All Bank Risks are Idiosyncratic, Until They Are Not: The Case of Operational Risk



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Discussion:



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Idea: OpRisk contributes to systemic risk

Traditionally, OpRisk has been believed to be largely idiosyncratic to a bank.



Background: Regulatory capital requirements

Old approach (Basel II): Advanced Measurement Approach (AMA).

- Regulatory capital is based on internal loss models.
- o Risk-sensitive.
- But... Banks use different internal modeling practices. Lack of comparability of riskweighted assets calculations across banks.

New approach (Basel III): Standardized Measurement Approach (SMA).

- Regulatory capital is an increasing function of BI Component and Loss Component.
- o Different weights to different bank activities.
- Risk-sensitive (but less than AMA?). Simple. Greater comparability across banks.

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The paper suggests that bank activities with greater contribution of OpRisk to systemic risk should be assigned higher weights.

Main findings

Biggest contributors to systemic risk are:

By OpRisk event type

- Internal fraud (IF, 1% *)
- Clients, Products, and Business Practices (CPBM, 78%)
- Execution, Delivery, and Process Management (EDPM, 14%)

By business line

- Retail banking (47%)
- Corporate other (18%)

By operational loss amount/frequency

- Total amount of tail losses (>99th, 99.5th, 99.9th perc.)
- o Frequency of tail events
 - * % all operational losses.

- I enjoyed reading the paper!
- Well written.
- Comprehensive data (2002-2016), close to 300K op. loss events.
- Great idea, big economic implications.
- Of interest to regulators in light of Basel III.

Motivation (1)

What is the **channel** through which **OpRisk** contributes to systemic risk?



• From the paper: "if idiosyncratic losses from any risk source for a large financial institution are big enough, they can pose a threat within an interconnected financial system"

How? Give theory / examples / empirical evidence.

 Examples in the introduction (SocGen, Barings, WF) are all examples that impact just one institution. Need better examples (e.g., 2008 fin. crisis?)

Motivation (2)

Why would an operational loss (\$\$) lead to systemic risk?

- Operational loss (\$\$) = direct cost of operational risk
- But there are also indirect costs: Reputational damage, e.g., loss of market value), loss of customers, loss of personnel (e.g., CEO resignation)
- Most banks can easily handle a multi-mln \$ loss, even a bln \$ loss. Most banks fail *not* because of direct material loss, but because of reputational damage (e.g., WF).
- <u>Suggestion</u>: Think about measures of indirect operational loss, e.g., loss in market value or CDS spread increase around/after operational risk announcements.

Empirical methodology (1)

Premise: This quarter's op. losses help predict this quarter's (?) systemic risk.



In most instances, there seems to be a several quarters' lag.

Suggestion: Include lagged op. losses in regression models. 4 lags.

Empirical methodology (2)

One key model relies on Acharya et al.'s (2017) estimate of systemic risk:

 $SES_{i,t} = 0.15MES_{i,t-1} + 0.04LVG_{i,t-1}$

Acharya et al. (2017)

Use realized stock returns during crisis (2007-2009) A bank's stock returns on market's worst 5% days before crisis (2006-2007) A bank's leverage before crisis (2006-2007)

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(Do not estimate. Use coefficient estimates 0.15 and 0.04 to compute systemic risk... but over what period?) A bank's stock returns on market's worst 5% days before crisis (2006-2007)

A bank's stock returns on market's worst 5% days last quarter A bank's leverage before crisis (2006-2007)

A bank's leverage this quarter

(2017)

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- Acharya et al. use 1 obs per bank: one before and one after crisis. This paper uses Ο quarterly data 2002Q1 - 2016Q4. \rightarrow Same coefficients won't hold.
- This paper's sample contains the financial crisis... Same coefficients won't hold. Ο
- Acharya's sample: 102 banks >\$5bln market cap. This paper: 26 BHCs >\$50bln assets. Ο
- Acharya's model included additional controls, but they are statistically insignificant, so ok, Ο but should still include for consistency.

Suggestion: Re-estimate the model.

Empirical methodology (3)

Development of hypotheses. Section 3 lacks theoretical arguments and one-sided hypotheses.

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For example,

Hypothesis 2A: Operational risks associated with different event types and business lines have different impacts on the systemic risk of large financial institutions, ceteris paribus.

- <u>Suggestion</u>: Need theoretical/empirical arguments why op. losses in certain ETs and BLs would contribute more to systemic risk.
- The paper finds that *Internal Fraud* events are significantly associated with systemic risk. Unintuitive... Why?
- The paper finds that op. losses in *Retail Banking* business line are significantly associated with systemic risk. Why?

Empirical methodology (4)

Timing.

- 1. Occurrence date.
- 2. Discovery date.
- 3. Accounting date. = The first date financial impact is recorded or legal reserves for future losses are estimated.

There are also:

- 4. Public announcement date.
- 5. Settlement date.

Empirical methodology (4)

Timing.

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There are also:

- 4. Public announcement date.
- 5. Settlement date.
- Concerns:
 - Authors aggregate all losses from the same event at accounting date.
 <u>Suggestion</u>: Treat them as separate losses to avoid look-ahead bias?
 - Is it possible to accurately estimate future financial impact around or prior to financial crisis? How close is your estimation vs. true realized loss?

Other

- Acharya et al. (2017) also examined systemic risk in the CDS market.
- Implicit government guarantees. What's the effect on systemic risk?
- Are your estimates economically significant?

"A 10% increase in operational losses is associated with a 0.0067 increase in Systemic Risk or 0.048 standard deviations."

"A one standard deviation increase in Ln(OpLoss) is associated with 0.14 standard deviations increase in Systemic Risk."

- To define 'tail events,' you use 99th, 99.5th, and 99.9th percentiles. Why? Can use Extreme Value Theory.
- Risk Management Index is used as a control variable. Endogeneity?
- Distance-to-default and its interaction with op. loss may be problematic. Endogeneity. Perhaps, it's not DD that is amplifying the effect of op. losses on systemic risk, but a 3rd variable that's driving both DD and op. losses.
- Choice of IV variables in robustness tests is unclear. It seems that all 3 are directly related to systemic risk.
- Would greater regulatory capital for op. risk reduce systemic risk?

Thank you for your attention!

Comments or questions:



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EXTRA

Operational risk event types

ET1: Internal Fraud

- unauthorized activity, theft & fraud involving at least 1 internal party

ET2: External Fraud

- theft & fraud by a 3rd party, systems security

ET3: Employment Practices and Workplace Safety

- discrimination, general liability, compensation
- ET4: Clients, Products, and Business Practices

- improper business & market practices, model errors

ET5: Damage to Physical Assets

- natural and man-made disasters, vandalism

ET6: Business Disruption and Systems Failures

- hardware & software failures, telecommunications

ET7: Execution, Delivery, and Process Management

- data entry error, missed deadline, delivery failure