



LSE

Macroprudential Stress Tests and Policies: Searching for Robust and Implementable Frameworks

R. Anderson, C. Baba, J. Danielsson, U. Das, H. Kang and M. Segoviano

Financial Crises: Predictability, Causes and Consequences

April 10, 2018

Systemic Risk Centre, London School of Economics



The views expressed in this report are those of the authors and do not necessarily represent the views of the IMF, its Executive Board, or IMF Management



Introduction

Collaboration between the Monetary and Capital Markets Department (MCM) of the IMF and the Systemic Risk Centre (SRC) of London School of Economics (LSE)

Report was prepared for the MCM-SRC symposium “Macroprudential Stress Test and Policies” held at the IMF, Washington DC, December 2016

Official sector

Alex Brazier (Bank of England)
Jill Cetina (Office of Financial Research)
Ian Christensen (Bank of Canada)
Alan Elizondo (Banco de México)
Cho Hoi Hui (Hong Kong Monetary Authority)
Malcolm Knight (SRC-LSE)
Hitoshi Mio (Bank of Japan)
Deepak Mohanty (Reserve Bank of India)
Sergio Nicoletti-Altamari (European Central Bank)

Academics

Rama Cont (Imperial College)
Charles Goodhart (LSE)
Itay Goldstein (Wharton School)
Casper de Vries (Erasmus University of Rotterdam)



MACROPRUDENTIAL STRESS TESTS AND POLICIES: SEARCHING FOR ROBUST AND IMPLEMENTABLE FRAMEWORKS

Ron Anderson, Chikako Baba, Jon Danielsson, Udaibir S. Das, Heedon Kang and Miguel Segoviano

February 2018



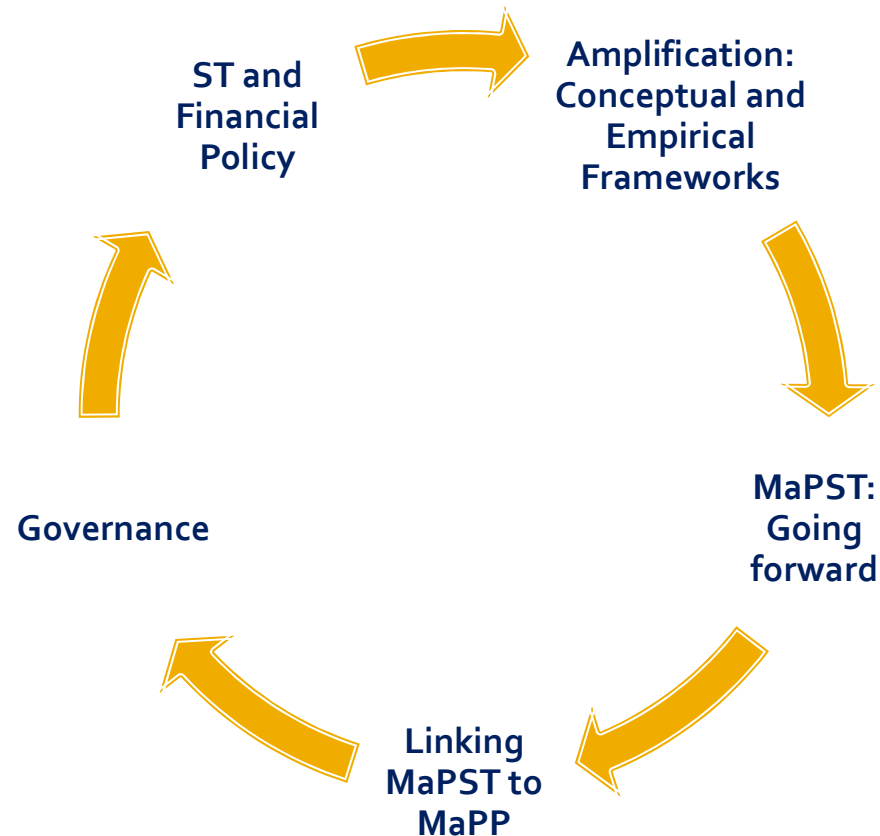
Objective

LSE

Present state-of-the-art MaPST methodologies discussing modelling and implementation challenges;

Provide a **roadmap for future research** and **practical implementations** in stress testing and;

Guide authorities on the **use of MaPST to support macroprudential tool calibration**.





ST and Amplification Mechanisms



- Most stress testing is microprudential, focusing on **individual institutions** and their **resiliency to exogenous** shocks.
- But almost all stress events and crises are caused by **endogenous risk** — the interaction of all market participants in equilibrium;
- Thus, need to account for **amplification mechanisms** due to the **interaction** between the variety of **financial institutions and markets**



Why do we care of MaPSTs?

LSE

MaPSTs are beginning to play an increasingly major role in financial sector policymaking.

- Losses that have the potential to **magnify moderate exogenous shocks** into **substantial negative financial outcomes** with **significant welfare losses**.

A properly designed MaPST can generate valuable information for policymakers.

- Provide forward-looking quantitative assessment of the **resilience of individual banks and financial system as a whole**
- **Inform the use/calibration of relevant macroprudential policy instruments.**
- Generate **useful information for risk management and decision making processes** in periods of financial distress
- Contribute to the **design/improvement of recovery and resolution frameworks.**



Challenges SR quantification: Definition

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

Initial Interpretations of SR

Direct Contagion

Indirect Contagion

Generalized shocks.

Bartholomew & Whalen (1995).

Relationship between the financial system and the real economy.

Mishkin (1995), Bartholomew & Whalen (1995).

Domino effects.

BIS (1994), Kaufman (1995)

However DE do not seem to provide the full explanation.

Adrian and Shin (2008)



Challenges SR quantification: Definition

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

Initial Interpretations of SR

Direct Contagion

Indirect Contagion

Generalized shocks.

Bartholomew & Whalen (1995).

Relationship between the financial system and the real economy.

Mishkin (1995), Bartholomew & Whalen (1995).

Domino effects.

BIS (1994), Kaufman (1995)

However DE do not seem to provide the full explanation.

Adrian and Shin (2008)



Challenges SR quantification: Definition

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

Initial Interpretations of SR

Direct Contagion

Indirect Contagion

Amplification Mechanisms

Fire sales in financial markets.

Exposures to common risk factors

Collateralized agreements. Shleifer and Vishny (2011).

Interactions across Banks and Non-banks. Khandani and Lo (2011), Cortes et al, (2017).

Illiquidity spirals. Brunnermeier and Pedersen (2009).

Deleveraging. Greenwood, et al. (2015)., Cont and Schaanning (2016) .



Challenges SR quantification: Definition

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

Initial Interpretations of SR

Direct Contagion

Indirect Contagion

Amplification Mechanisms

Fire sales in financial markets.

Exposures to common risk factors

Information Asymmetry Channel

Collateralized agreements. Shleifer and Vishny (2011).

Interactions across Banks and Non-banks. Khandani and Lo (2011), Cortes et al, (2017).

Illiquidity spirals. Brunnermeier and Pedersen (2009).

Deleveraging. Greenwood, et al. (2015)., Cont and Schaanning (2016) .

I-A key source of bank runs. Jacklin and Bhattacharya (1988), Khandani and Lo (2011).

Under high uncertainty, the impact of I-A becomes more severe. Kapadia, et al. (2012), Khandani and Lo (2011)

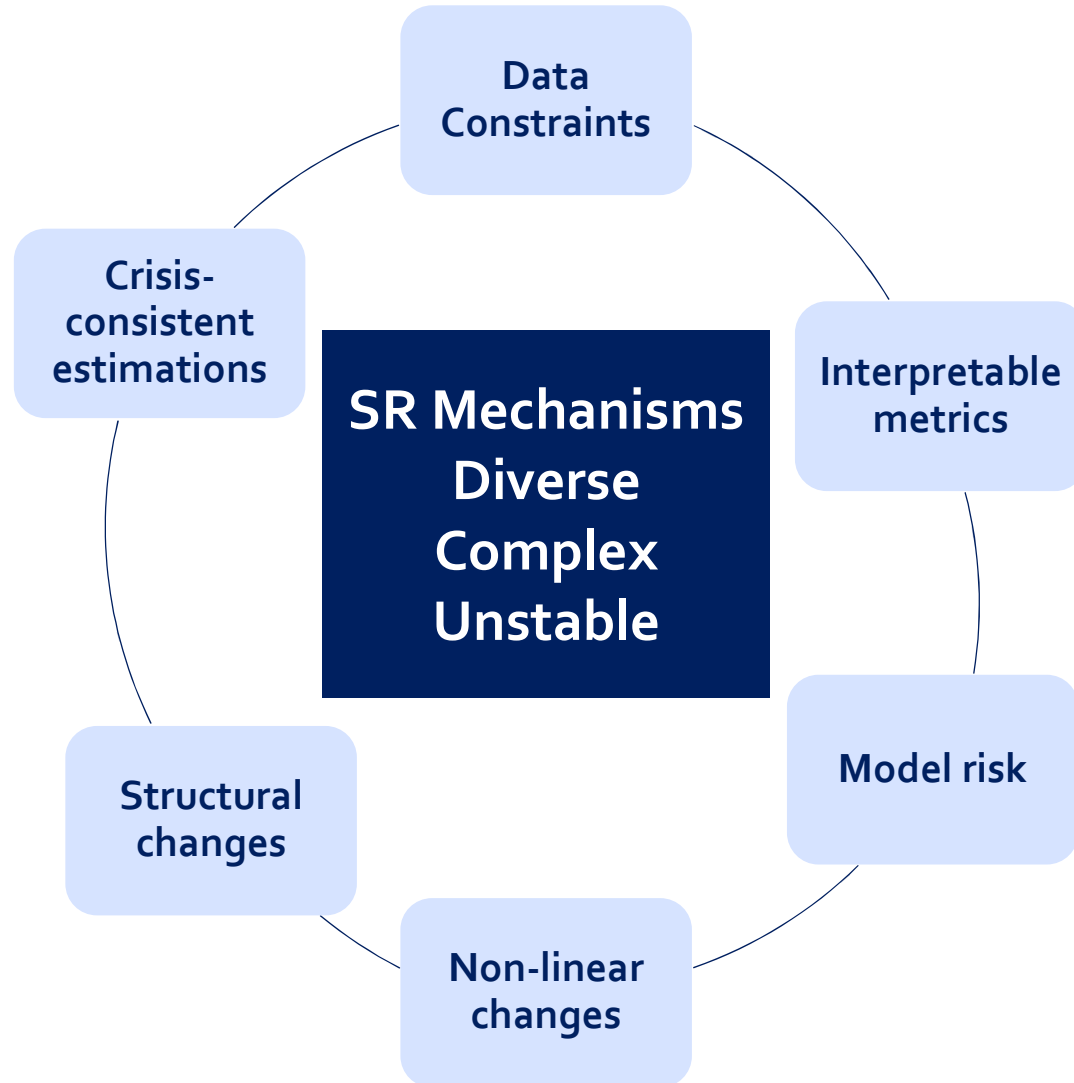
Financial Imbalances
Minsky (1992)
(Adrian, Covitz, Liang, 2013)



Challenges SR quantification: Implementation



Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers





SR quantification: Modeling Approaches



Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

REDUCED-FORM MODELS

Pros	Cons
<p><i>Infer from market data the effect of agents' behavior</i></p> <ul style="list-style-type: none">• Publicly available data• Capture all possible channels accounted by markets• No assumptions on agents' behaviors/market structures• Frequent updating	<ul style="list-style-type: none">• Market data maybe "noisy"• No information on mechanisms• Difficult to embed into stress tests

STRUCTURAL/SIMULATED MODELS

Pros	Cons
<p><i>Explicitly model agents' behavior</i></p> <ul style="list-style-type: none">• Identification of Specific amplification channels• Rooted in theory	<ul style="list-style-type: none">• Limited sets of amplification mechanisms• Complex• Need granular data• Difficult to calibrate

No model or data are completely satisfactory



Encompassing Frameworks

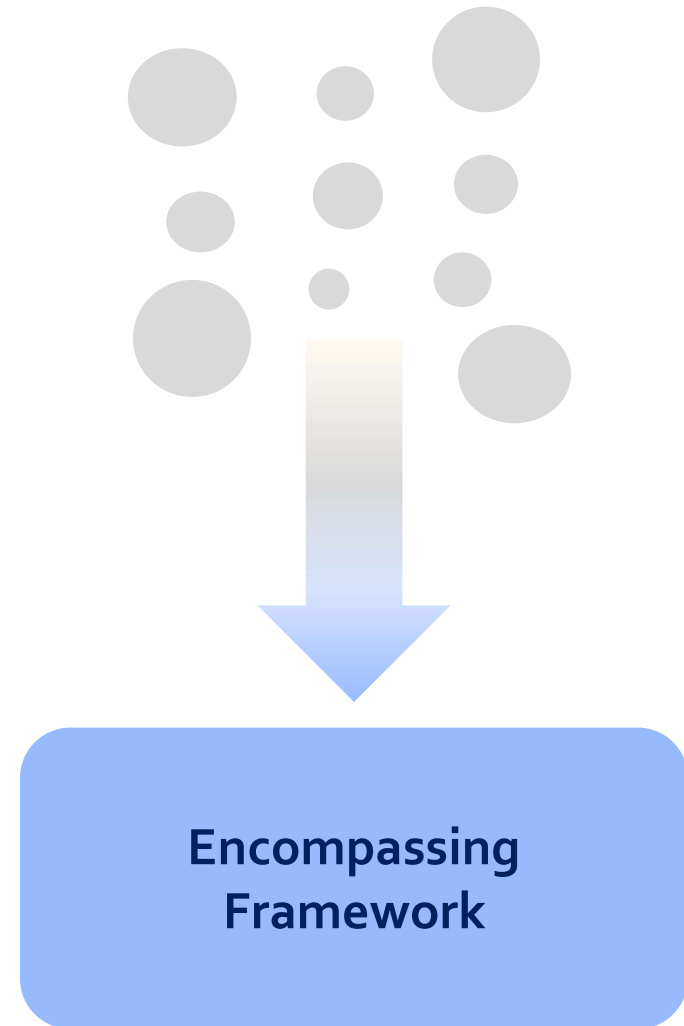
LSE

Systemic Risk Challenges to Modeling Systemic Risk Encompassing Frameworks IMF-EF

No data or model is completely satisfactory for capturing SRA mechanisms

We should try to capture the **best elements of a variety of approaches**

Flexible, yet organized approaches to combining separate analyzes





Encompassing Frameworks



Systemic Risk Challenges to Modeling Systemic Risk Encompassing Frameworks IMF-EF

Cornerstone Benefits of
Assessments of Risk across
Encompassing Frameworks
Heterogeneous Systems

Transferable frameworks
Advance analysis cooperatively
using diverse sets of data and
methods

Reduced Risk
of Model Error

Improved
Assessments

Complementary
Perspectives on Risk

Frameworks **implemented with** a combination of
publicly available and supervisory-based data and
embed **diverse types of methods.**

Fund staff often work under
highly restrictive data constraints, especially for
SRA mechanisms
Need to analyze **heterogeneous financial markets**



Microprudential ST

First order effects of adverse scenarios on individual entities

Diverse methods: ST implemented by the IMF (workbox), National authorities, Firms, jointly

Combination of data: Publicly available, supervisory

SRA Losses

Multivariate perspective of financial system

"Crisis consistent conditional losses" based on markets' perceptions

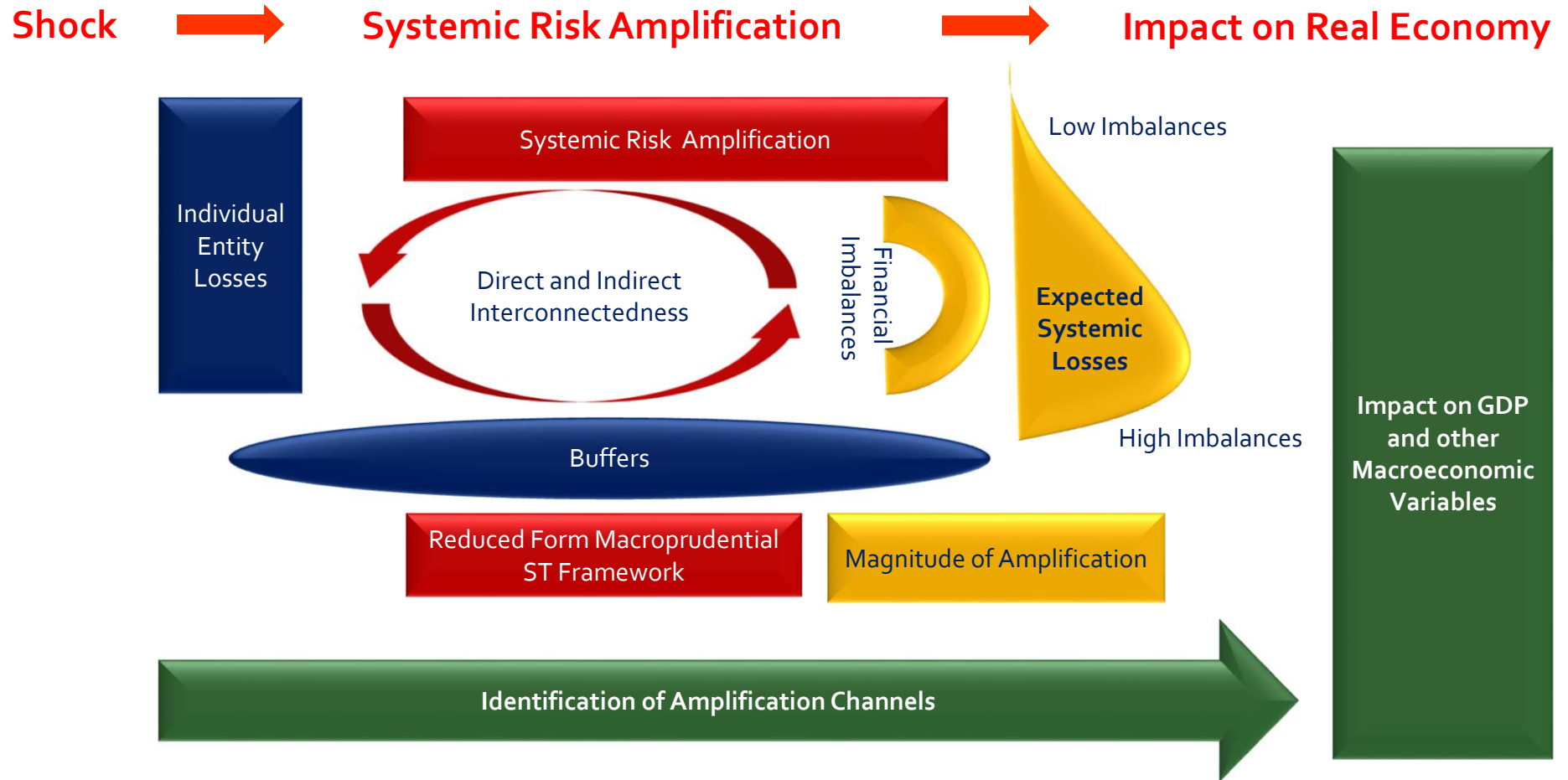
Publicly available data

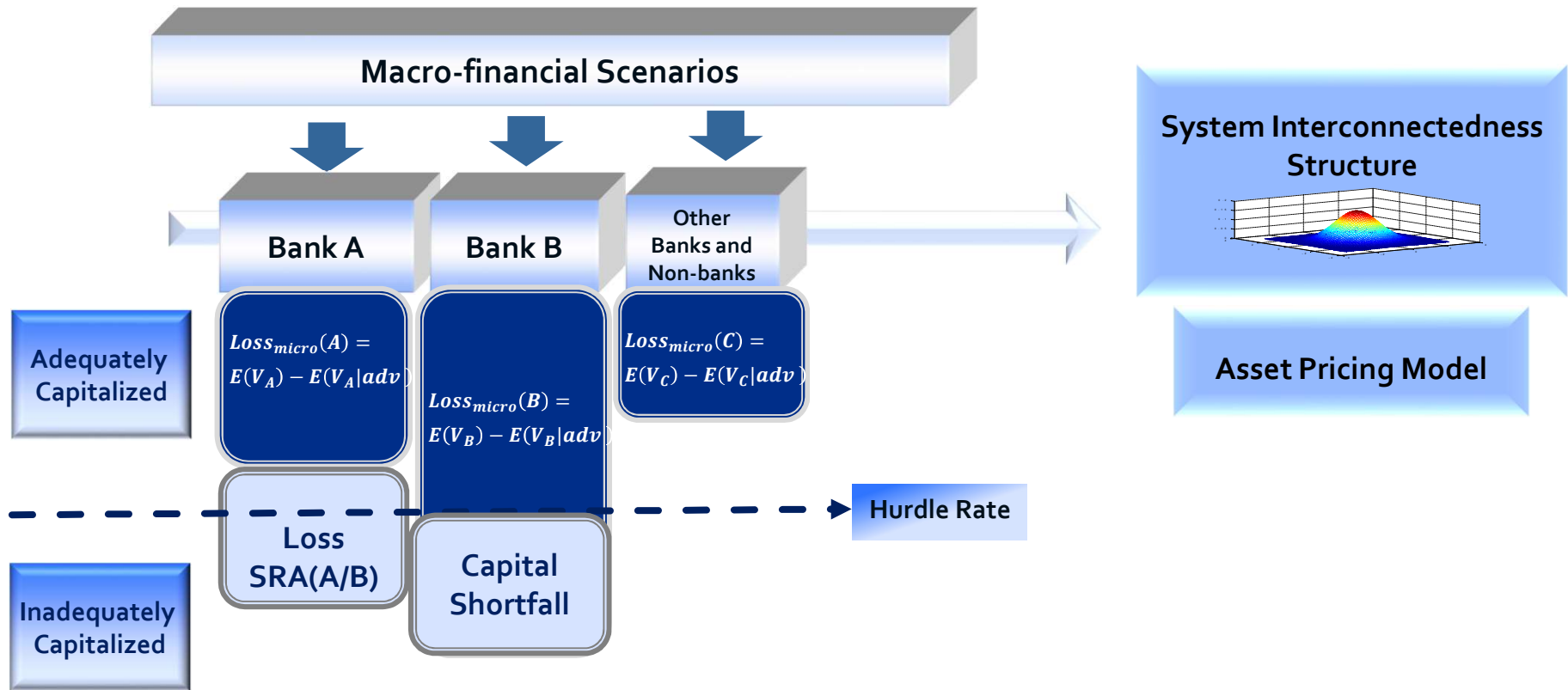


Conceptual Framework: Systemic Risk Assessment

LSE

Challenges to Systemic Risk Modeling **Reduced-Form Macroprudential ST** An example Use for policy makers





Systemic Risk Losses (SR)

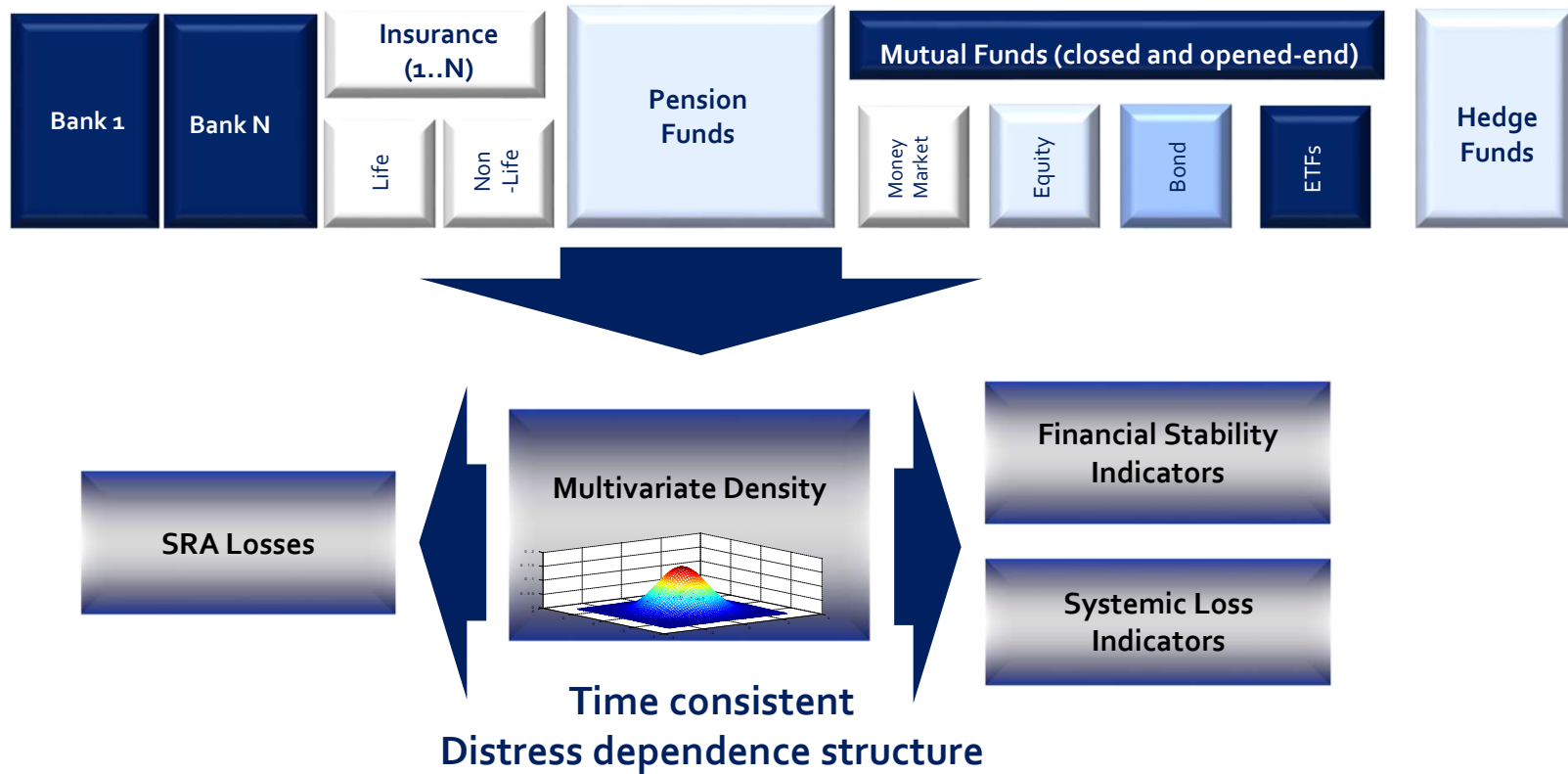
Expected losses given the realization of a given event: $Loss_{SR}(Ai|S) = E(V_{Ai}|adv) - E(V_{Ai}|adv \cap S)$



IMF EF Multivariate Perspective



Systemic Risk Challenges to Modeling Systemic Risk Encompassing Frameworks IMF-EF





Characterization

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

MicroST Loss. Difference between the value of bank A in normal times, and its value under an adverse macroeconomic scenario:

$$Loss_{micro}(A) = E(V_A) - E(V_A|adv) ;$$

SR Loss. Assuming the realization of a given financial contagion event S

$$Loss_{SR}(A|S) = E(V_A|adv) - E(V_A|adv \cap S) ;$$

Total Loss. Assuming the realization of a the financial event S

$$\begin{aligned} Loss_{TS}(A|S) &= Loss_{micro}(A) + Loss_{SR}(A|S) \\ &= E(V_A) - E(V_A|adv \cap S) \end{aligned}$$



SRA Loss: Decomposition

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

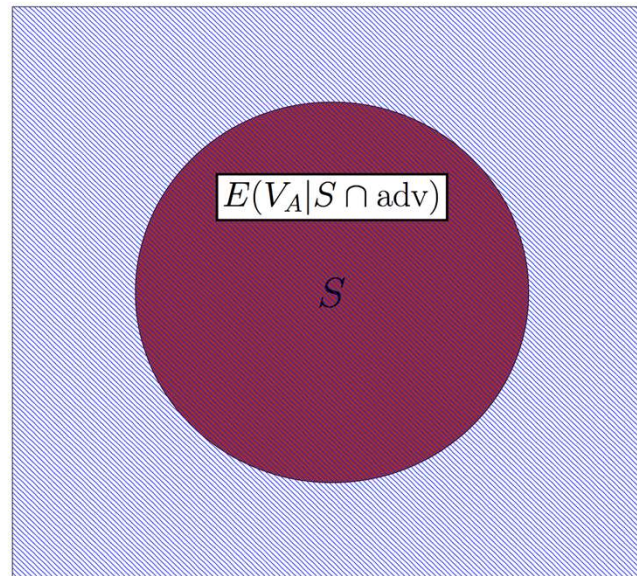
- The SR Loss accounts for all the potential connections across all entities
- A high SR Loss (A/B) does *not* necessarily mean that there is a strong straight connection between A and B.
- The contagion path may include another bank, which is strongly connected to A and/or B and explains the high conditional loss of A/B.
- Using the law of total expectations, we can **identify the connecting entities** between two given entities.



Identification of the SR loss in a Venn Diagram

LSE

Challenges to Systemic Risk Modeling **Reduced-Form Macroprudential ST** An example Use for policy makers



MicroST Loss of a given bank.

Difference between its value in normal times and its value in the adverse M.S.;
This state of nature is represented by the **hatched rectangle in the Figure**.

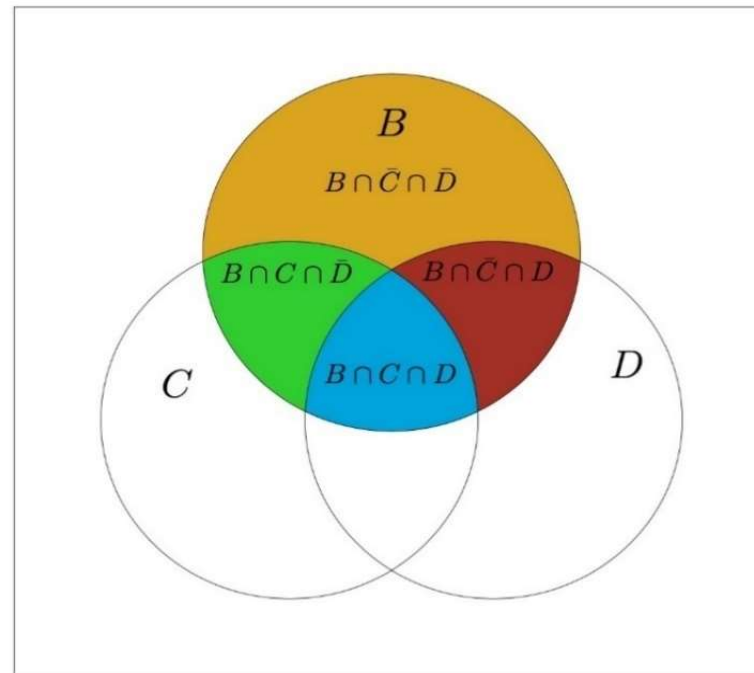
SR Loss.

Difference between the value of bank assuming an adverse M.S., and its value assuming an adverse M.S. **and** the realization of the event S.

The event S is represented by the dark-circled area in the Figure 1.

Decomposing the SR Loss, we can quantify the **likelihood** and **intensity** of “contagion” events.

$$\begin{aligned}
 &Loss_{SR}(A|B) \\
 &= P(B \cap C \cap D|B)Loss_{SR}(A|B \cap C \cap D) + P(B \cap \bar{C} \cap D|B)Loss_{SR}(A|B \cap \bar{C} \cap D) \\
 &+ P(B \cap C \cap \bar{D}|B)Loss_{SR}(A|B \cap C \cap \bar{D}) + P(B \cap \bar{C} \cap \bar{D}|B)Loss_{SR}(A|B \cap \bar{C} \cap \bar{D})
 \end{aligned}$$



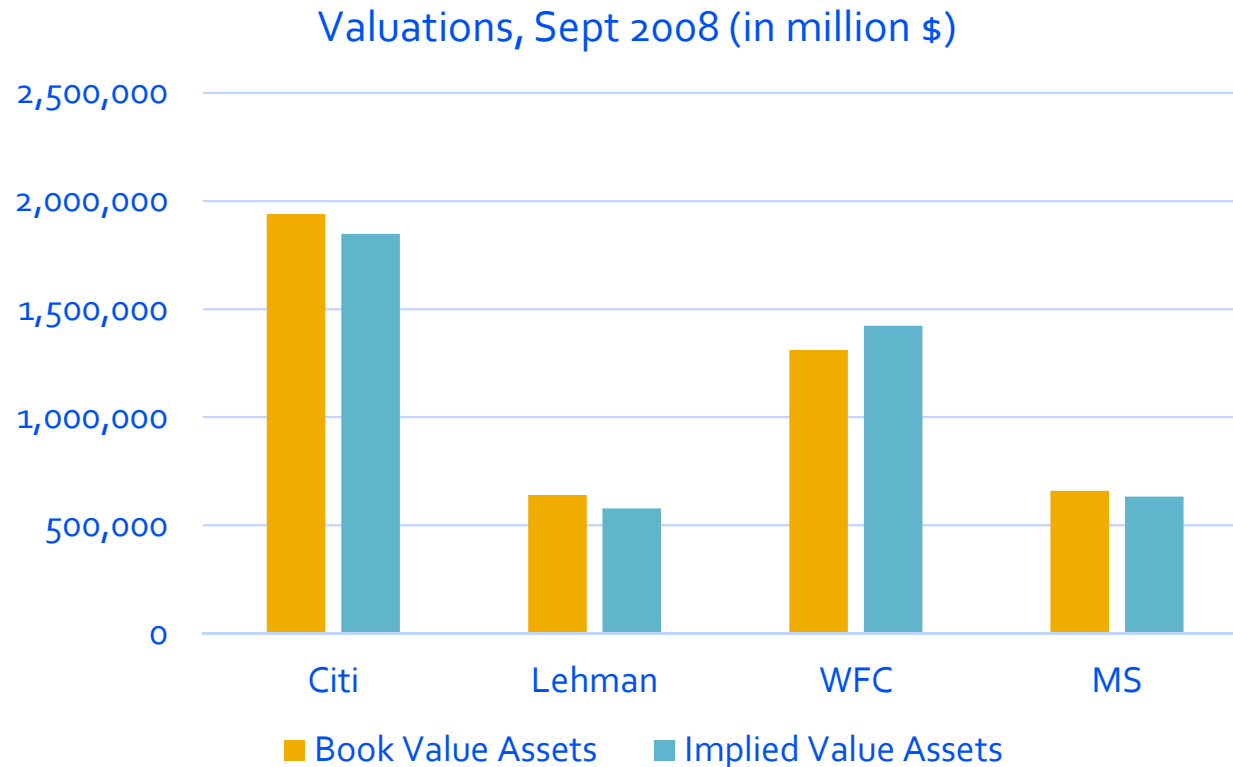


Consistency Checks



Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

Comparing TA with asset pricing model estimate of expected asset values





Results



Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

Conditional losses are increasing in the size of the defaulting set

Conditional Loss for Citi (Sept 2008, in million \$)



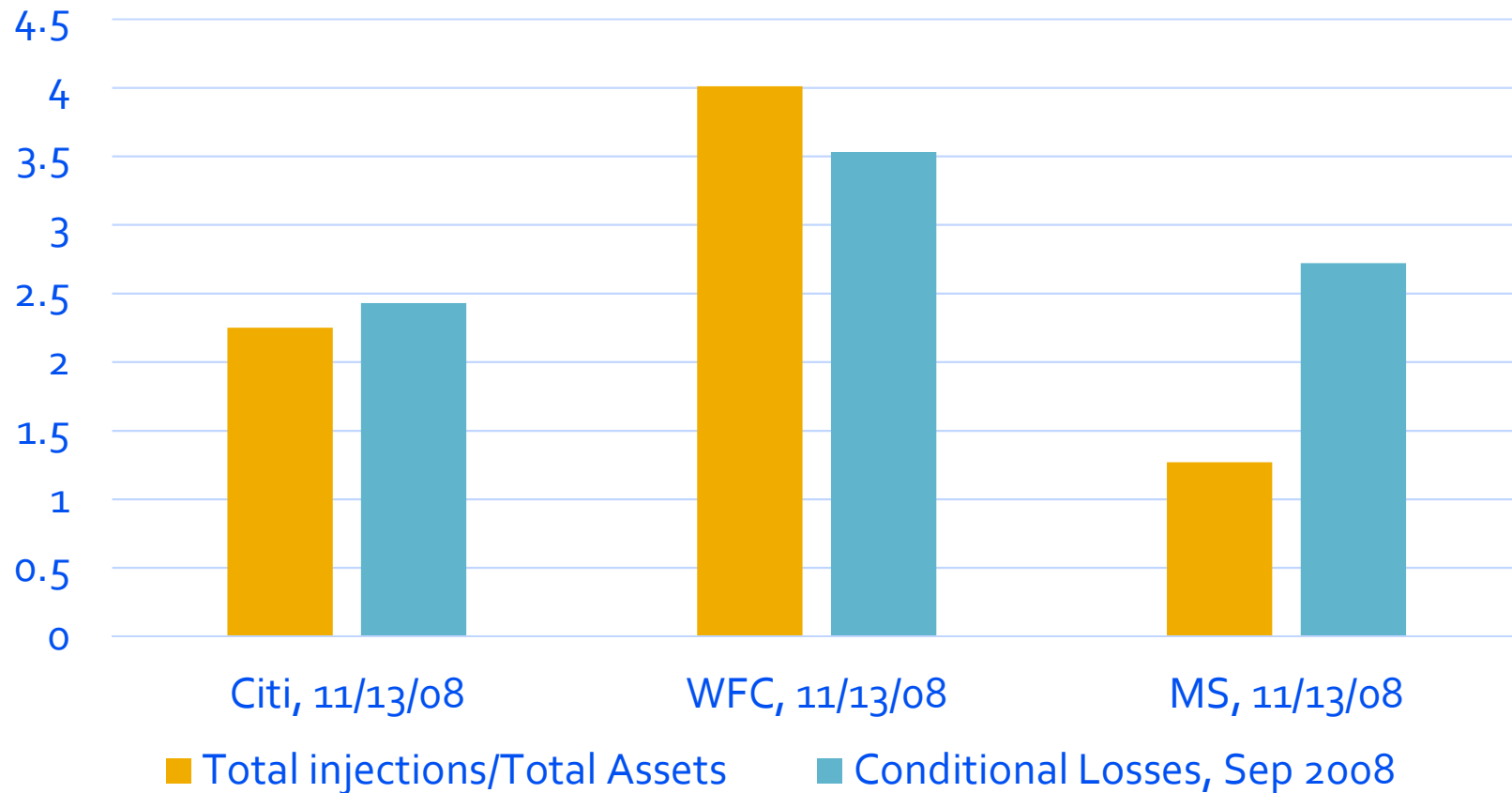


Consistency check: Conditional Losses vs Government Injection



Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST **An example** Use for policy makers

Capital injections and Losses conditional on Lehman default



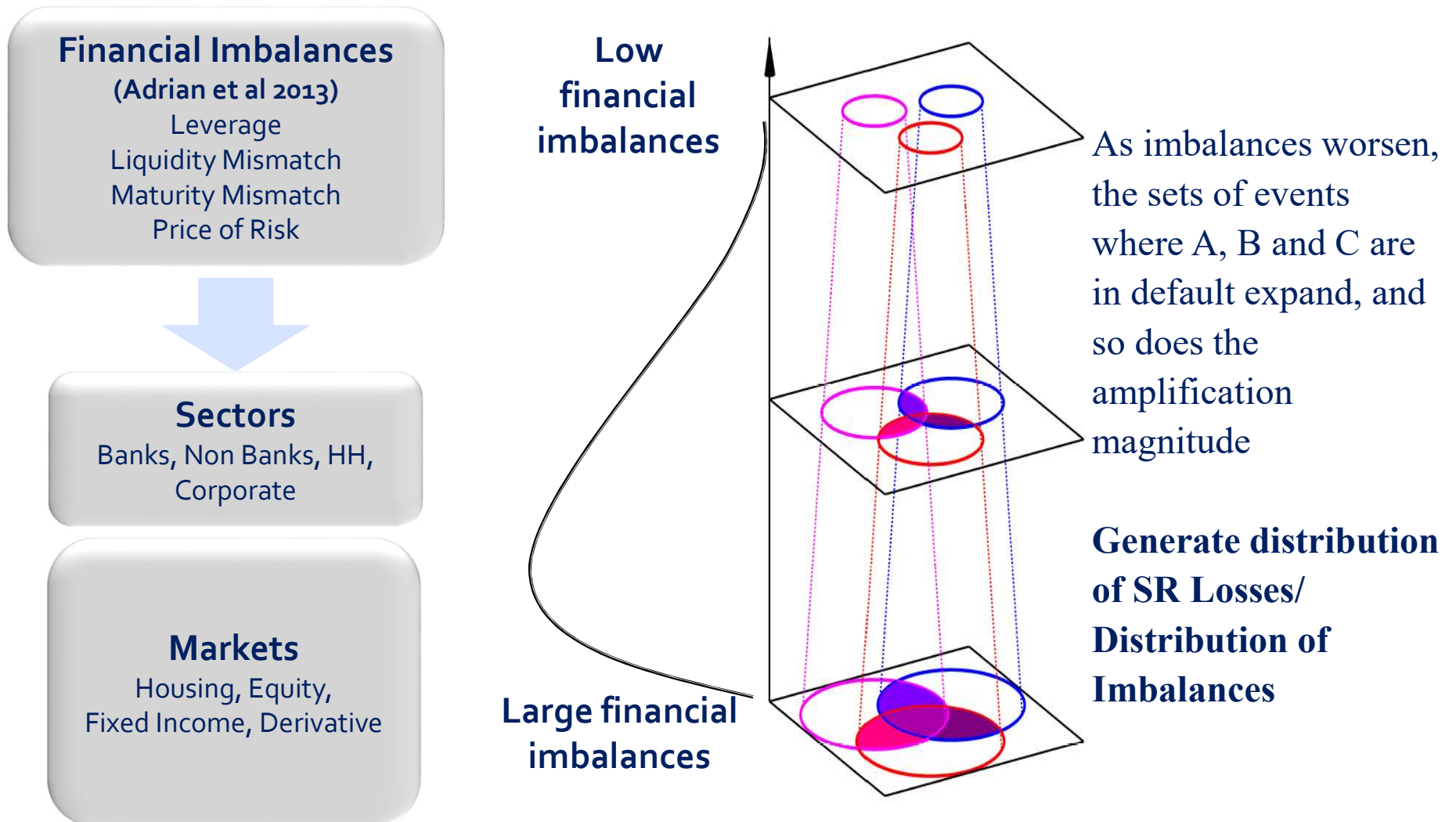


Work in Progress: Magnitude of Amplification

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Work in Progress Use for policy makers

Hiebert, Schueler, Segoviano, Zhao, "Systemic Risk Amplification Magnitude: Conditioning on Financial Imbalances"



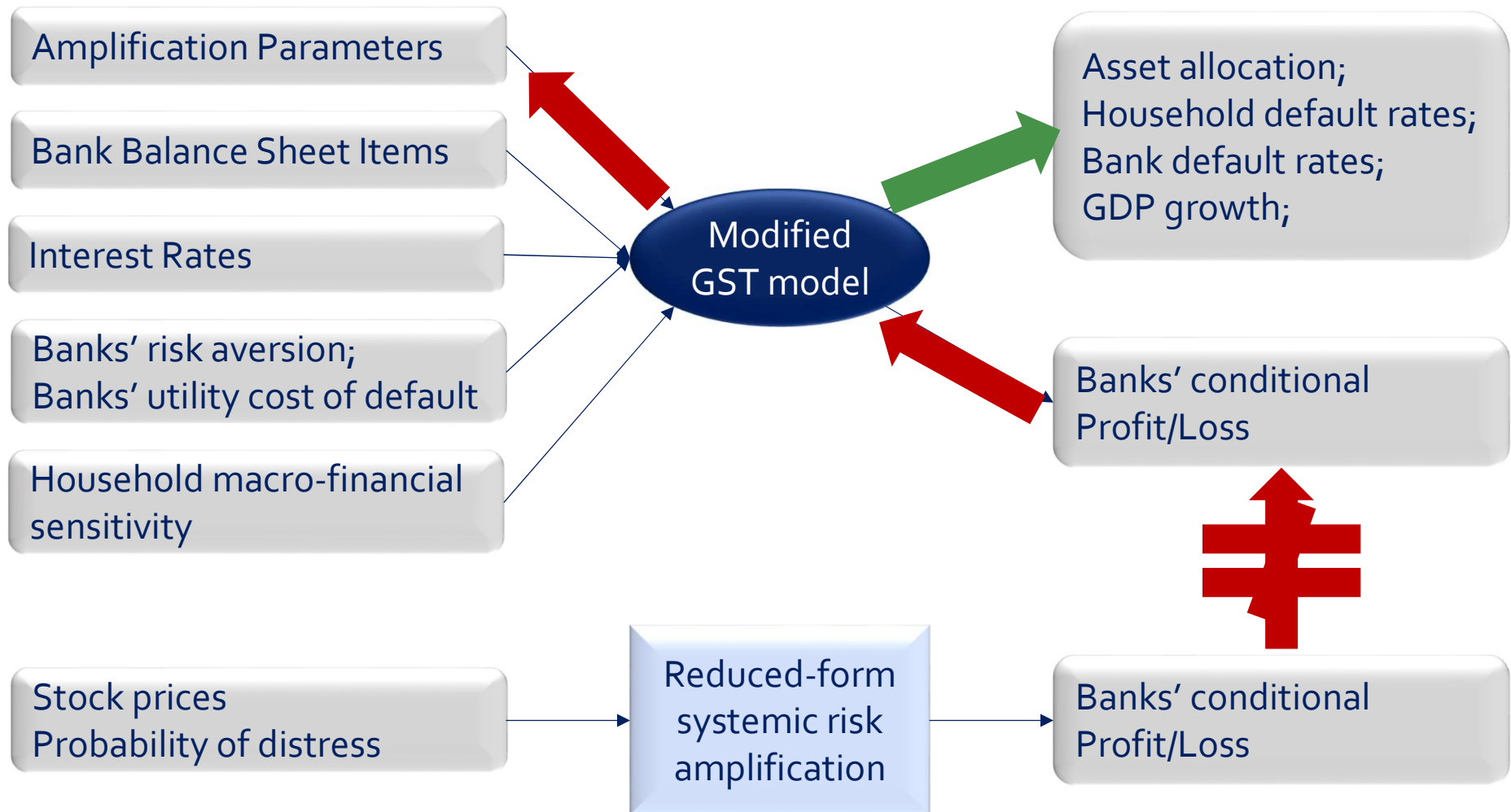


Work in Progress: Bringing together SR Theory and Empirics

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Work in Progress Use for policy makers

Espinoza, Segoviano, Yan, "Systemic Risk: Bringing Together Theory and Measurement"



Goodhart, Sunirand, Tsomocos (2005)



Use for Policy Makers: Calibration of Capital



Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

Buffers	Additional buffers for systemic banks (bank specific)	0-3.5 % ²	Buffers that can be calibrated by stress tests
	Countercyclical capital buffer (all banks)	0-2.5 % ³	
	Capital conservation buffer	2.5%	
	Minimum capital requirement	8%	

1/ The above illustrates the minimum requirements presented in the Basel III framework. National authorities may have additional minimum capital requirements or other types of buffer requirements.

2/ National authorities can impose a capital buffer requirement on SIBs that is higher than 3.5 percent. The Basel framework introduces capital surcharges for G-SIBs ranging from 1 to 3.5 percent. For banks that are systemically important both globally and domestically, the higher of G-SIB and D-SIB capital surcharges applies.

3/ National authorities can impose a CCyB higher than 2.5 percent, while the mandatory international reciprocity applies only up to 2.5 percent.

Source: Anderson, et al, 2017, "Macroprudential Stress Tests and Policies: Searching for Robust and Implementable Frameworks", Systemic Risk Centre, London School of Economics, forthcoming Discussion Paper.



There are many challenges for calibrating a capital buffer strategy.

- **Time consistency.** Aikman, Haldane, and Nelson (2015).
- **Regulatory discretion vs. quantitative calibration.**
- **Robustness of methods.**
- **Consistency of alternative uses of stress tests.**

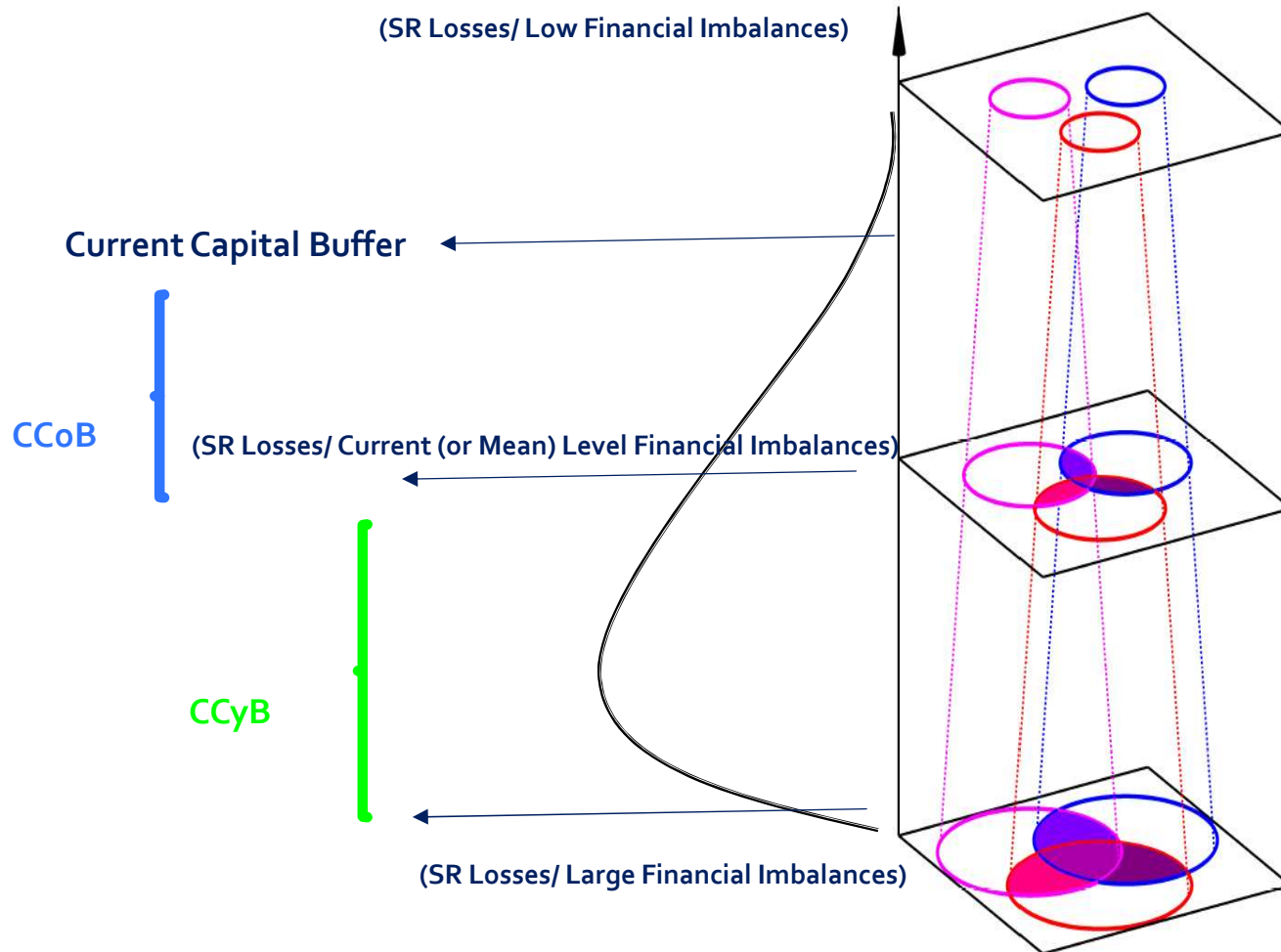


Use for Policy Makers: CCoB and CCyB Surcharge



Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Work in Progress Use for policy makers

SR Amplification magnitude to calibrate capital buffers





Use for Policy Makers: SIB Surcharge

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

- Currently, SIB surcharges unrelated to macroprudential ST
 - SIB surcharges are justified on *perceived externalities*.

- This contradicts risk management perspective of capital:
 - Banks should hold **capital to withstand stress (unexpected) losses**, embedded in the Basel framework.
 - **Difficult to identify causality**: should requirements be on debtors, creditors, transactions?
 - **Capital to withstand vulnerabilities due to SR losses**. All banks subjected to different degrees.

- Important to question
 - Should only SIBs or G-SIBs be subjected to capital charges due to SR vulnerabilities?
 - *Are other instruments better suited to address externalities?*
 - Regulation to **alter the magnitude of financial imbalances**? leverage, liquidity mismatch, etc.
 - Or policies to **alter structural features** of the financial system; e.g., Central clearing, bilateral margining, large exposure limits, etc.



Use for Policy Makers: Other Uses

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

- **Identification of firms that could cause the most severe externalities or be most vulnerable to systemic shocks.**
- **Lending standards.**
- **MaPP responses targeting systems' structural features.**
- **Improving the design of recovery and resolution frameworks.** Goodhart and Segoviano (2015).
- **Understanding of the impact of regulatory constraints.** Divya Kirti and Vijay Narasiman (IMF Working Paper 17/68).



Conclusion

LSE

Challenges to Systemic Risk Modeling Reduced-Form Macroprudential ST An example Use for policy makers

- The proposed framework **makes use of micro stress tests** already implemented
- SR Loss based on **publicly available data**.
- **Cost-efficient.** Computationally simple and relatively light on data requirements.
- **Reduced-form.**
 - We can quantify SR Losses.
 - We can identify “connecting entities”
 - We can estimate likelihood and intensity of contagion effects
 - We **cannot** provide insights into the channels of SR amplification.
- **Conditioning on Financial Imbalances.** Improvement of estimation of **magnitude of amplification** and possibility to estimate a **density of SR losses**.
- **Combining theoretical models with reduced-form measurement.** Identification of **amplification channels** with improved measurement of SR.



References

The logo for the London School of Economics (LSE) is located in the top right corner. It consists of the letters "LSE" in white, set against a red square background.

- Alla, Z., R. Espinoza, Q.H. Li and M. Segoviano, 2017, "Macroprudential Stress Tests: A Reduced-Form Approach to Quantifying Systemic Risk Losses," forthcoming IMF Working Paper 18/49, Washington DC: International Monetary Fund
- Anderson, R., Danielsson, J., Baba, C., Das, U., Kang, H., and Miguel Segoviano, 2017, "Macroprudential Stress Tests and Policies: Searching for Robust and Implementable Frameworks", Systemic Risk Centre, London School of Economics, forthcoming Discussion Paper.
- Bazinas, V., Segoviano, M., 2017, "Assessing Time-varying Macrofinancial Linkages", forthcoming , IMF Working Paper.
- Cáceres, C., Guzzo, V., Segoviano, M., (2010), "Sovereign Spreads: Global Risk Aversion, Contagion or Fundamentals?", IMF Working Paper WP/10/120.
- Cortes, F., Lindner, P., Malik, S., M. Segoviano, "A Comprehensive Multi-Sector Framework for Surveillance of Systemic Risk and Interconnectedness (SyRIN)", forthcoming IMF Working Paper 18/14, Washington DC, International Monetary Fund
- Espinoza, R. and Segoviano, M. (2011). "Probabilities of Default and the Market Price of Risk in a Distressed Economy", IMF Working Paper WP/11/75.
- Espinoza, R., M. Segoviano and J. Yan, (2018) "Systemic Risk: Bringing Together Theory and Measurement", forthcoming Working Paper, Oxford University.
- Goodhart, C., Hofmann B., and Segoviano M., (2006), "Default, Credit Growth, and Asset Prices", IMF Working Paper 06/223.
- Charles AE Goodhart, Pojanart Sunirand, and Dimitrios P Tsomocos. A model to analyse financial fragility. *Economic Theory*, 27(1):107{142, 2006a.
- Hiebert, P., Schueler, Y., Segoviano, M., Zhao, Y., (2018) "Systemic Risk Amplification Magnitude: Conditioning on Financial Imbalances", forthcoming Discussion Paper, Systemic Risk Centre, London School of Economics.
- Segoviano, M. (2006). "Consistent Information Multivariate Density Optimizing Methodology". Financial Markets Group, London School of Economics, Discussion Paper No. 557.
- Segoviano, M., (2006), "The Conditional Probability of Default Methodology," Financial Markets Group, London School of Economics, Discussion Paper 558.
- Segoviano, M. and Padilla, P., (2006), "Portfolio Credit risk and Macroeconomic Shocks: Applications to Stress Testing under Data Restricted Environments," IMF WP/06/283.
- Segoviano, M. and Goodhart, C. (2009). "Banking Stability Measures", IMF WP/09/4.
- Segoviano, M., Espinoza, R., (2017)., "Consistent Measures of Systemic Risk"., Systemic Risk Centre, London School of Economics Discussion Paper 74