maximizing B/V (leverage) from being chosen independently of σ_V (volatility). To contain risk-shifting at all, counterparties and regulators must require a bank's B/V to fall when and as σ_V increases.

The DMS model consists of two equations (2) and (3):

$$B/V = \alpha_0 + \alpha_1 \sigma_V + \varepsilon_1 . \quad (2)$$

$$IPP = \beta_0 + \beta_1 \sigma_V + \varepsilon_2 . \quad (3)$$

² Carbo, Kane, and Rodriguez (2008, 2011) incorporate both extensions in using the DMS model to compare regulatory and merger policy performance across countries.

Equation (3) is a quasi-reduced form equation that comes from substituting (2) into (1). It treats the endogenous variable σ_V as a predetermined regressor. This assumption produces a recursive model in which bank B/V and IPP both depend on σ_V .

Equation (2) expresses the idea that regulators and creditors constrain bank risk-taking to a mutually acceptable locus of leverage and volatility pairs. If safety-net managers had no incentive conflicts and could observe σ_V and control B/V perfectly, they would set B/V so that IPP equaled the value of the sum of explicit and implicit premiums they could impose on the bank. Taking total derivatives, the slope coefficients in (2) and (3) may be interpreted as follows:

$$\alpha_1 = \frac{d(B/V)}{d\sigma_V}, \quad (4)$$

$$\beta_1 = \frac{\partial IPP}{\partial \sigma_V} + \frac{\partial IPP}{\partial (B/V)} \alpha_1 = \gamma_1 + \gamma_2 \alpha_1. \quad (5)$$

The partial derivatives γ_1 and γ_2 that appear in equation (5) are positive. They describe the incremental value that bank stockholders could extract from the safety net if bankers were free to make unconstrained adjustments in volatility and leverage, respectively. To prevent a corner solution, either or both of two conditions must be met. Imperfectly informed safety-net officials and private counterparties must monitor and constrain bank risk taking at the margin or managers must believe that unbridled pursuit of safety-net subsidies would work against their career interests. Equations (4) and (5) express the effects of “outside discipline.” At DFU banks during the years we examine, the depth of the crisis indicates that managerial restraint or “inside discipline” was prepared to accommodate a substantial amount of tail risk.

Given the external discipline a bank faces, the sign of β_1 in equation (2) indicates whether, in a country’s particular contracting environment and economic circumstances, increases in asset volatility can increase the value of the implicit and explicit access to safety-net

support that is imbedded in the bank's stock price. To neutralize risk-shifting incentives at the margin, disciplinary penalties intended to induce a decline in B/V must be large enough to offset fully whatever increase in IPP might otherwise be generated by choosing a higher σ_V . In firms for which the total derivative β_1 is positive, risk-shifting incentives are not completely neutralized by inside and outside discipline.

For market and regulatory pressure to discipline and potentially to neutralize incremental risk-shifting incentives, two conditions must be met:

Bank capital increases with volatility: $\alpha_1 < 0$,

Guarantee values do not rise with volatility: $\beta_1 \leq 0$.

None of the three variables featured in the DMS model is directly observable.

However, Marcus and Shaked (1984) show how to use option-based models of deposit insurance to track these variables synthetically. The first step in the Marcus-Shaked procedure is to obtain tracking values for V and σ_V by numerical methods. These values are then used to estimate IPP conservatively as the value of a one-year put option on bank assets (Merton's "default put"). As explained more fully in Hovakimian and Kane (2000), a key step in the procedure is to use Ito's lemma to transform σ_V into σ_E , the instantaneous standard deviation of equity returns.

3. Preliminary Look at Mean Sample Experience

Table I lists the sources of the data we analyze. It also introduces and defines some control and shift variables (such as DFU status) that we incorporate into our regression experiments. DFU status is proxied ex ante by a size criterion (DFUxa) and ex post by the receipt of open-bank assistance (DFUxp). As for the selection of DFUxp banks, we rely on two main sources. To identify the receipt of aid by European Union banks, we accept the European

Commission (EC) definition of State aid.³ This includes capital injections/recapitalization and debt guarantees⁴. To identify US DFUxp banks, we rely on US Treasury data covering participation in the Asset Guarantee Program, the Capital Assistance Program and the Capital Purchase Program⁵. Federal Reserve aid to US and European banks was substantial, but was not reported publicly until long after our analysis was finished. A detailed list of sources is provided in the Appendix.

Table II lists the number of observations in our sample by country and shows the evolution of leverage, volatility, and the insurance premium in 2003, 2005 and 2008. Over a third of the observations come from the US and Germany and roughly 80 percent come from the last six countries listed in the table.

Table III describes the mean behavior of leverage, volatility, and the fair insurance premium percentage for different groupings of banks. As a robustness check, IPP, σ_v , and B/V are calculated in two different ways: by the Ronn and Verma (RV) procedure and by a maximum-likelihood (ML) method developed by Duan (1994). Table III also records the results of t-tests for differences in the mean values found between US and European banks and between DFU and other banks in various regions. Mean differences are significant at conventional levels in every instance.

Mean safety-net benefits range between 10 and 22 basis points. Mean leverage proves uniformly higher under the ML procedure, while volatility and IPP are often lower. From a

³ Regulatory interventions were managed and supervised by the EC. According to the annual reports on State aid expenditure for 2009, Member States reported aid of € 351.7 billion. This corresponds to 2.98% of EU-27 GDP. Approximately half of State aid was provided in the form of recapitalization (€ 139.43 billion). Other aid took the form of guarantees (€ 128.15 billion), impaired assets (€ 75.27 billion) and direct liquidity funding (€ 8.8 billion).

⁴ The main link to the different national sources can be found at:
<http://europa.eu/rapid/pressReleasesAction.do?reference=MEMO/11/68&format=PDF&aged=0&language=EN&guiLanguage=en>

⁵ The list of banks participating in the different programs is provided at:
<http://www.treasury.gov/initiatives/financial-stability/investment-programs/Pages/default.aspx>

policy perspective, either procedure finds that both kinds of DFU banks show higher safety-net benefits than other banks in each region and time frame. In most cases, DFU institutions show more leverage, too. Both before and during the crisis, DFU banks in Europe show more leverage and safety-net benefits than DFU banks in the US and DFUxp banks extracted more benefits than DFUxa firms. During the first two years of the crisis, DFU banks in Europe and the US decreased volatility, reduced their leverage, and did suffer procyclical cuts in the mean size of ex ante safety net benefits.

4. Regression Analysis

Subsequent tables report difference-on-difference regression experiments that expand equations (2) and (3) to introduce three control variables and three parameter-shift indicators for DFU banks⁶. The log of asset size is introduced as a hard-to-interpret proxy that aggregates the influence of political clout, complexity, and public awareness separately from measures of ex ante and ex post evidence of DFU status. Transparency International's Corruption Perception Index (10-CPI) is used to represent cross-country differences in a government's susceptibility to regulatory capture. We include the so-called "fear index" (VIX) as a way to distinguish the impacts of marketwide and idiosyncratic volatility.

Pooling precrisis and crisis years, Table IV applies this model separately to panels of US and European banks and bank holding companies. The signs of all coefficients and the rough magnitude of p-values are similar in all parallel RV and ML runs.

Given the large size of these samples and the near-zero value of focal coefficients, the Lindley Paradox suggests that we employ a more rigorous standard for statistical significance

⁶ Han and Phillips (2011) provide a comprehensive discussion of the advantages of using fixed-effects estimation and difference-on-difference regression to analyze dynamic panel datasets.

than the conventional 5 percent. Our discussions benchmark significance at 2 percent, but the reader is free to adopt a lighter or tougher standard.

The shift variable in the size effect for DFU banks is never significant and is dropped from subsequent runs. Except for VIX and the corruption-perceptions index (which proves significant only in Europe where there is cross-section as well as time-series variation), covariance tests for differences in individual coefficients between US and European leverage equations usually meet the significance standard of 2 percent. The effects of asset size on safety-net benefits (i.e., on IPP) are similar across countries, but at the margin DFUxa banks in the US extract slightly more benefits than their European counterparts.

Table V.A shows the effect of employing the ex post definition of the DFU shift variable. In this experiment, DFUxp banks are banks that received aid during the crisis. Although R-squared remains much the same, this definition renders differences between Europe and the US in coefficients for idiosyncratic volatility, asset size, corruption, and the intensified role of volatility for DFU banks sharper and uniformly more significant. In particular, even though DFUxp banks in the US find themselves penalized more heavily for increased volatility via the leverage equation than DFUxp banks in the EU, they manage to extract incremental benefits from the safety net more successfully (0.035 in the US vs. 0.029 in Europe according to the ML procedure). Additionally, the proxy for regulatory capture (10-CPI) is significant only for the EU sample.

Table V.B re-runs the Table V.A regression experiment using Heckman's (1976, 1978) procedure for endogenizing the ex post selection process for providing capital and/or liquidity support to DFU banks. This procedure adds a third equation to our model. This selection equation is linked to the other equations by a variable that Heckman calls Lambda (also known

as the Mills Odds Ratio for selection) which is calculated from the selection model. This linking variable is then added to the list of the potential determinants of leverage and safety-net benefits in expanded versions of equations (2) and (3).

Although the value and significance of individual coefficients in the B/V and IPP models are not much different from those in Table V.A, coefficient differences in probit selection models for receiving State aid prove markedly different. In Europe, asset size has no significant effect: Idiosyncratic volatility and the corruption index dominate government bailout decisions. In particular, the impact of σ_V on the probability of receiving State aid in Europe is 0.815 according to data based on the ML procedure, while the impact of the regulatory capture proxy (CPI-10) is 0.916. In the US, idiosyncratic volatility is more or less equally important, but size has a large effect. The coefficient for size at US banks estimated with either the RV and ML procedure is roughly 1.50. Limited to minor time-series variation, the corruption index shows no predictive power in the US.

Tables VI.A and VI.B run the abridged model of Table IV separately for pre-crisis and crisis years: i.e., for 2003-2006 and 2007-2008. The most interesting differences are those in which the subperiod coefficients both lie substantially above or below those found in the pooled equation. Such a finding establishes a prima facie case against pooling data across separate regimes. This phenomenon occurs for the incremental effects on IPP of the DFU_xa shift variable in both regions (+), for corruption (+) in Europe, and for size (-) and volatility (+) in the US. In particular, taking the ML model as a reference, the coefficient of the critical DFU shift variable in the IPP equation changes only slightly (from 0.026 to 0.029) in the EU sample between the precrisis and crisis periods. But the coefficient for the US sample shows increased subsidization at the margin, jumping from 0.034 to 0.044. As for the corruption-perceptions index, the

coefficient in the IPP equation almost doubles (from 0.007 to 0.012), while the index continues to be insignificant in the US sample. The coefficient of σ_v in the IPP equation increases for the EU sample from 0.007 in precrisis years to 0.011 in the crisis period while the coefficient for the US sample increases only half as much, from 0.013 to 0.015.

Table VII reports the significance of differences between coefficients in precrisis and crisis years for US and European banks separately. In Europe, crisis years show an intensification in incremental subsidization for a few variables and equations: for idiosyncratic volatility on safety-net benefits; for the DFU shift variable on leverage under the ML procedure and on IPP using the RV approach; and for corruption in the leverage equation and in the ML model for IPP. In the US, the incremental effects of asset size and the DFU shift variable intensify for both variables under both procedures.

Tables VIII.A and VIII.B re-run the experiments of Tables VI.A and VI.B using the Heckman procedure and the ex post DFU indicator. For both the precrisis and crisis eras, the signs of all coefficients for the European and US samples remain the same. However, the magnitude of individual coefficients is often reduced. For the precrisis samples, coefficient differences between US and Europe for the corruption and VIX index are seldom significant, but the greater role for market volatility in explaining US bank leverage in crisis years continues to be significant.

As in Table V, the importance of the Heckman experiments lies in creating the opportunity to examine the selection equations. Asset size (and to a lesser extent, idiosyncratic volatility) is a more important determinant of bailout assistance in the US than in Europe, while a European country's corruption index strongly influences its bailout decisions. The inference is that banks may not be too big to fail in Europe, but they might be too politically connected.

For European and US sample banks, Table IX shows that the leverage and IPP equations underwent many statistically significant changes between precrisis and crisis periods.

Economically, the effects on IPP generation are the most interesting. In Europe, the shift in the volatility slope for DFU banks explicitly receiving State aid increased by roughly 50 percent under both procedures. In the US, this coefficient also increased, but the effect is smaller and significant only under the RV procedure.

5. Special Cases of Portugal, Ireland, Italy, and Spain

In several European countries affected severely by the crisis, doubt arose about the government's ability to resolve the losses experienced by its largest banks. Greece (for which we lack data), Ireland, Portugal, Italy, and Spain all saw substantial increases in the credit premium paid on their sovereign debt. Tables X.A and X.B apply the expanded DMS model to the high-premium countries for which we have data.

Although idiosyncratic volatility is always significant in these four countries, market volatility is not. Time-series variation in the index of perceived corruption almost always impacts leverage, IPP, and selection significantly. However, the economic significance of the proxy for susceptibility to regulatory capture (10-CPI) is higher in Ireland (0.021 in the ML version of the IPP equation) than in Portugal (0.011), Spain (0.008) or Italy (0.006). Size plays a significant role in the selection equation except in Portugal. Idiosyncratic volatility increases safety-net benefits more in Portugal (0.010) and Ireland (0.018) than in Spain (0.008) and Italy (0.006).

Table X.C shows that almost all coefficient differences are significant across country pairs. Ignoring coefficient differences and discarding the market-volatility term, Table XI tests for differences that apply in precrisis and crisis periods when the DMS model fitted to the DFU

banks that were bailed out in these four countries is compared with the Table VI.A and VI.B models estimated across the full sample of European banks. The most striking differences between these two panels and periods is the much greater importance found during the crisis years for asset size and the proxy for susceptibility to regulatory capture.

6. Lessons and Policy Implications

Three important lessons emerge from our work. The first concerns authorities' convenient claim that crisis pressures could not be foreseen. Despite being limited to annual data for key variables, changes in volatility and leverage consistently help to predict changes in the flow of safety-net benefits across different models, regions, and time periods. The second lesson is that the mean flow of ex ante benefits declined in the face of the increased public accountability generated by the transparency of ex post bailout expense. Finally, the cross-country proxy for susceptibility to regulatory capture (the index of perceived corruption) helps to explain safety-net benefits and bailout decisions in Europe.

The clear policy implication of these lessons is that authorities could be incentivized to do a better job of controlling safety-net benefits if they and DFU institutions expanded their information systems so that they could track IPP in a transparent manner. As intricate as it may seem, the stochastic and econometric plumbing underlying our equities-based estimates of volatility and safety-net benefits is still at an early stage of evolution. Complementary estimates could be engineered using richer stochastic processes and datasets that incorporated the prices of debt and derivative instruments. We encourage others to do this and are confident that they will.

Another way to improve information flow would be to require bank managers to report data on earnings and net worth more frequently and under civil or even criminal penalties for

deliberate or negligent misrepresentation. Data on market capitalization are available in real time, as are data on stock-market returns. If the values of on-balance-sheet and off-balance-sheet positions were reported weekly or monthly to national authorities, rolling regression models could be used to estimate changes in the flow of safety-net benefits in ways that would allow regulators to observe, manage, report, and service explicitly taxpayers' stake in the safety net in a timely manner.

REFERENCES

- Brewer, Elijah, and Julapa Jagliani, 2009. "How Much Did Banks Pay to Become Too-Big-to-Fail and to Become Systematically Important?" Federal Reserve Bank of Philadelphia Working Paper No. 09-34 (December 3).
- Carbo, Santiago, Edward Kane, and Francisco Rodriguez, 2008. Evidence of Differences in the Effectiveness of Safety-Net Management in European Union Countries, *Journal of Financial Services Research*, 34, 151-76.
- _____ 2011. Regulatory Arbitrage in Cross-Border Mergers Within the EU, *Journal of Money Credit and Banking* (forthcoming).
- Duan, Jin-Chuan, 1994. Maximum Likelihood Estimation Using Price Data of the Derivative Contract, *Mathematical Finance*, 4, 155-67.
- Duan, J-C, Arthur F. Moreau, and C. William Sealey, 1992. Fixed-Rate Deposit Insurance and Risk-Shifting Behavior at Commercial Banks, *Journal of Banking and Finance*, 16, 715-42.
- Duan, Jin-Chuan, and Jean-Guy Simonato, 2002. Maximum Likelihood Estimation of Deposit Insurance Value with Interest-Rate Risk, *Journal of Empirical Finance*, 9, 109-32.
- Eberlein, E. and D.B. Madan, 2010. Capital requirements, and taxpayer put option values for the major US banks. Mimeo.
- Gropp, Reint, Hakenes, Hendrik and Isabel Schnabel, 2011, Competition, Risk-shifting, and Public bail-out Policies, *Review of Financial Studies*, forthcoming.
- Han, Chirok and Peter C.B. Phillips, 2011. First difference MLE and dynamic panel estimation, Cowles Foundation Discussion Paper No. 1780.

- Heckman, James, 1976. The Common Structure of Statistical Models of Truncation, Sample Selection and Limited Dependent Variables and a Sample Estimator for Such Models, *Annals of Economic and Social Measurement*, 5, 475-92.
- _____. 1978 “Dummy Endogenous Variables in a Simultaneous Equation System,” *Econometrica*, 46, 931-59.
- Hovakimian, Armen, and Edward J. Kane, 2000. Effectiveness of Capital Regulation at U.S. Commercial Banks, 1985-1994, *Journal of Finance*, 55(March), 451-469.
- Kane, Edward J., 1980. Market Incompleteness and Divergences Between Forward and Future Interest Rates, *Journal of Finance*, 35, 221-234.
- _____, 1986. Appearance and reality for deposit institutions: The case for reform, *Journal of Banking and Finance*, 175-188.
- _____, 2009. Extracting Nontransparent Safety Net Subsidies by Strategically Expanding and Contracting a Financial Institution’s Accounting Balance Sheet, *Journal of Financial Services Research*, 36, 161-68.
- Marcus, Alan, and Israel Shaked, 1984. The Valuation of FDIC Deposit Insurance Using Option-Pricing Estimates, *Journal of Money, Credit, and Banking*, 16, 446-460.
- Matthews, Steve, and Joshua Zumbrun (2010). “Greenspan Says Fed, Regulators ‘Failed’ During Crisis (Update2),” *Bloomberg Businessweek*, Feb. 18.
- Merton, Robert C., 1977. An Analytic Derivation of the Cost of Deposit Insurance and Loan Guarantees, *Journal of Banking and Finance*, 1, 3-11.
- _____, 1978. on the Cost of Deposit Insurance When There Are Surveillance Costs, *Journal of Business*, 51, 439-52.
- Molyneux, Philip, Klaus Schaeck, and Tim Mi Zhou, 2010. *‘Too-Big-to-Fail’ and Its Impact on Safety Net Subsidies and Systemic Risk*, Bangor, Wales: Bangor University Working Paper.
- Penas, Maria F., and Haluk Unal, 2004. “Gains in Bank Mergers: Evidence from the Bond Markets,” *Journal of Financial Economics*, 74, 149-179.
- Pennacchi, George, 1987a, A Reexamination of the over- or under-pricing of deposit insurance, *Journal of Money, Credit, and Banking* 19, 340-360.
- _____, 1987b, Alternative forms of deposit insurance: Pricing and bank incentive issues, *Journal of Banking and Finance* 11, 291-312.

Ronn, Ehud, and A.R. Verma, 1986. Pricing Risk-Adjusted Deposit Insurance: An Option-Based Model, *Journal of Finance*, 41, 871-95.

Völz, Manja, and Michael Wedow, 2009. “Does Banks’ Size Distort Market Prices? Evidence of Too-Big-to-Fail in the CDS Market.” Frankfurt: Deutsche Bundesbank Discussion Paper (June).

TABLE I
SAMPLE SIZE (NUMBER OF OBSERVATIONS) AND EVOLUTION OF THE LEVERAGE RATIO (B/V), MEAN FAIR PREMIUM (IPP), AND VOLATILITY OF RETURN ON ASSETS (σ_v) ACROSS COUNTRIES AND OVER TIME (2003, 2005, 2008)

ML: maximum likelihood; RV: Ronn-Verma
 Frequency of the data: annual

	Obs. 2003- 2008	2003						2005						2008					
		B/V (%)		IPP (%)		σ_v (%)		B/V (%)		IPP (%)		σ_v (%)		B/V (%)		IPP (%)		σ_v (%)	
		ML	RV	ML	RV	ML	RV	ML	RV	ML	RV	ML	RV	ML	RV	ML	RV	ML	RV
<i>Austria</i>	476	89.04	90.52	0.157	0.162	1.400	1.358	87.77	90.28	0.157	0.161	1.379	1.354	87.68	90.13	0.154	0.158	1.354	1.334
<i>Belgium</i>	627	92.65	94.80	0.151	0.154	1.843	1.808	91.53	93.89	0.149	0.153	1.833	1.799	90.86	93.47	0.147	0.151	1.802	1.778
<i>Denmark</i>	206	90.29	92.04	0.282	0.283	3.016	2.978	88.91	91.49	0.278	0.282	2.964	2.931	88.24	91.01	0.275	0.276	2.916	2.875
<i>Finland</i>	78	92.05	94.09	0.194	0.199	2.371	2.311	91.77	92.88	0.193	0.198	2.345	2.286	90.91	92.63	0.191	0.195	2.304	2.252
<i>Luxembourg</i>	426	95.91	96.57	0.133	0.134	1.588	1.556	94.75	96.40	0.131	0.133	1.565	1.541	93.86	95.65	0.129	0.132	1.547	1.511
<i>Netherlands</i>	203	89.92	91.41	0.152	0.153	1.899	1.865	88.30	90.34	0.151	0.153	1.882	1.854	87.97	89.48	0.150	0.151	1.855	1.825
<i>Portugal</i>	158	90.03	92.80	0.153	0.156	1.745	1.738	89.76	91.43	0.151	0.154	1.737	1.714	88.98	91.71	0.149	0.152	1.708	1.688
<i>Sweden</i>	263	92.09	94.64	0.178	0.178	1.746	1.716	91.88	94.25	0.176	0.178	1.722	1.687	92.53	93.39	0.172	0.176	1.703	1.669
<i>Ireland</i>	157	90.03	91.19	0.153	0.155	1.763	1.712	88.36	89.62	0.151	0.153	1.729	1.701	88.57	88.94	0.149	0.150	1.699	1.672
<i>United Kingdom</i>	864	86.58	89.19	0.221	0.227	3.295	3.253	86.20	88.19	0.218	0.223	3.252	3.211	85.38	89.02	0.213	0.220	3.195	3.167
<i>Spain</i>	531	83.77	85.83	0.218	0.220	1.501	1.463	83.51	84.57	0.215	0.219	1.477	1.434	82.92	85.19	0.211	0.216	1.451	1.420
<i>France</i>	1112	89.75	91.56	0.214	0.214	1.438	1.424	88.21	90.18	0.211	0.213	1.434	1.406	88.99	90.89	0.208	0.210	1.418	1.388
<i>Italy</i>	1236	88.76	90.38	0.203	0.208	1.331	1.285	88.36	90.09	0.200	0.206	1.305	1.268	87.97	89.82	0.197	0.203	1.281	1.252
<i>Germany</i>	2227	88.96	89.72	0.145	0.148	1.652	1.603	87.65	88.87	0.145	0.147	1.631	1.594	88.06	89.57	0.143	0.144	1.600	1.563
<i>United States</i>	2153	82.57	83.72	0.134	0.138	1.380	1.346	82.31	83.52	0.134	0.136	1.379	1.344	83.06	82.85	0.131	0.133	1.353	1.327
TOTAL	11117	85.29	85.74	0.144	0.148	1.755	1.731	84.00	85.20	0.143	0.147	1.750	1.707	84.83	84.97	0.141	0.144	1.727	1.679

TABLE II
DEFINITIONS AND SOURCES FOR VARIABLES

<i>Variable</i>	<i>Definition</i>	<i>Source</i>
B/V (%)	Leverage, measured as the ratio of the book value (B) of deposits and other debt to the market value of a bank's assets (V).	Bank-level data to compute this variable are obtained from the Bureau-Van Dijk Bankscope database.
IPP (%)	"Fair" insurance premium percentage, defined as the per-period flow of safety-net benefits that bank stockholders enjoy.	Bank-level data to compute this variable are obtained from the Bureau-Van Dijk Bankscope database.
σ_V (%)	Volatility, defined as the standard deviation of the return on bank assets	Bank-level data to compute this variable are obtained from the Bureau-Van Dijk Bankscope database.
<i>Size (log total assets)</i> <i>(Eur mill)</i>	Size of the banks measured by total book value of assets.	Bank-level data to compute this variable are obtained from the Bureau-Van Dijk Bankscope database.
<i>Corruption perception index</i> <i>(10-CPI)</i>	Transparency International's Corruptions Perceptions Index (CPI) is an aggregate indicator that ranks countries in terms of the degree to which corruption is perceived to exist among public officials and politicians. It is a composite index drawing on corruption-related data by a variety of independent and reputable institutions. The main reason for using an aggregated index of individual sources is that a combination of sources measuring the same phenomenon is more reliable than each source taken separately. The CPI ranges 1 to 10. Higher values of the index show less corruption. In order to normalize the values we have redefined the indicator as 10-CPI so that higher values show more corruption.	Transparency international (www.transparency.org)
<i>Market volatility (VIX)</i>	The VIX is calculated and disseminated in real-time by the Chicago Board Options Exchange. It is a weighted blend of prices for a range of options on the S&P 500 index. On March 26, 2004, the first-ever trading in futures on the VIX Index began on CBOE Futures Exchange (CFE). The formula uses a kernel-smoothed estimator that takes as inputs the current market prices for all out-of-the-money calls and puts for the front month and second month expirations.[1] The goal is to estimate the implied volatility of the S&P 500 index over the next 30 days. The VIX is the square root of the par variance swap rate for a 30 day term initiated today. Note that the VIX is the volatility of a variance swap and not that of a volatility swap (volatility being the square root of variance).	Chicago Board of Exchange (http://www.cboe.com/micro/vix/introduction.aspx)
<i>DFU Status</i>	A binary variable that takes on the value of unity for banks that alternately either received open-bank assistance (DFUxp) or fell in the first decile of average 2003-2008 asset size for US and European banks in the Bankscope database (DFUxa).	Deciles are calculated by the authors. Identity of banks receiving equity injections is hand-collected.

TABLE III
MEAN LEVERAGE RATIO (B/V), MEAN FAIR PREMIUM (IPP), AND VOLATILITY OF RETURN ON ASSETS (σ_v): ALL BANKS, DFUxa and DFUxp BANKS IN EUROPE AND IN THE US

Country	B/V (%)		IPP (%)		σ_v (%)	
	RV	ML	RV	ML	RV	ML
<i>ALL BANKS (FULL SAMPLE)</i>	84.8	87.1	0.143	0.119	1.815	1.582
<i>ALL BANKS IN EUROPE</i>	85.3	86.0	0.153	0.134	1.988	1.727
<i>ALL BANKS IN THE US</i>	82.5	83.9	0.139	0.127	1.490	1.368
<i>DFUxa BANKS (FULL SAMPLE)</i>	86.9	89.8	0.167	0.145	1.593	1.597
<i>DFUxp BANKS (FULL SAMPLE)</i>	88.0	90.9	0.174	0.156	1.669	1.490
<i>DFUxa BANKS IN EUROPE</i>	88.1	90.0	0.179	0.164	1.696	1.487
<i>DFUxp BANKS IN EUROPE</i>	89.3	91.6	0.189	0.180	1.792	1.594
<i>DFUxa BANKS IN THE US</i>	80.5	82.2	0.127	0.116	1.396	1.284
<i>DFUxp BANKS IN THE US</i>	83.4	84.2	0.140	0.134	1.503	1.411
<i>ALL BANKS IN EUROPE (PRE 2007)</i>	86.7	88.0	0.157	0.163	2.134	2.166
<i>ALL BANKS IN THE US (PRE 2007)</i>	83.2	84.3	0.149	0.156	1.529	1.632
<i>ALL BANKS IN EUROPE (2007-2008)</i>	83.9	84.3	0.132	0.138	1.842	1.931
<i>ALL BANKS IN THE US (2007-2008)</i>	81.1	81.5	0.128	0.137	1.344	1.388
<i>DFUxa BANKS IN EUROPE (PRE 2007)</i>	90.4	92.6	0.198	0.185	1.591	1.403
<i>DFUxa BANKS IN THE US (PRE 2007)</i>	81.5	82.4	0.158	0.146	1.343	1.211
<i>DFUxa BANKS IN EUROPE (2007-2008)</i>	85.7	88.6	0.165	0.150	1.967	1.663
<i>DFUxa BANKS IN THE US (2007-2008)</i>	78.2	80.1	0.119	0.102	1.491	1.396
<i>DFUxp BANKS IN EUROPE (PRE 2007)</i>	92.3	93.4	0.215	0.220	1.635	1.523
<i>DFUxp BANKS IN THE US (PRE 2007)</i>	83.8	84.1	0.176	0.160	1.428	1.323
<i>DFUxp BANKS IN EUROPE (2007-2008)</i>	89.9	90.1	0.179	0.162	2.123	1.815
<i>DFUxp BANKS IN THE US (2007-2008)</i>	82.3	83.1	0.129	0.118	1.538	1.493
<i>Mean difference tests: ALL BANKS IN EUROPE vs. ALL BANKS IN THE US</i>	0.006	0.007	0.008	0.005	0.009	0.007
<i>Mean difference tests: ALL BANKS vs. DFUxa BANKS (FULL SAMPLE)</i>	0.012	0.011	0.008	0.004	0.005	0.006
<i>Mean difference tests: ALL BANKS vs. DFUxp BANKS (FULL SAMPLE)</i>	0.008	0.006	0.005	0.003	0.002	0.001
<i>Mean difference tests: DFUxa vs. DFUxp BANKS (FULL SAMPLE)</i>	0.005	0.004	0.003	0.001	0.001	0.001
<i>Mean difference tests: ALL BANKS IN EUROPE vs. DFUxa BANKS IN EUROPE</i>	0.009	0.007	0.004	0.003	0.004	0.002
<i>Mean difference tests: ALL BANKS IN THE US vs. DFUxa BANKS IN THE US</i>	0.012	0.013	0.013	0.015	0.019	0.016
<i>Mean difference tests: ALL BANKS IN EUROPE vs. DFUxp BANKS IN EUROPE</i>	0.007	0.005	0.0003	0.001	0.002	0.002
<i>Mean difference tests: ALL BANKS IN THE US vs. DFUxa BANKS IN THE US</i>	0.009	0.010	0.010	0.011	0.013	0.010
<i>Mean difference tests: DFUxa BANKS IN EUROPE vs. DFUxp BANKS IN THE US</i>	0.002	0.003	0.005	0.004	0.007	0.006
<i>Mean difference tests: DFUxp BANKS IN EUROPE vs. DFUxp BANKS IN THE US</i>	0.001	0.001	0.003	0.002	0.004	0.004
<i>Mean difference tests: ALL BANKS IN EUROPE (PRE 2007) vs. DFUxa BANKS IN EUROPE (PRE 2007)</i>	0.005	0.003	0.004	0.002	0.004	0.003
<i>Mean difference tests: ALL BANKS IN THE US (PRE 2007) vs. DFUxa BANKS IN THE US (PRE 2007)</i>	0.012	0.014	0.005	0.004	0.006	0.004
<i>Mean difference tests: ALL BANKS IN EUROPE (2007-2008) vs. DFUxa BANKS IN EUROPE (2007-2008)</i>	0.004	0.003	0.005	0.003	0.002	0.004
<i>Mean difference tests: ALL BANKS IN THE US (2007-2008) vs. DFUxa BANKS IN THE US (2007-2008)</i>	0.008	0.011	0.005	0.007	0.004	0.005
<i>Mean difference tests: DFUxa IN EUROPE (PRE 2007) VS. DFUxa IN EUROPE (2007-2008)</i>	0.010	0.009	0.012	0.015	0.011	0.017
<i>Mean difference tests: DFU IN THE US (PRE 2007) VS. DFUxa IN THE US (2007-2008)</i>	0.008	0.007	0.010	0.005	0.013	0.010
<i>Mean difference tests: ALL BANKS IN EUROPE (PRE 2007) vs. DFUxp BANKS IN EUROPE (PRE 2007)</i>	0.001	0.001	0.001	0.001	0.002	0.002
<i>Mean difference tests: ALL BANKS IN THE US (PRE 2007) vs. DFUxp BANKS IN THE US (PRE 2007)</i>	0.008	0.009	0.003	0.002	0.004	0.003
<i>Mean difference tests: ALL BANKS IN EUROPE (2007-2008) vs. DFUxp BANKS IN EUROPE (2007-2008)</i>	0.002	0.002	0.003	0.001	0.001	0.002
<i>Mean difference tests: ALL BANKS IN THE US (2007-2008) vs. DFUxp BANKS IN THE US (2007-2008)</i>	0.002	0.001	0.002	0.004	0.002	0.003
<i>Mean difference tests: DFUxp IN EUROPE (PRE 2007) VS. DFUxp IN EUROPE (2007-2008)</i>	0.008	0.006	0.010	0.011	0.008	0.013
<i>Mean difference tests: DFU IN THE US (PRE 2007) VS. DFUxp IN THE US (2007-2008)</i>	0.006	0.004	0.006	0.004	0.010	0.007

All estimated parameters are significant at the 1% level

The test statistics report the p-value of a one-tailed t-test of the hypothesis that the means are equal for the indicated groups.

TABLE IV
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL: ALL BANKS AND DFU_{xa} BANKS IN EUROPE AND IN THE US

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to the changes in volatility of its assets, $\Delta\sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.002** (-26.14)	-0.004** (-34.17)	0.007** (19.83)	0.008** (25.16)
<i>Size (log total assets)</i>	0.013** (14.31)	0.016** (17.90)	-0.015** (-14.51)	-0.011** (16.31)
$\Delta\sigma_V \times DFU_{xa}$ banks Europe	-0.020** (-6.53)	-0.025** (-8.83)	0.019** (6.50)	0.020** (7.28)
<i>Size</i> $\times DFU_{xa}$ banks Europe	0.003 (1.23)	0.001 (1.01)	0.003 (1.23)	0.003 (1.23)
<i>Corruption perception index (10-CPI)</i>	0.008** (3.29)	0.011** (4.88)	0.016** (6.04)	0.008** (3.29)
<i>Market volatility (VIX)</i>	-0.001* (1.93)	-0.001* (2.16)	0.012 (0.27)	0.018 (0.14)
<i>Observations</i>	8,964	8,964	8,964	8,964
<i>Number of banks</i>	1,494	1,494	1,494	1,494
R^2	0.517	0.604	0.685	0.643
<i>US sample</i>				
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.006** (-18.07)	-0.007** (-31.20)	0.009** (18.51)	0.011** (25.14)
<i>Size (log total assets)</i>	0.029** (14.13)	0.024** (17.53)	-0.016** (-11.15)	-0.014** (22.23)
$\Delta\sigma_V \times DFU_{xa}$ banks US	-0.038** (-5.57)	-0.032** (-8.92)	0.024** (3.63)	0.029** (3.97)
<i>Size</i> $\times DFU_{xa}$ banks US	0.002 (1.12)	0.004 (1.25)	0.007 (0.44)	0.003 (0.78)
<i>Corruption perception index (10-CPI)</i>	0.004 (1.18)	0.007 (0.96)	0.010 (0.85)	0.006 (0.72)
<i>Market volatility (VIX)</i>	-0.003** (2.85)	-0.004** (3.49)	0.010 (0.68)	0.012 (0.19)
<i>Observations</i>	2,153	2,153	2,153	2,153
<i>Number of banks</i>	358	358	358	358
R^2	0.693	0.618	0.688	0.715
<i>Test of the differences between the European and the US sample (p-value)</i>				
$\Delta\sigma_V$	0.020	0.018	0.013	0.014
<i>Size (log total assets)</i>	0.004	0.009	0.198	0.032
$\Delta\sigma_V \times DFU_{xa}$ banks US	0.003	0.036	0.013	0.009
<i>Size</i> $\times DFU_{xa}$ banks US	0.002	0.011	0.011	0.396
<i>Corruption perception index (10-CPI)</i>	0.023	0.028	0.021	0.024
<i>Market volatility (VIX)</i>	0.059	0.053	0.061	0.036
* Statistically significant at 5% level				
** Statistically significant at 1% level				

TABLE V.A
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL: ALL BANKS AND DFUxp BANKS (BENEFITING FROM STATE AID) IN EUROPE AND IN THE US

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to changes in the volatility of its assets, $\Delta\sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.003** (-18.31)	-0.005** (-22.51)	0.006** (14.02)	0.007** (33.08)
<i>Size (log total assets)</i>	0.011** (12.24)	0.014** (18.88)	-0.013** (-17.29)	-0.010** (14.25)
$\Delta\sigma_V \times DFUxp$ banks in Europe	-0.009** (-7.12)	-0.012** (-7.31)	0.027** (8.15)	0.029** (6.10)
<i>Corruption perception index (10-CPI)</i>	0.010** (2.98)	0.011** (4.88)	0.016** (6.04)	0.008** (3.29)
<i>Market volatility (VIX)</i>	-0.002* (2.20)	-0.007** (2.96)	0.013 (0.08)	0.011 (0.19)
<i>Observations</i>	8,964	8,964	8,964	8,964
<i>Number of banks</i>	1,494	1,494	1,494	1,494
R^2	0.616	0.594	0.702	0.625
<i>US sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.006** (-17.12)	-0.008** (-28.68)	0.010** (17.27)	0.013** (22.65)
<i>Size (log total assets)</i>	0.025** (16.77)	0.019** (14.31)	-0.018** (-12.72)	-0.017** (25.90)
$\Delta\sigma_V \times DFUxp$ banks in the US	-0.022** (-6.19)	-0.028** (-6.84)	0.033* (2.14)	0.035** (4.42)
<i>Corruption perception index (10-CPI)</i>	0.003 (0.82)	0.005 (0.48)	0.014 (1.12)	0.010 (0.95)
<i>Market volatility (VIX)</i>	-0.006** (3.48)	-0.005** (3.89)	0.014 (0.71)	0.011 (0.28)
<i>Observations</i>	2,153	2,153	2,153	2,153
<i>Number of banks</i>	358	358	358	358
R^2	0.685	0.624	0.603	0.745
<i>Test of the differences between the European and the US sample (p-value)</i>				
$\Delta\sigma_V$	0.016	0.014	0.016	0.014
<i>Size (log total assets)</i>	0.004	0.006	0.004	0.006
$\Delta\sigma_V \times DFUxp$ banks in Europe	0.003	0.002	0.003	0.002
<i>Corruption perception index (10-CPI)</i>	0.002	0.005	0.002	0.005
<i>Market volatility (VIX)</i>	0.005	0.016	0.005	0.016
* Statistically significant at 5% level				
** Statistically significant at 1% level				

TABLE V.B
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL: ALL BANKS AND DFU_{XP} BANKS IN EUROPE AND THE US

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to changes in the volatility of its assets, $\Delta\sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.004** (-14.26)	-0.006** (-21.05)	0.005** (13.04)	0.007** (28.14)
<i>Lambda (Mills ratio)</i>	-0.058* (1.99)	-0.081** (3.93)	-0.028** (10.13)	-0.034** (7.82)
<i>Size (log total assets)</i>	0.010** (11.51)	0.016** (17.23)	-0.011** (-17.50)	-0.013** (-13.85)
$\Delta\sigma_V \times DFU_{XP}$ banks in Europe	-0.009** (-6.14)	-0.013** (-7.18)	0.029** (8.96)	0.025** (5.08)
<i>Corruption perception index (10-CPI)</i>	0.011** (2.08)	0.014** (5.15)	0.013** (6.17)	0.004** (3.22)
<i>Market volatility (VIX)</i>	-0.002* (2.14)	-0.006** (3.17)	0.012 (0.19)	0.014 (0.11)
<i>Observations</i>	8,964	8,964	8,964	8,964
<i>Number of banks</i>	1,494	1,494	1,494	1,494
R^2	0.649	0.629	0.718	0.632
FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_V	0.963** (12.39)	0.815** (7.05)	0.963** (12.39)	0.815** (7.05)
<i>Size (log total assets)</i>	0.013 (1.16)	0.004 (0.96)	0.013 (1.16)	0.004 (0.96)
<i>Corruption perception index (10-CPI)</i>	0.823** (6.28)	0.916** (8.62)	0.823** (6.28)	0.916** (8.62)
<i>Observations</i>	826	826	826	826
<i>Number of DFU_{XA} banks</i>	137	137	137	137
<i>Number of DFU_{XP} banks</i>	43	43	43	43
<i>Log-likelihood</i>	-626.3	-458.5	-626.3	-458.5
<i>Fraction of correct predictions</i>	88.5	90.4	88.5	90.4
<i>US sample</i>				
$\Delta\sigma_V$	-0.007** (-14.06)	-0.24(06)	0.011** (13.08)	0.012** (21.04)
<i>Lambda (Mills ratio)</i>	-0.094** (4.41)	-0.078** (5.13)	-0.028** (6.40)	-0.034** (6.21)
<i>Size (log total assets)</i>	0.028** (15.93)	0.020** (11.10)	-0.016** (-12.13)	-0.013** (-23.03)
$\Delta\sigma_V \times DFU_{XP}$ banks in the US	-0.021** (-7.05)	-0.031** (-7.13)	0.034* (2.10)	0.030** (5.06)
<i>Corruption perception index (10-CPI)</i>	0.005 (0.88)	0.006 (0.51)	0.013 (1.08)	0.009 (0.72)
<i>Market volatility (VIX)</i>	-0.006** (3.20)	-0.007** (4.13)	0.014 (0.62)	0.012 (0.33)
<i>Observations</i>	2,153	2,153	2,153	2,153
<i>Number of banks</i>	358	358	358	358
R^2	0.690	0.645	0.615	0.758
FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_V	0.703** (18.05)	0.626** (12.35)	0.703** (18.05)	0.626** (12.35)
<i>Size (log total assets)</i>	1.624** (6.51)	1.498** (7.18)	1.624** (6.51)	1.498** (7.18)
<i>Corruption perception index (10-CPI)</i>	0.621 (0.44)	0.521 (0.76)	0.621 (0.44)	0.521 (0.76)
<i>Observations</i>	203	203	203	203
<i>Number of DFU_{XA} banks</i>	33	33	33	33
<i>Number of DFU_{XP} banks</i>	22	22	22	22
<i>Log-likelihood</i>	-484.0	-507.2	-484.0	-507.2
<i>Fraction of correct predictions</i>	89.9	88.5	89.9	88.5

<i>Test of the differences between the European and the US sample (p-value)</i>				
$\Delta\sigma_V$	0.015	0.016	0.021	0.012
<i>Size (log total assets)</i>	0.003	0.005	0.007	0.006
$\Delta\sigma_V X DFUxp$ banks	0.005	0.002	0.011	0.004
<i>Corruption perception index (10-CPI)</i>	0.002	0.004	0.041	0.045
<i>Market volatility (VIX)</i>	0.007	0.015	0.596	0.624
* Statistically significant at 5% level				
** Statistically significant at 1% level				

TABLE VI.A
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL: PRE-CRISIS ENVIRONMENT (2003-2006)

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to changes in the volatility of its assets, $\Delta\sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.003** (-29.47)	-0.005** (-33.42)	0.005** (14.24)	0.007** (22.73)
<i>Size (log total assets)</i>	0.015** (12.10)	0.018** (19.81)	-0.018** (-15.46)	-0.014** (13.78)
$\Delta\sigma_V \times DFUxa$ banks Europe	-0.029** (-7.77)	-0.030** (-6.76)	0.024** (4.72)	0.026** (4.84)
<i>Corruption perception index (10-CPI)</i>	0.011** (5.18)	0.005** (4.93)	0.011** (5.91)	0.007** (4.58)
<i>Market volatility (VIX)</i>	-0.001* (2.19)	-0.001** (2.84)	0.018 (0.52)	0.025 (0.27)
<i>Observations</i>	6,156	6,156	6,156	6,156
<i>Number of banks</i>	1,539	1,539	1,539	1,539
R^2	0.517	0.562	0.597	0.534
<i>US sample</i>				
	RV	ML	RV	ML
	$\Delta\sigma_V$	-0.005** (-16.35)	-0.009** (-24.15)	0.011** (12.16)
<i>Size (log total assets)</i>	0.037** (15.56)	0.032** (16.74)	-0.013** (-14.20)	-0.011** (18.26)
$\Delta\sigma_V \times DFUxa$ banks US	-0.041* (-2.21)	-0.035** (-2.19)	0.032** (5.84)	0.034** (3.13)
<i>Corruption perception index (10-CPI)</i>	0.002 (0.77)	0.003 (0.53)	0.012 (0.97)	0.008 (0.68)
<i>Market volatility (VIX)</i>	-0.004** (3.99)	-0.005** (5.18)	0.008 (0.76)	0.018 (0.21)
<i>Observations</i>	1,398	1,398	1,398	1,398
<i>Number of banks</i>	349	349	349	349
R^2	0.584	0.494	0.652	0.626
<i>Test of the differences between the European and the US sample (p-value)</i>				
$\Delta\sigma_V$	0.014	0.011	0.015	0.012
<i>Size (log total assets)</i>	0.002	0.003	0.006	0.026
$\Delta\sigma_V \times DFUxa$ banks Europe	0.003	0.031	0.078	0.003
<i>Corruption perception index (10-CPI)</i>	0.006	0.031	0.362	0.408
<i>Market volatility (VIX)</i>	0.023	0.013	0.014	0.008
* Statistically significant at 5% level				
** Statistically significant at 1% level				

TABLE VI.B
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL: CRISIS ENVIRONMENT (2007-2008)

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to changes in the volatility of its assets, $\Delta\sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level.

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.005** (-20.30)	-0.006** (-31.51)	0.009** (12.51)	0.011** (16.83)
<i>Size (log total assets)</i>	0.015** (16.71)	0.018** (13.95)	-0.014** (-13.38)	-0.009** (10.14)
$\Delta\sigma_V \times DFUxa$ banks Europe	-0.025** (-4.96)	-0.020* (-2.17)	0.033** (6.36)	0.029** (4.57)
<i>Corruption perception index (10-CPI)</i>	0.015** (6.85)	0.009** (4.08)	0.014** (3.63)	0.012** (3.90)
<i>Market volatility (VIX)</i>	-0.004** (3.52)	-0.003** (2.92)	0.025 (0.63)	0.032 (0.44)
<i>Observations</i>	2,808	2,808	2,808	2,808
<i>Number of banks</i>	1,404	1,404	1,404	1,404
R^2	0.501	0.495	0.542	0.512
<i>US sample</i>				
	RV	ML	RV	ML
	$\Delta\sigma_V$	-0.008** (-12.64)	-0.009** (-31.20)	0.012** (14.24)
<i>Size (log total assets)</i>	0.028** (12.41)	0.027** (11.14)	-0.010** (-12.35)	-0.017** (16.51)
$\Delta\sigma_V \times DFUxa$ banks US	-0.024** (-4.50)	-0.023** (8.92)	0.039** (4.27)	0.044** (5.23)
<i>Corruption perception index (10-CPI)</i>	0.001 (0.63)	0.002 (0.32)	0.010 (0.59)	0.006 (0.70)
<i>Market volatility (VIX)</i>	-0.005** (5.02)	-0.004** (4.28)	0.006 (0.44)	0.012 (0.58)
<i>Observations</i>	755	755	755	755
<i>Number of banks</i>	377	377	377	377
R^2	0.602	0.528	0.538	0.586
<i>Test of the differences in $\Delta\sigma_V$ between the European and the US sample (p-value)</i>				
$\Delta\sigma_V$	0.012	0.015	0.013	0.010
<i>Size (log total assets)</i>	0.002	0.003	0.010	0.003
$\Delta\sigma_V \times DFUxa$ banks Europe	0.126	0.037	0.004	0.002
<i>Corruption perception index (10-CPI)</i>	0.002	0.012	0.006	0.007
<i>Market volatility (VIX)</i>	0.586	0.489	0.008	0.004
* Statistically significant at 5% level				
** Statistically significant at 1% level				

TABLE VII
TESTS OF DIFFERENCES BETWEEN PRECRISIS (2003-2006) AND CRISIS YEARS (2007-2008) FOR
THE US AND EUROPE SEPARATELY

The table shows p-values of covariance tests for coefficient differences as well as Chow's F-test of the overall differences between the sub-samples

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	0.023	0.042	0.007	0.012
<i>Size (log total assets)</i>	0.653	0.728	0.046	0.007
$\Delta\sigma_V \times DFUxa$ banks Europe	0.124	0.005	0.006	0.138
<i>Corruption perception index (10-CPI)</i>	0.014	0.011	0.088	0.005
<i>Market volatility (VIX)</i>	0.009	0.016	0.008	0.018
<i>Overall coefficients F-test</i>	0.018	0.013	0.011	0.016
<i>US sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	0.046	0.196	0.140	0.051
<i>Size (log total assets)</i>	0.007	0.006	0.013	0.004
$\Delta\sigma_V \times DFUxa$ banks US	0.003	0.008	0.053	0.009
<i>Corruption perception index (10-CPI)</i>	0.963	0.694	0.121	0.160
<i>Market volatility (VIX)</i>	0.864	0.658	0.134	0.079
<i>Overall coefficients F-test</i>	0.019	0.034	0.038	0.030

TABLE VIII.A
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL: ALL
BANKS AND DFU_{xp} BANKS IN EUROPE AND IN THE US
PRE-CRISIS ENVIRONMENT (2003-2006)

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to changes in the riskiness of its assets, $\Delta \sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta \sigma_V$	-0.003** (-15.64)	-0.004** (-12.59)	0.005** (11.90)	0.004** (21.03)
<i>Lambda (Mills ratio)</i>	-0.061* (1.95)	-0.073** (3.26)	-0.024** (10.03)	-0.038** (7.30)
<i>Size (log total assets)</i>	0.008** (8.41)	0.009** (11.57)	-0.014** (-13.82)	-0.007** (10.13)
<i>$\Delta \sigma_V$ X DFU_{xp} banks in Europe</i>	-0.006** (-4.14)	-0.011** (-7.23)	0.021** (5.63)	0.025** (4.52)
<i>Corruption perception index (10-CPI)</i>	0.005** (4.23)	0.008** (3.48)	0.011** (4.52)	0.014** (5.27)
<i>Market volatility (VIX)</i>	-0.004** (3.94)	-0.005** (2.86)	0.007 (0.28)	0.005 (0.25)
<i>Observations</i>	6,156	6,156	6,156	6,156
<i>Number of banks</i>	1,539	1,539	1,539	1,539
R^2	0.492	0.536	0.559	0.580
FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_V	0.826** (8.13)	0.803** (4.28)	0.826** (8.13)	0.803** (4.28)
<i>Size (log total assets)</i>	0.010 (1.03)	0.002 (0.31)	0.010 (1.03)	0.002 (0.31)
<i>Corruption perception index (10-CPI)</i>	0.426** (4.13)	0.531** (4.26)	0.426** (4.13)	0.531** (4.26)
<i>Observations</i>	534	534	534	534
<i>Number of DFU_{xa} banks</i>	133	133	133	133
<i>Number of DFU_{xp} banks</i>	43	43	43	43
<i>Log-likelihood</i>	-412.8	-469.1	-412.8	-469.1
<i>Fraction of correct predictions</i>	87.4	88.5	87.4	88.5
<i>US sample</i>				
$\Delta \sigma_V$	-0.004** (-11.77)	-0.21(0.7)	0.006** (10.13)	0.011** (14.93)
<i>Lambda (Mills ratio)</i>	-0.074** (4.14)	-0.060** (4.92)	-0.024** (5.31)	-0.031** (5.55)
<i>Size (log total assets)</i>	0.013** (12.65)	0.014* (2.01)	-0.014* (-2.38)	-0.013** (19.04)
<i>$\Delta \sigma_V$ X DFU_{xp} banks in the US</i>	-0.016** (-8.20)	-0.020** (-5.40)	0.030** (2.73)	0.034** (3.81)
<i>Corruption perception index (10-CPI)</i>	0.004 (0.93)	0.003 (0.53)	0.010 (1.52)	0.006 (1.20)
<i>Market volatility (VIX)</i>	-0.007** (3.07)	-0.005** (4.29)	0.007 (0.93)	0.012 (0.32)
<i>Observations</i>	1398	1398	1398	1398
R^2	0.586	0.469	0.626	0.590
FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_V	0.423** (12.31)	0.415** (7.80)	0.423** (12.31)	0.415** (7.80)
<i>Size (log total assets)</i>	0.840** (8.13)	0.902** (4.03)	0.840** (8.13)	0.902** (4.03)
<i>Corruption perception index (10-CPI)</i>	0.403 (0.32)	0.491 (0.86)	0.403 (0.32)	0.491 (0.86)
<i>Observations</i>	128	128	128	128
<i>Number of DFU_{xa} banks</i>	32	32	32	32
<i>Number of DFU_{xp} banks</i>	22	22	22	22
<i>Log-likelihood</i>	-412.7	-477.7	-412.7	-477.7
<i>Fraction of correct predictions</i>	86.8	88.2	86.8	88.2

<i>Test of the differences between the European and the US sample (p-value)</i>				
$\Delta\sigma_Y$	0.025	0.020	0.019	0.016
Size (log total assets)	0.002	0.012	0.006	0.008
$\Delta\sigma_Y$ X DFUxp banks in the US	0.005	0.007	0.026	0.003
Corruption perception index (10-CPI)	0.117	0.016	0.059	0.036
Market volatility (VIX)	0.029	0.014	0.631	0.017
<i>Test of the differences between the selection equations for the European and the US sample (p-value)</i>				
σ_Y	0.002	0.001	0.003	0.004
Size (log total assets)	0.001	0.002	0.001	0.001
Corruption perception index (10-CPI)	0.013	0.019	0.003	0.001
* Statistically significant at 5% level				
** Statistically significant at 1% level				

TABLE VIII.B
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL: ALL
BANKS AND DFU_{xp} BANKS IN EUROPE AND IN THE US
DURING CRISIS YEARS (2007-2008)

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to changes in the riskiness of its assets, $\Delta\sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.004** (-12.03)	-0.006** (-14.28)	0.006** (10.37)	0.007** (24.28)
<i>Lambda (Mills ratio)</i>	-0.052* (1.84)	-0.062** (3.81)	-0.014** (4.53)	-0.035** (7.60)
<i>Size (log total assets)</i>	0.006** (8.21)	0.015** (9.51)	-0.011** (-11.17)	-0.012** (13.68)
<i>$\Delta\sigma_V$ X DFU_{xp} banks in Europe</i>	-0.010** (-7.35)	-0.018* (-2.13)	0.033** (7.08)	0.036** (5.27)
<i>Corruption perception index (10-CPI)</i>	0.007** (6.13)	0.007** (4.27)	0.012** (3.29)	0.018** (3.70)
<i>Market volatility (VIX)</i>	-0.005** (3.23)	-0.004** (4.02)	0.006 (0.25)	0.008 (0.56)
<i>Observations</i>	2,808	2,808	2,808	2,808
<i>Number of banks</i>	1,404	1,404	1,404	1,404
R^2	0.475	0.443	0.518	0.468
FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_V	0.285** (6.01)	0.327** (5.32)	0.285** (6.01)	0.327** (5.32)
<i>Size (log total assets)</i>	0.004 (1.23)	0.005 (0.90)	0.004 (1.23)	0.005 (0.90)
<i>Corruption perception index (10-CPI)</i>	0.415* (1.97)	0.885** (6.17)	0.415* (1.97)	0.885** (6.17)
<i>Observations</i>	292	292	292	292
<i>Number of DFU_{xa} banks</i>	146	146	146	146
<i>Number of DFU_{xp} banks</i>	43	43	43	43
<i>Log-likelihood</i>	-348.3	-435.3	-348.3	-435.3
<i>Fraction of correct predictions</i>	85.2	87.9	85.2	87.9
<i>US sample</i>				
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.007** (-12.16)	-0.009** (-12.57)	0.015** (11.16)	0.013** (12.83)
<i>Lambda (Mills ratio)</i>	-0.046** (6.13)	-0.079** (5.28)	-0.024** (6.32)	-0.036** (5.04)
<i>Size (log total assets)</i>	0.030* (2.16)	0.020** (4.86)	-0.019* (-4.27)	-0.018* (-1.63)
<i>$\Delta\sigma_V$ X DFU_{xp} banks in the US</i>	-0.033** (-7.95)	-0.008** (-3.53)	0.034* (2.08)	0.042** (4.20)
<i>Corruption perception index (10-CPI)</i>	0.007 (0.63)	0.006 (0.63)	0.015 (1.03)	0.010 (0.90)
<i>Market volatility (VIX)</i>	-0.010** (3.38)	-0.012** (4.02)	0.006 (0.20)	0.008 (0.24)
<i>Observations</i>	755	755	755	755
<i>Number of banks</i>	377	377	377	377
R^2	0.572	0.477	0.506	0.547

FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_v	0.982** (4.01)	0.785** (3.68)	0.982** (4.01)	0.785** (3.68)
<i>Size (log total assets)</i>	1.626** (5.13)	1.494** (5.92)	1.626** (5.13)	1.494** (5.92)
<i>Corruption perception index (10-CPI)</i>	0.574 (0.42)	0.546 (0.58)	0.574 (0.42)	0.546 (0.58)
<i>Observations</i>	75	75	75	75
<i>Number of DFUxa banks</i>	37	37	37	37
<i>Number of DFUxp banks</i>	22	22	22	22
<i>Log-likelihood</i>	-390.7	-349.7	-390.7	-349.7
<i>Fraction of correct predictions</i>	86.3	88.7	86.3	88.7
<i>Test of the differences between the European and the US sample (p-value)</i>				
$\Delta\sigma_v$	0.020	0.018	0.015	0.013
<i>Size (log total assets)</i>	0.003	0.006	0.004	0.007
$\Delta\sigma_v \times$ DFUxp banks in the US	0.005	0.003	0.044	0.038
<i>Corruption perception index (10-CPI)</i>	0.705	0.020	0.029	0.027
<i>Market volatility (VIX)</i>	0.013	0.014	0.657	0.266
<i>Test of the differences between the selection equations for the European and the US sample (p-value)</i>				
σ_v	0.001	0.001	0.001	0.001
<i>Size (log total assets)</i>	0.001	0.001	0.001	0.001
<i>Corruption perception index (10-CPI)</i>	0.023	0.006	0.002	0.002
* Statistically significant at 5% level				
** Statistically significant at 1% level				

TABLE IX
TESTS OF THE DIFFERENCES BETWEEN THE PRECRISIS (2003-2006) AND CRISIS
YEARS (2007-2008): DFU_{xp} BANKS IN EUROPE AND THE US

The table show the p-values of covariance tests for coefficient differences as well as Chow's F-test of the overall differences between the subsamples

<i>European sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	0.006	0.012	0.006	0.003
<i>Size (log total assets)</i>	0.011	0.004	0.107	0.052
$\Delta\sigma_V$ X DFU _{xp} banks in Europe	0.008	0.009	0.004	0.006
<i>Corruption perception index (10-CPI)</i>	0.019	0.294	0.013	0.011
<i>Market volatility (VIX)</i>	0.048	0.031	0.002	0.003
<i>Overall coefficients F-test</i>	0.010	0.013	0.008	0.009
<i>US sample</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	0.005	0.011	0.004	0.003
<i>Size (log total assets)</i>	0.003	0.003	0.003	0.004
$\Delta\sigma_V$ X DFU _{xp} banks in the US	0.004	0.005	0.005	0.128
<i>Corruption perception index (10-CPI)</i>	0.043	0.238	0.013	0.029
<i>Market volatility (VIX)</i>	0.031	0.011	0.002	0.003
<i>Overall coefficients F-test</i>	0.008	0.010	0.006	0.008

TABLE X.A
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL:
PORTUGAL AND IRELAND

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to changes in the riskiness of its assets, $\Delta\sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level

<i>Portugal</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.003** (-8.64)	-0.004** (-22.16)	0.011** (10.51)	0.010** (9.87)
<i>Lambda (Mills ratio)</i>	-0.043** (2.98)	-0.071** (5.53)	-0.019** (6.18)	-0.024** (5.96)
<i>Size (log total assets)</i>	0.006** (11.13)	0.008** (13.82)	-0.014** (-9.32)	-0.013** (12.63)
$\Delta\sigma_V \times DFUxp$ banks in Portugal	-0.012** (-3.98)	-0.015** (-5.83)	0.038** (2.94)	0.030** (2.61)
<i>Corruption perception index (10-CPI)</i>	0.010** (3.31)	0.009** (2.31)	0.014** (5.02)	0.011** (3.36)
<i>Market volatility (VIX)</i>	-0.004 (1.03)	-0.003 (1.27)	0.005 (0.33)	0.003 (0.28)
<i>Observations</i>	158	158	158	158
<i>Number of banks</i>	26	26	26	26
R^2	0.403	0.460	0.484	0.520
FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_V	0.423** (5.18)	0.432** (6.15)	0.423** (5.18)	0.432** (6.15)
<i>Size (log total assets)</i>	0.026 (1.63)	0.012 (0.94)	0.026 (1.63)	0.012 (0.94)
<i>Corruption perception index (10-CPI)</i>	0.661* (2.23)	0.891** (6.02)	0.661* (2.23)	0.891** (6.02)
<i>Observations</i>	24	24	24	24
<i>Number of DFUxa banks</i>	4	4	4	4
<i>Number of DFUxp banks</i>	2	2	2	2
<i>Log-likelihood</i>	-326.7	-460.3	-326.7	-460.3
<i>Fraction of correct predictions</i>	86.4	88.4	86.4	88.4
<i>Ireland</i>				
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.002** (-11.01)	-0.002** (-16.50)	0.015** (13.08)	0.018** (14.82)
<i>Lambda (Mills ratio)</i>	-0.039* (2.63)	-0.050** (4.09)	-0.028** (6.08)	-0.042** (9.02)
<i>Size (log total assets)</i>	0.023** (12.60)	0.013** (19.42)	-0.018** (-10.09)	-0.020** (14.37)
$\Delta\sigma_V \times DFUxp$ banks in Ireland	-0.018** (-6.54)	-0.019** (-3.88)	0.049** (3.62)	0.064** (4.03)
<i>Corruption perception index (10-CPI)</i>	0.084** (5.23)	0.041** (5.03)	0.018** (5.21)	0.021** (5.52)
<i>Market volatility (VIX)</i>	-0.006 (1.32)	-0.005 (1.62)	0.016 (0.40)	0.019 (0.28)
<i>Observations</i>	157	157	157	157
<i>Number of banks</i>	25	25	25	25
R^2	0.447	0.416	0.593	0.496

FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_v	0.225** (4.36)	0.212** (5.28)	0.225** (4.36)	0.212** (5.28)
<i>Size (log total assets)</i>	0.096* (2.13)	0.077* (4.82)	0.096* (2.13)	0.077* (4.82)
<i>Corruption perception index (10-CPI)</i>	0.686** (2.91)	0.719** (4.64)	0.686** (2.91)	0.719** (4.64)
<i>Observations</i>	24	24	24	24
<i>Number of DFUxa banks</i>	4	4	4	4
<i>Number of DFUxp banks</i>	3	3	3	3
<i>Log-likelihood</i>	-3263	-412.0	-3263	-412.0
<i>Fraction of correct predictions</i>	83.7	85.3	83.7	85.3
* Statistically significant at 5% level				
** Statistically significant at 1% level				

TABLE X.B
SINGLE-EQUATION ESTIMATES OF THE EFFECTIVENESS OF SAFETY-NET CONTROL:
SPAIN AND ITALY

Fixed-effects panel regressions relating changes in a bank's leverage, $\Delta B/V$, and changes in its fair insurance premium percentage, ΔIPP , to changes in the riskiness of its assets, $\Delta\sigma_V$. B is the face value of bank's debt, including deposits. V is the market value of bank assets. The errors are clustered at the firm level

<i>Spain</i>				
	$\Delta(B/V)$		ΔIPP	
	RV	ML	RV	ML
$\Delta\sigma_V$	-0.004** (-15.04)	-0.005** (-25.18)	0.006** (16.12)	0.008** (16.27)
<i>Lambda (Mills ratio)</i>	-0.021** (4.88)	-0.025** (6.58)	-0.016** (3.31)	-0.014** (4.19)
<i>Size (log total assets)</i>	0.037 (0.85)	0.027 (0.31)	0.012 (0.63)	0.010 (0.40)
$\Delta\sigma_V \times DFUxp$ banks in Spain	-0.018** (-6.03)	-0.020** (-5.13)	0.019** (3.34)	0.023** (4.16)
<i>Corruption perception index (10-CPI)</i>	0.004** (3.18)	0.003** (2.58)	0.007** (5.01)	0.008** (3.14)
<i>Market volatility (VIX)</i>	-0.002 (0.73)	-0.001 (0.34)	0.005 (0.20)	0.004 (0.33)
<i>Observations</i>	531	531	531	531
<i>Number of banks</i>	86	86	86	86
R^2	0.503	0.550	0.519	0.523
FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_V	0.131** (4.56)	0.131** (5.01)	0.131** (4.56)	0.131** (5.01)
<i>Size (log total assets)</i>	0.008 (0.55)	0.009 (0.65)	0.008 (0.55)	0.009 (0.65)
<i>Corruption perception index (10-CPI)</i>	0.285** (2.76)	0.826** (6.04)	0.285** (2.76)	0.826** (6.04)
<i>Observations</i>	52	52	52	52
<i>Number of DFUxa banks</i>	8	8	8	8
<i>Number of DFUxp banks</i>	4	4	4	4
<i>Log-likelihood</i>	-318.5	-401.7	-318.5	-401.7
<i>Fraction of correct predictions</i>	83.7	84.2	83.7	84.2
<i>Italy</i>				
$\Delta\sigma_V$	-0.007** (-10.13)	-0.008** (-15.06)	0.008** (16.67)	0.006** (12.34)
<i>Lambda (Mills ratio)</i>	-0.059* (2.31)	-0.063* (2.13)	-0.052** (4.83)	-0.060** (7.15)
<i>Size (log total assets)</i>	0.032** (13.84)	0.038** (14.13)	-0.029** (-7.15)	-0.010** (14.32)
$\Delta\sigma_V \times DFUxp$ banks in Italy	-0.016** (-7.50)	-0.023** (-8.31)	0.026** (4.77)	0.020** (5.32)
<i>Corruption perception index (10-CPI)</i>	0.008** (4.83)	0.006** (4.94)	0.005** (3.34)	0.006** (4.05)
<i>Market volatility (VIX)</i>	-0.001 (0.30)	-0.002 (0.35)	0.003 (0.62)	0.004 (0.30)
<i>Observations</i>	1236	1236	1236	1236
<i>Number of DFU banks</i>	206	206	206	206
R^2	0.576	0.593	0.580	0.613

FIXED-EFFECTS PROBIT SELECTION MODELS FOR ZERO-ONE BINARY VARIABLES DISTINGUISHING DFU BANKS BENEFITING FROM STATE AID (1) FROM THE REST OF DFU BANKS (0)				
σ_v	0.721** (7.58)	0.850** (4.16)	0.721** (7.58)	0.850** (4.16)
<i>Size (log total assets)</i>	0.013* (1.96)	0.010** (3.05)	0.013* (1.96)	0.010** (3.05)
<i>Corruption perception index (10-CPI)</i>	0.421** (3.75)	0.478** (4.01)	0.421** (3.75)	0.478** (4.01)
<i>Observations</i>	120	120	120	120
<i>Number of DFUxa banks</i>	20	20	20	20
<i>Number of DFUxp banks</i>	6	6	6	6
<i>Log-likelihood</i>	-360.5	-390.0	-360.5	-390.0
<i>Fraction of correct predictions</i>	85.0	86.5	85.0	86.5

TABLE X.C
TESTS OF THE DIFFERENCES BETWEEN PORTUGAL, IRELAND,
SPAIN AND ITALY DFUxp BANKS
p-values in parentheses

<i>Test of the differences between Portugal and Ireland</i>				
$\Delta\sigma_V$	0.057	0.046	0.026	0.022
Size (log total assets)	0.016	0.010	0.005	0.004
$\Delta\sigma_V \times DFUxp$ banks in Spain	0.002	0.006	0.030	0.023
Corruption perception index (10-CPI)	0.006	0.002	0.028	0.019
Market volatility (VIX)	0.014	0.030	0.016	0.008
<i>Test of the differences in the selection equation between Portugal and Ireland</i>				
σ_V	0.004	0.003	0.003	0.002
Size (log total assets)	0.002	0.001	0.002	0.010
Corruption perception index (10-CPI)	0.006	0.023	0.003	0.003
<i>Test of the differences between Portugal and Spain</i>				
$\Delta\sigma_V$	0.326	0.286	0.015	0.039
Lambda (Mills ratio)	0.003	0.004	0.042	0.026
Size (log total assets)	0.002	0.002	0.123	0.086
$\Delta\sigma_V \times DFUxp$ banks in Italy	0.024	0.027	0.006	0.007
Corruption perception index (10-CPI)	0.028	0.031	0.001	0.001
Market volatility (VIX)	0.263	0.385	0.698	0.582
<i>Test of the differences in the selection equation between Portugal and Spain</i>				
σ_V	0.002	0.002	0.001	0.003
Size (log total assets)	0.003	0.004	0.638	0.125
Corruption perception index (10-CPI)	0.005	0.042	0.029	0.045
<i>Test of the differences between Portugal and Italy</i>				
$\Delta\sigma_V$	0.005	0.004	0.035	0.028
Size (log total assets)	0.008	0.005	0.006	0.004
$\Delta\sigma_V \times DFUxp$ banks in Spain	0.003	0.002	0.010	0.013
Corruption perception index (10-CPI)	0.235	0.094	0.012	0.014
Market volatility (VIX)	0.131	0.122	0.193	0.281
<i>Test of the differences in the selection equation between Portugal and Italy</i>				
$\Delta\sigma_V$	0.002	0.003	0.003	0.002
Size (log total assets)	0.008	0.032	0.002	0.001
Corruption perception index (10-CPI)	0.005	0.004	0.004	0.003
<i>Test of the differences between Ireland and Spain</i>				
$\Delta\sigma_V$	0.043	0.034	0.013	0.011
Size (log total assets)	0.006	0.005	0.003	0.002
$\Delta\sigma_V \times DFUxp$ banks in Spain	0.004	0.002	0.008	0.010
Corruption perception index (10-CPI)	0.001	0.001	0.006	0.004
Market volatility (VIX)	0.028	0.023	0.004	0.006
<i>Test of the differences in the selection equation between Ireland and Spain</i>				
$\Delta\sigma_V$	0.005	0.006	0.002	0.002
Size (log total assets)	0.001	0.001	0.003	0.002
Corruption perception index (10-CPI)	0.002	0.001	0.001	0.001
<i>Test of the differences between Ireland and Italy</i>				
$\Delta\sigma_V$	0.006	0.008	0.013	0.008
Size (log total assets)	0.013	0.016	0.005	0.003
$\Delta\sigma_V \times DFUxp$ banks in Spain	0.026	0.031	0.007	0.009
Corruption perception index (10-CPI)	0.018	0.014	0.012	0.013
Market volatility (VIX)	0.026	0.034	0.015	0.019
<i>Test of the differences in the selection equation between Ireland and Italy</i>				
$\Delta\sigma_V$	0.001	0.001	0.001	0.001
Size (log total assets)	0.001	0.001	0.010	0.028
Corruption perception index (10-CPI)	0.003	0.002	0.007	0.006
<i>Test of the differences between Spain and Italy</i>				
$\Delta\sigma_V$	0.026	0.031	0.008	0.029
Size (log total assets)	0.005	0.008	0.004	0.003
$\Delta\sigma_V \times DFUxp$ banks in Spain	0.236	0.184	0.026	0.021
Corruption perception index (10-CPI)	0.013	0.010	0.236	0.208
Market volatility (VIX)	0.650	0.635	0.336	0.381
<i>Test of the differences in the selection equation between Spain and Italy</i>				
$\Delta\sigma_V$	0.001	0.001	0.001	0.001
Size (log total assets)	0.089	0.136	0.007	0.015
Corruption perception index (10-CPI)	0.012	0.009	0.003	0.002

TABLE XI

TESTS OF COEFFICIENT DIFFERENCES BETWEEN DMS MODELS ESTIMATED FOR PRECRISIS YEARS (2003-2006) AND CRISIS YEARS (2007-2008): DFU BANKS RECEIVING STATE AID IN PORTUGAL, IRELAND, SPAIN AND ITALY VS. DFU BANKS IN THE EUROPEAN SAMPLE

The table show the p-values of the tests for coefficient differences as well as Chow's F-test of the overall differences between the subsamples

<i>Portugal, Ireland, Spain and Italy vs. European sample (PRECRISIS)</i>				
	RV	ML	RV	ML
$\Delta\sigma_y$	0.014	0.017	0.005	0.006
<i>Size (log total assets)</i>	0.009	0.008	0.128	0.682
$\Delta\sigma_y \times$ DFU banks	0.013	0.010	0.016	0.014
<i>Corruption perception index (10-CPI)</i>	0.143	0.138	0.231	0.228
<i>Market volatility (VIX)</i>	-	-	-	-
<i>Overall coefficients F-test</i>	0.006	0.010	0.011	0.013
<i>Portugal, Ireland, Spain and Italy vs. European sample (CRISIS)</i>				
	RV	ML	RV	ML
$\Delta\sigma_y$	0.005	0.004	0.005	0.006
<i>Size (log total assets)</i>	0.008	0.010	0.014	0.026
$\Delta\sigma_y \times$ DFU banks	0.403	0.396	0.002	0.012
<i>Corruption perception index (10-CPI)</i>	0.010	0.013	0.036	0.054
<i>Market volatility (VIX)</i>	-	-	-	-
<i>Overall coefficients F-test</i>	0.013	0.011	0.014	0.016

APPENDIX: IDENTIFYING DFU_{xp} BANKS. MAIN SOURCES

The following list of links includes references to all the guarantee schemes and capital injection/recapitalization schemes in the sample countries as well as list of banks receiving the State aid:

- Austria:

Capital injections:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/928&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1989&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1998&format=HTML&aged=0&language=EN&guiLanguage=en>

Guarantee schemes:

http://www.parlament.gv.at/PAKT/SW/index.shtml?jsMode=&xdocumentUri=&GP=XXIV&SW=82&SUCH=&listeId=119&LISTE=Anzeigen&FBEZ=FP_019
http://www.parlament.gv.at/PAKT/SW/index.shtml?jsMode=&xdocumentUri=&GP=XXIII&SW=82&SUCH=&listeId=119&LISTE=Anzeigen&FBEZ=FP_019

- Belgium:

Guarantee schemes:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1745&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1745&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/399&format=HTML&aged=0&language=EN&guiLanguage=en>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1746&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1884&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/743&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/2033&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/254&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1063&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1730&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/592&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/201&format=HTML&aged=0&language=EN&guiLanguage=en>

- Denmark

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1222&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1633&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1483&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/819&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/1374&format=HTML&aged=0&language=EN&guiLanguage=en>

Guarantee schemes:

<http://www.finansraadet.dk/bankkunde/det-private-beredskab/medlemmer-i-det-private-beredskab.aspx>

- Finland:

Guarantee schemes:

<http://www.eduskunta.fi/valtiopaivaasiat/he+255/2009>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1705&format=HTML&aged=0&language=EN&guiLanguage=en>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/82&format=HTML&aged=0&language=EN&guiLanguage=en>

- Luxembourg:

Capital injections/recapitalization

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1107&format=HTML&aged=0&language=EN&guiLanguage=en>

- Netherlands:

Guarantee schemes:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1610&format=HTML&aged=0&language=EN&guiLanguage=en>

http://www.dnb.nl/binaries/The%20Financial%20Crisis%20in%20the%20Netherlands_tcm46-224708.pdf

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/514&format=HTML&aged=0&language=EN&guiLanguage=en>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1699&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1951&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/138&format=HTML&aged=0&language=EN&guiLanguage=en>

- Portugal:

Guarantee schemes:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/400&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://www.mayerbrown.com/publications/article.asp?id=7856&nid=6>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/818&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://www.mayerbrown.com/publications/article.asp?id=7856&nid=6>

- Sweden:

Guarantee schemes:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1600&format=HTML&aged=0&language=EN&guiLanguage=en>

https://www.riksgalden.se/templates/RGK_Templates/TwoColumnPage_17104.aspx

<https://www.riksgalden.se/Dokument/Riksgaldskontoret%20och%20statsfinanserna/Finansiell%20stabilitet/Garantiprogrammet/Information%20garantiprogram/General%20Information%20concerning%20the%20Swedish%20Guarantee%20Scheme.pdf>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1600&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://www.sweden.gov.se/sb/d/10213/a/120252>

https://www.riksgalden.se/templates/RGK_Templates/TwoColumnPage_17537.aspx

- Ireland:

Capital injections/recapitalization:

<http://www.nama.ie/>

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/50&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/744&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/1765&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/1765&format=HTML&aged=0&language=EN&guiLanguage=en>

Guarantee schemes:

<http://www.financialregulator.ie/industry-sectors/credit-institutions/Documents/List%20of%20covered%20institutions.pdf>

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/1154&format=HTML&aged=0&language=EN&guiLanguage=en>

- United Kingdom:

Capital injections/recapitalization/Guarantee schemes:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1496&format=HTML&aged=0&language=EN&guiLanguage=en>

<http://www.mondaq.com/article.asp?articleid=87924>

http://www.dmo.gov.uk/index.aspx?page=CGS/ABS_about

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1600&format=HTML&aged=0&language=EN&guiLanguage=en>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1437&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1728&format=HTML&aged=0&language=EN&guiLanguage=en/1728&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1915&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/47&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/48&format=HTML&aged=0&language=EN&guiLanguage=en>

- Spain:

Guarantee schemes:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1630&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://www.fondoaaf.es/SP/Subastas.html>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/2049&format=HTML&aged=0&language=EN&guiLanguage=en>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/70&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/855&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/10/1479&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://www.frob.es/notas/notas.html>

- France:

Guarantee schemes and capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1609&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1609&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://www.hoganlovells.co.uk/files/Publication/e05cef7f-fc2a-4711-82a5-c7c53359443a/Presentation/PublicationAttachment/3ac27fc0-b167-4418-b2e0-cb0726624a60/res9597539E97B647948DBE56CBADB62D80.pdf>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/722&format=HTML&aged=0&language=EN&guiLanguage=en>

- Italy:

Guarantee schemes:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/1706&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://www.mayerbrown.com/publications/article.asp?id=7852&nid=6>

Capital injections/recapitalization:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/2059&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://www.mayerbrown.com/publications/article.asp?id=7852&nid=6>

- Germany:

Capital injections:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/722&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/722&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/722&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/114&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/114&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/114&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/854&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/1058&format=HTML&aged=0&language=EN&guiLanguage=en>

Guarantee schemes:

<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/2056&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/08/2055&format=HTML&aged=0&language=EN&guiLanguage=en>
<http://europa.eu/rapid/pressReleasesAction.do?reference=IP/09/114&format=HTML&aged=0&language=EN&guiLanguage=en>

- United States:

Asset Guarantee Program:

<http://www.treasury.gov/initiatives/financial-stability/investment-programs/agp/Pages/assetguaranteeprogram.aspx>

Capital Assistance Program:

<http://www.treasury.gov/initiatives/financial-stability/investment-programs/cap/Pages/capitalassistance.aspx>

Capital Purchase Program:

<http://www.treasury.gov/initiatives/financial-stability/investment-programs/cpp/Pages/capitalpurchaseprogram.aspx>

<http://money.cnn.com/news/specials/storysupplement/bankbailout/>

<http://bailout.propublica.org/list/index>