



An Encompassing Framework to Estimate Systemic Risk Amplification Losses Based on Publicly Available Information*

**Symposium
Macroprudential Stress Test and Polices:
A Framework**

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**Miguel Angel Segoviano
International Monetary Fund
Monetary and Capital Markets Department**

* This presentations draws material from Alla et al 2017. The views expressed in this presentation are those of the authors solely and do not represent those of the IMF or IMF policy.

"Policy makers constantly face information and time constraints but in times of crisis, when volatility increases, these constraints hurt the most, as history becomes a poor guide (for the future), the significance of market perceptions increases, and reaction times decrease"

C. A. Goodhart, 1975

Outline

1. Systemic Risk and challenges to model it
2. Encompassing Frameworks
3. An EF Developed at the IMF

Challenges SR quantification: Definition

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

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)

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Initial Interpretations of SR

Direct Contagion

Indirect Contagion

Market Amplification Mechanisms

Fire sales in financial markets.

Collateralized agreements. Shleifer and Vishny (2011).

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

Illiquidity spirals. Brunnermeier and Pedersen (2009).

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

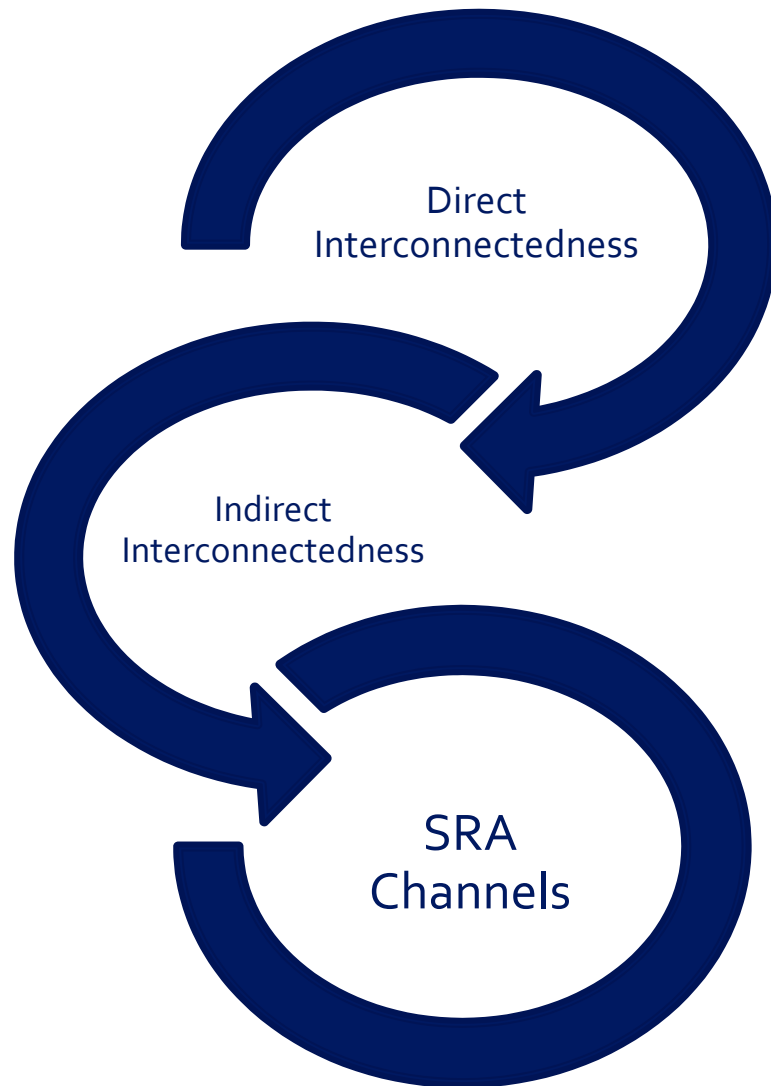
Information Asymmetry Channel

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)

Challenges SR quantification: Channels

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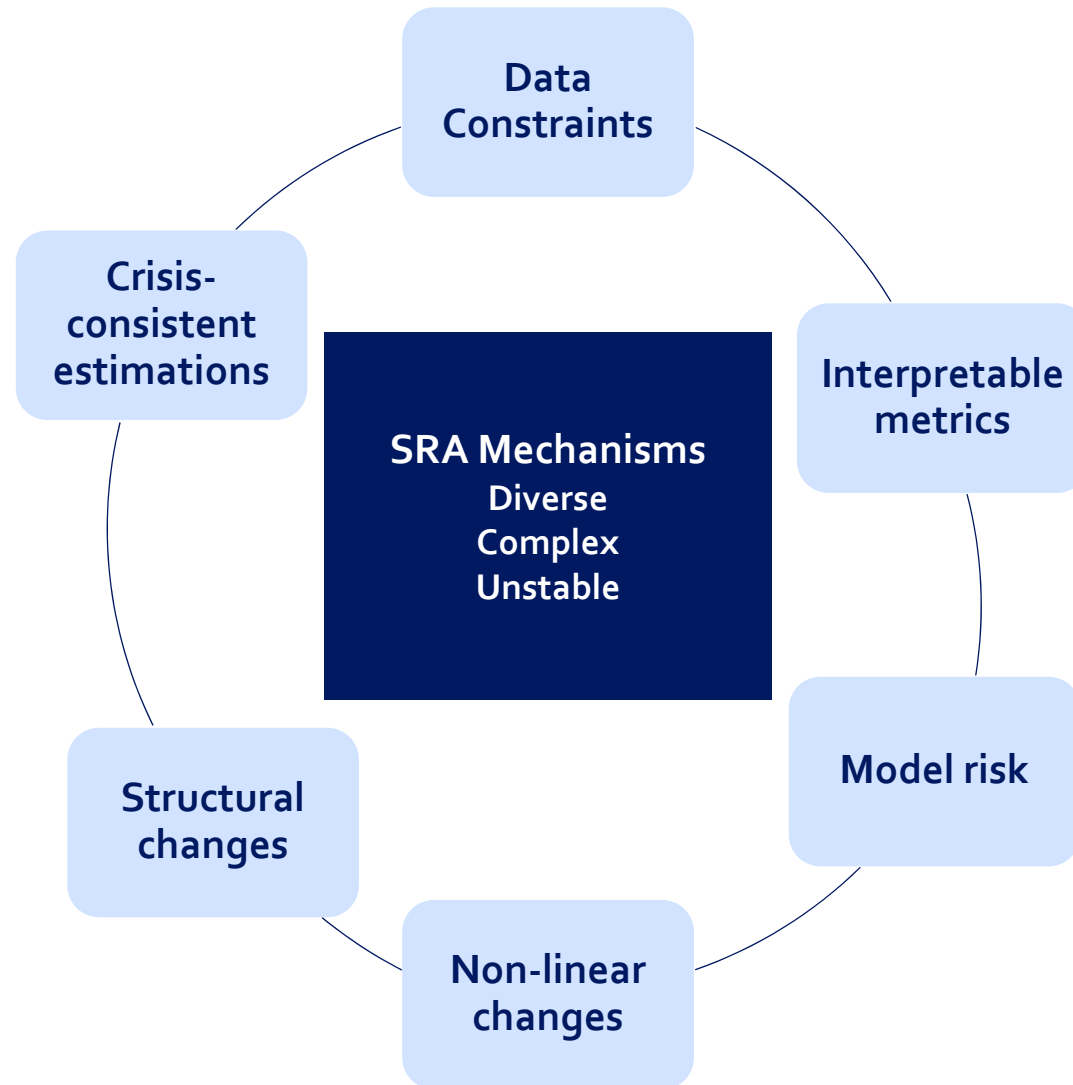


Domino Effects
Diverse Contractual Obligations

Market Price Channels
Asset Fire Sales
Information Asymmetry

Challenges SR quantification: Definition

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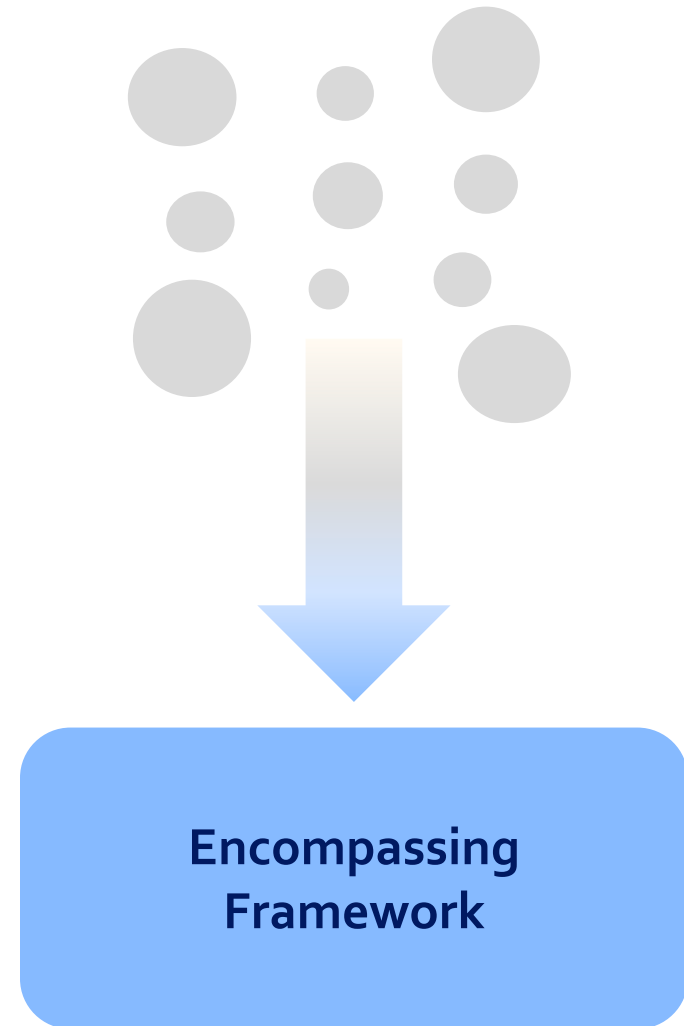
Encompassing Frameworks

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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

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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**

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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

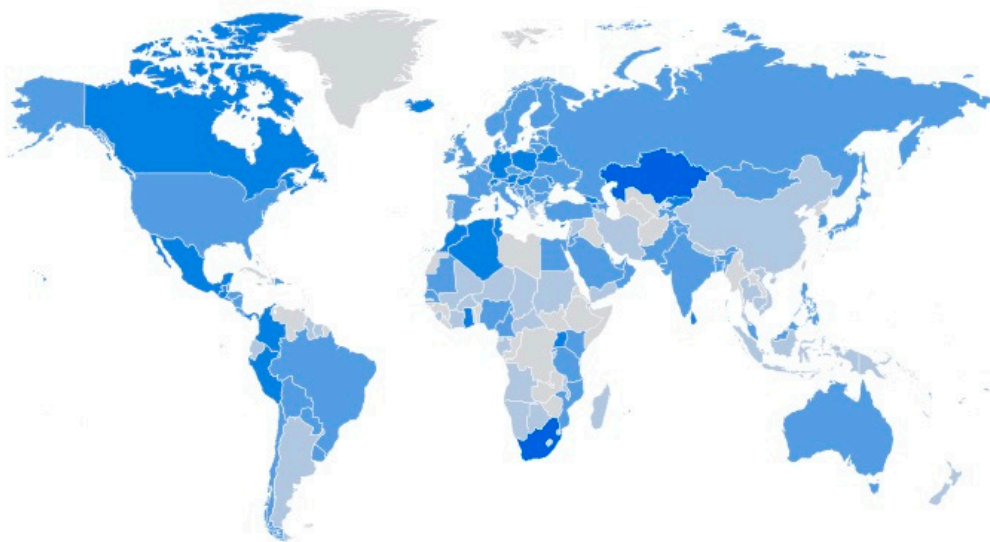
Publicly available data

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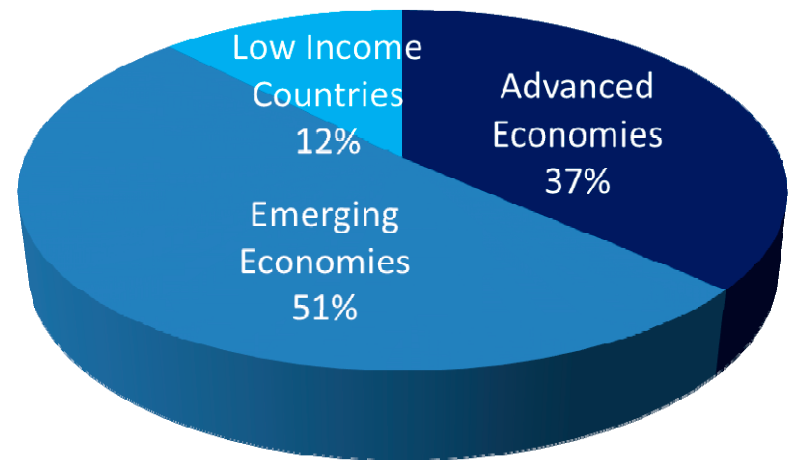
Microprudential ST: Diverse Financial Systems

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321 FSAPs in more than 160 Jurisdictions



0 1 2 3 4+
of FSAPs by Country



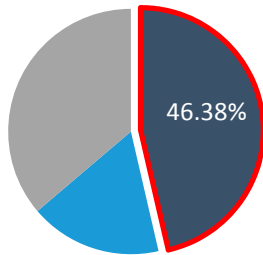
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Microprudential ST: Heterogeneous Methods and Data

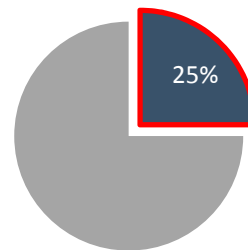
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Top Down

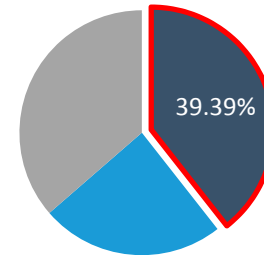
Bank solvency



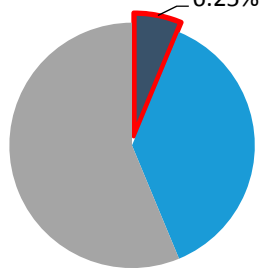
Insurance solvency



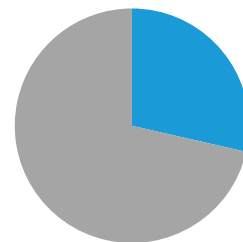
Bank liquidity



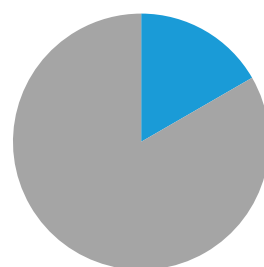
Bank solvency



Insurance solvency



Bank liquidity



Bottom Up

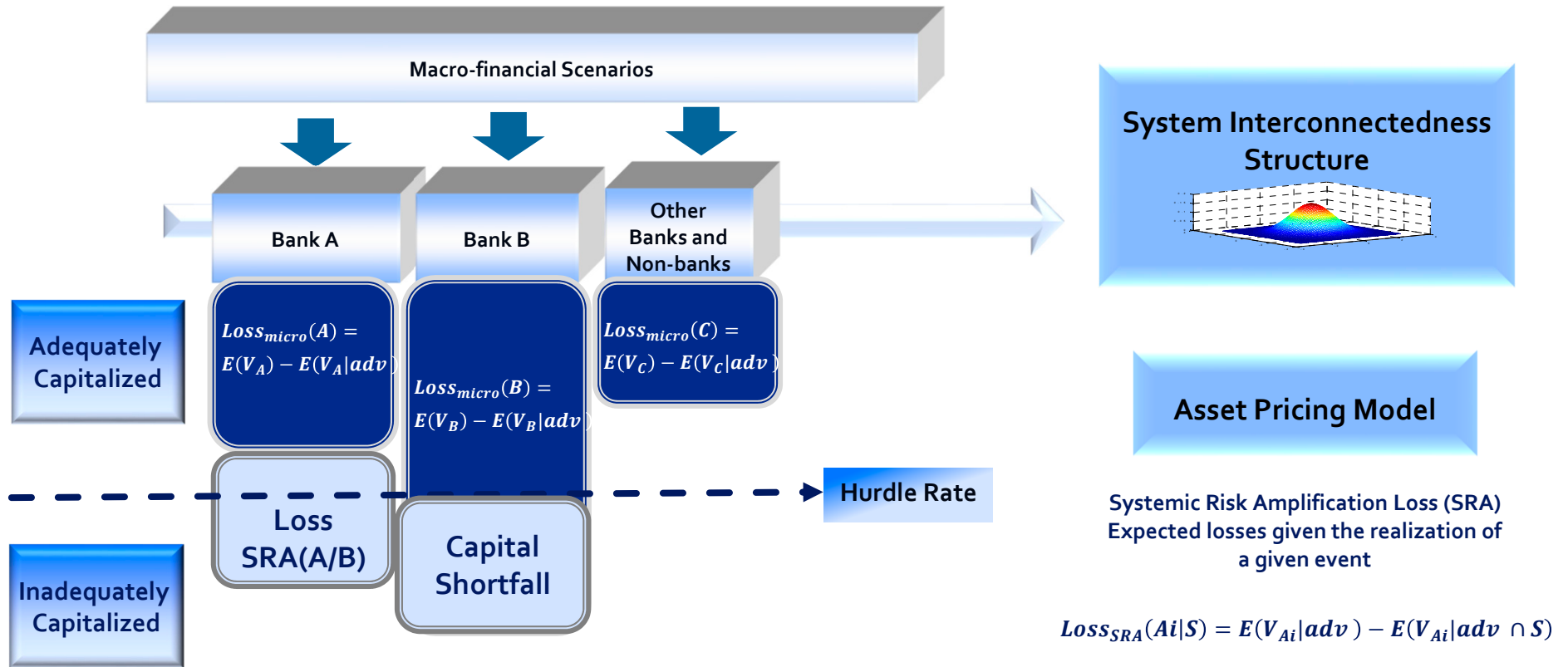
IMF (Workbox)

Jointly

National Authorities

IMF EF Characterization

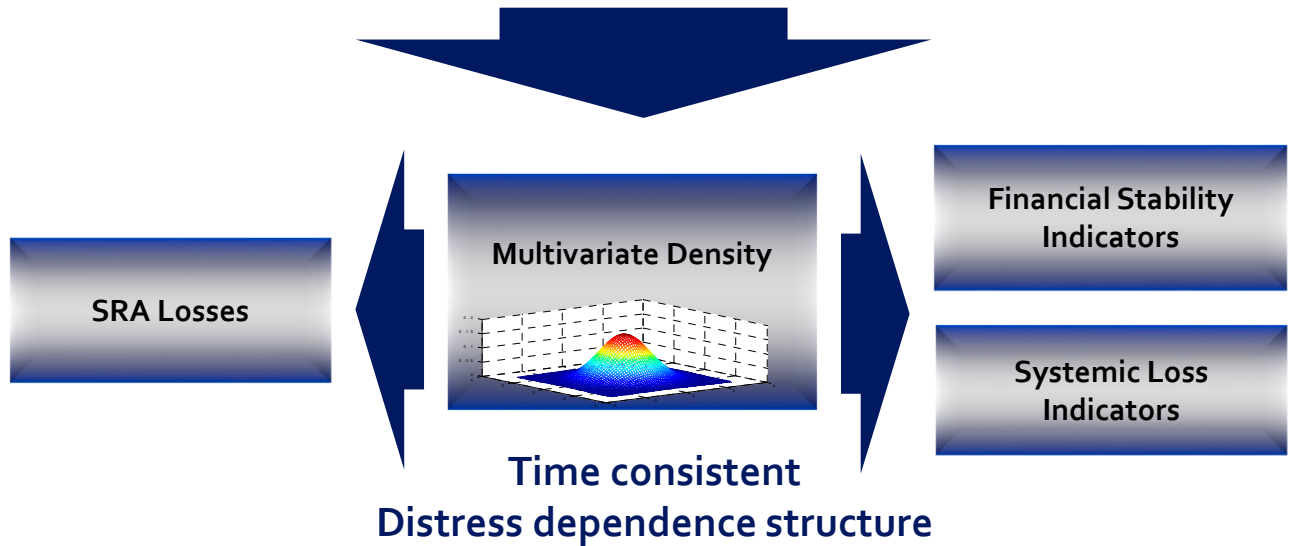
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IMF EF

Multivariate Perspective

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Characterization

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

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

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

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

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

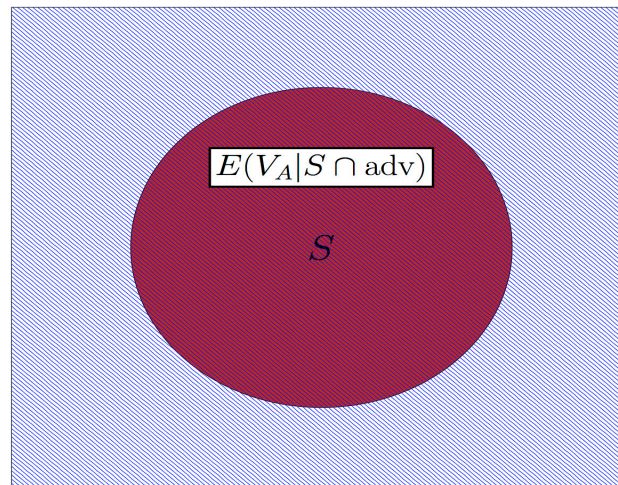
$$\begin{aligned} Loss_{TS}(A|S) &= Loss_{micro}(A) + Loss_{SRA}(A|S) \\ &= E(V_A) - E(V_A|adv \cap S) \end{aligned}$$

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SRA Loss: Venn Diagram

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Identification of the SRA loss in a Venn Diagram



MicroST Loss of a given bank.

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

SRA Loss.

Difference between the value of this bank assuming an adverse macroeconomic scenario, and its value assuming an adverse macroeconomic scenario and the realization of the event S .

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

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SRA Loss: Decomposition

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The **SRA Loss** takes into account **ALL** the potential **CONNECTIONS** across **ALL** entities and their effects on the bank whose loss is assessed.

A high **SRA Loss (A/B)** does **NOT** necessarily mean that there is a strong straight connection between financial entity (FE) A and B.

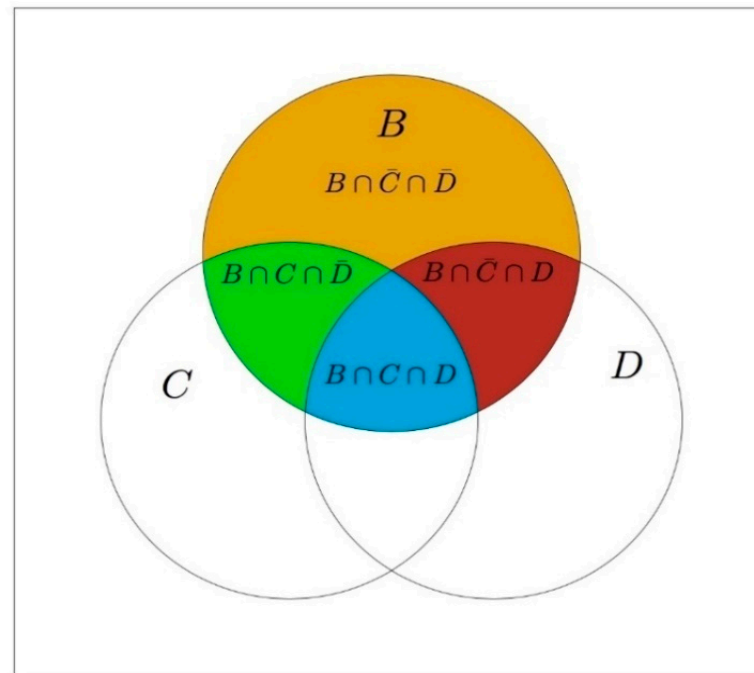
The **contagion path** may for instance include another FE, which is strongly connected to **A and/or B** and explains the high conditional loss of A/B.

Using the law of total expectations, we can **identify the connecting entities** between two given entities.

SRA Loss: Decomposition

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

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



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SRA Loss: Decomposition

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$$\begin{aligned}
 & LOSS_{SRA}(A|B) \\
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 &+ P(B \cap C \cap \bar{D}|B)LOSS_{SRA}(A|B \cap C \cap \bar{D}) + P(B \cap \bar{C} \cap \bar{D}|B)LOSS_{SRA}(A|B \cap \bar{C} \cap \bar{D})
 \end{aligned}$$

$$\text{Intensity Ratio: } R(D_j^i(k_1, \dots, k_l)|A_j) = \frac{LOSS_{SRA}(A_i|D_j^i(k_1, \dots, k_l))}{LOSS_{SRA}(A_i|A_j)}$$

$$C(D_j^i(k_1, \dots, k_l)|A_j) = \frac{P(D_j^i(k_1, \dots, k_l)|A_j) LOSS_{SRA}(A_i|D_j^i(k_1, \dots, k_l))}{LOSS_{SRA}(A_i|A_j)}$$

Defaulting sets D				
Metrics	$LB \cap \overline{WF} \cap \overline{MS}$	$LB \cap \overline{WF} \cap MS$	$LB \cap WF \cap \overline{MS}$	$LB \cap WF \cap MS$
$Prob(D LB)$	0.79	0.11	0.01	0.09
$R(D LB)$	0.45	2.62	2.30	3.69
$C(D LB)$	0.35	0.28	0.03	0.32

Citi (In millions of USD)	
TARP Injection	25,000 ¹
TA (Q2 2008)	2,100,385
TARP/TA	1.19%
Loss SRA (Citi/LB)/TA	1.14%

¹In December 2008, Citi received an additional 20 billion USD through TARP

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Conclusions

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- **The EF makes use of MiPST frameworks that are already implemented;**
- **SRA Loss based on publicly available data.**
 - Market perceptions of financial systems' distress dependence structures, **relevant for crisis contingent estimates.**
 - Given the data limitations faced by the IMF and some authorities (to measure SRA losses), **proper calibration** of methods that rely on ex-ante modeled structures becomes **very challenging**.
- **Cost-efficient.** Become computationally simple and relatively light on data requirements. However, not come at the expense of analytical rigor (Demekas, 2015).
- **Reduced-form.** While can quantify SRA Losses, identify "connecting entities", estimate likelihood and intensity of contagion effects, **does NOT provide insights** into the economic and financial causes behind SRA mechanisms.
- **Parallel running.** Even if alternative frameworks might be feasible given data availability, IMF-EF can be helpful for improving calibrations of alternative frameworks; e.g., parallel DSGE (structural) and VAR (reduced-form).

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