

Pitfalls in the use of systemic risk measures*

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The views herein do not necessarily reflect those of the Deutsche Bundesbank.

Definitions

—We favor an inclusive definition:

Systemic risk is the risk of a breakdown or severe dysfunction of the financial system - **no matter what the source of it is.**

—The **systemic risk measures (SRM)** we talk about try to

- Quantify how much an entity **contributes** to the vulnerability of the financial system
- Are based on **equity (or asset) returns.**

Return-based SRM can be relevant even if they are not used by supervisors

- Internal risk management
 - Systemic risk of counterparties
 - Macro risk analysis and stress testing
- Cost of capital: Market may use information from market prices even though regulators do not

Possible pitfalls in using systemic risk measures

- Estimation error
- Conceptual misunderstandings, e.g. systemic vs systematic
- Misaligned incentives
 - A bank wishing to lower its systemic risk may take an action that actually increases system risk
- Wrong diagnosis
 - Counterparty **A** appears to be more dangerous for system stability than **B** but the reverse is true

Systemic risk measures 1

- R_i ...return of bank i
 R_S ... market return, or „system“ return

ΔCoVaR (Adrian, Brunnermeier, Oct 2011):

- *Change of the system's VaR through bank i moving from a normal to a very bad state; formally: $Q_\alpha(\dots)$... α -quantile*
- $\Delta\text{CoVaR}_\alpha^{S|i} \equiv Q_\alpha(-R_S | R_i = Q_\alpha(R_i)) - Q_\alpha(-R_S | R_i = Q_{0.5}(R_i))$

$\text{Exposure } \Delta\text{CoVaR}$:

- *Change of bank i 's VaR through the system moving from a normal to a very bad state; formally:*
- $\Delta\text{CoVaR}_\alpha^{i|S} \equiv Q_\alpha(-R_i | R_S = Q_\alpha(R_S)) - Q_\alpha(-R_i | R_S = Q_{0.5}(R_S))$

Systemic risk measures 2

Marginal expected shortfall (MES)

- (Acharya, Pedersen, Philippon, Richardson, 2010)

$$MES_{\alpha}^i \equiv \mathbf{E} \left[-R_i \mid R_S < Q_{\alpha}(R_S) \right]$$

Beta

- Regression: $R_{i,t} = \alpha_i + \beta_i R_{S,t} + u_{i,t}$

Do SRM set the right incentives?

Sensitivities in a linear normal model

- Classic **market model**: N banks, returns:

$$R_i = \beta_i F + \varepsilon_i; \quad (F \sim N(\mu, \sigma_F^2), \varepsilon_i \sim N(0, \sigma_i^2), \text{ independent})$$

- Bank sector index $R_S = \sum_{j=1}^N w_j R_j$ represents „**the system**“

- Very simple representation of the SRM:

$$\Delta \text{CoVaR}_\alpha^{S|i} = \frac{\text{cov}(R_S, R_i)}{\sigma(R_i)} \Phi^{-1}(1 - \alpha)$$

$$\Delta \text{CoVaR}_\alpha^{i|S} = \frac{\text{cov}(R_S, R_i)}{\sigma(R_S)} \Phi^{-1}(1 - \alpha)$$

$$MES = -\beta_i \mu + \frac{\text{cov}(R_S, R_i)}{\sigma(R_S)} \frac{\phi(\Phi^{-1}(\alpha))}{\alpha}$$

$$\text{beta}_i = \frac{\text{cov}(R_S, R_i)}{\sigma^2(R_S)}$$

Do SRMs set the right incentives?

Sensitivities to risk parameters in a linear normal model

- Assumptions

- Banks can steer their idiosyncratic risk (σ_i), systematic risk (β_i) and relative size (w_i).
- Banks strive for low SRMs (e.g., in presence of SRM- based risk charges)

- Direct effect: on the own SRM:

$$\frac{\partial}{\partial p_i} [SRM_i], \quad p_i \in \{\sigma_i, \beta_i, w_i\}$$

- Relative effect: compared to another bank's SRM:

$$\frac{\partial}{\partial p_i} \left[\frac{SRM_i}{SRM_j} \right], \quad p_i \in \{\sigma_i, \beta_i, w_i\}$$

Do SRM set the right incentives?

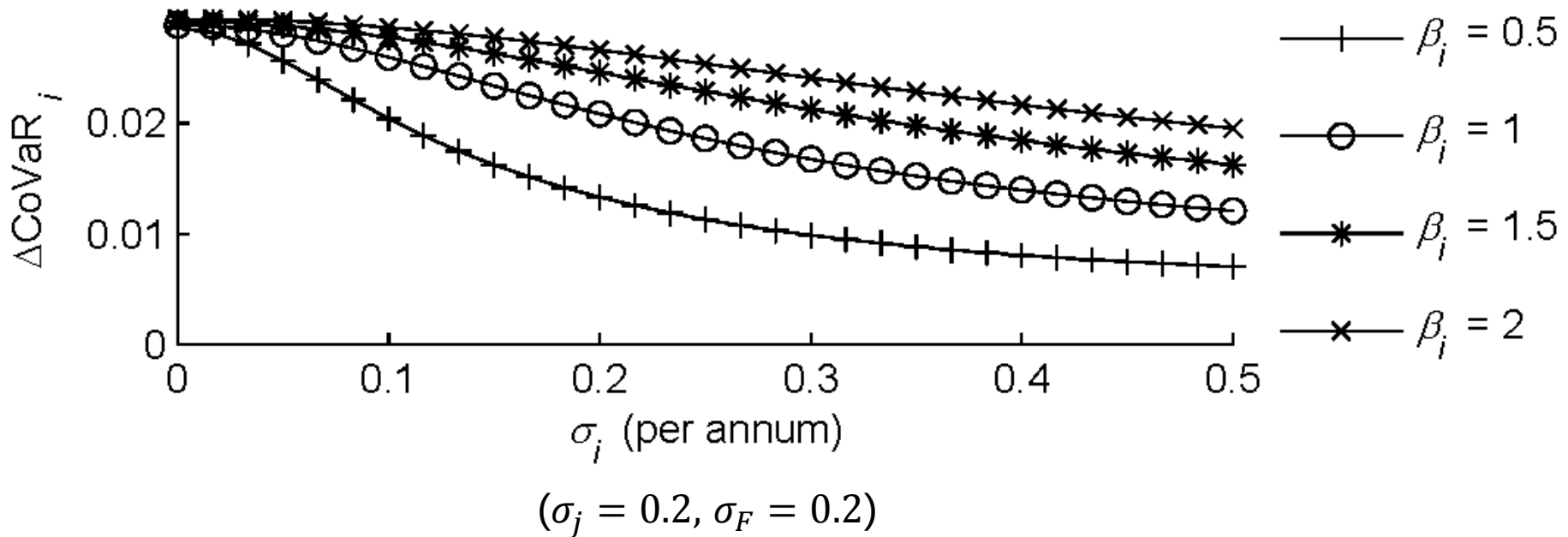
Linear model; sensitivity to risk parameters

Parameter	Effect type	ΔCoVaR (Bank i stressed)	Exposure ΔCoVaR (System stressed)	MES	Beta
idiosyncratic risk σ_i	direct	+/-	+	+	+
	relative	+/-	+	+ ⁿ	+
systematic risk β_i	direct	+	+	+ ⁿ	+
	relative	+/-	+/-	+/-	+/-
size w_i	direct	+/-	+	+	+/-
	relative	+ ⁿ	+ ⁿ	+ ⁿ	+ ⁿ

Legend: + SRM rises with risk parameter
 +/- SRM rises / falls, depending on other parameters
 +ⁿ SRM rises under non-exotic conditions
 Blue fields SRM and overall systemic risk can have opposite sensitivities

Do SRM set the right incentives?

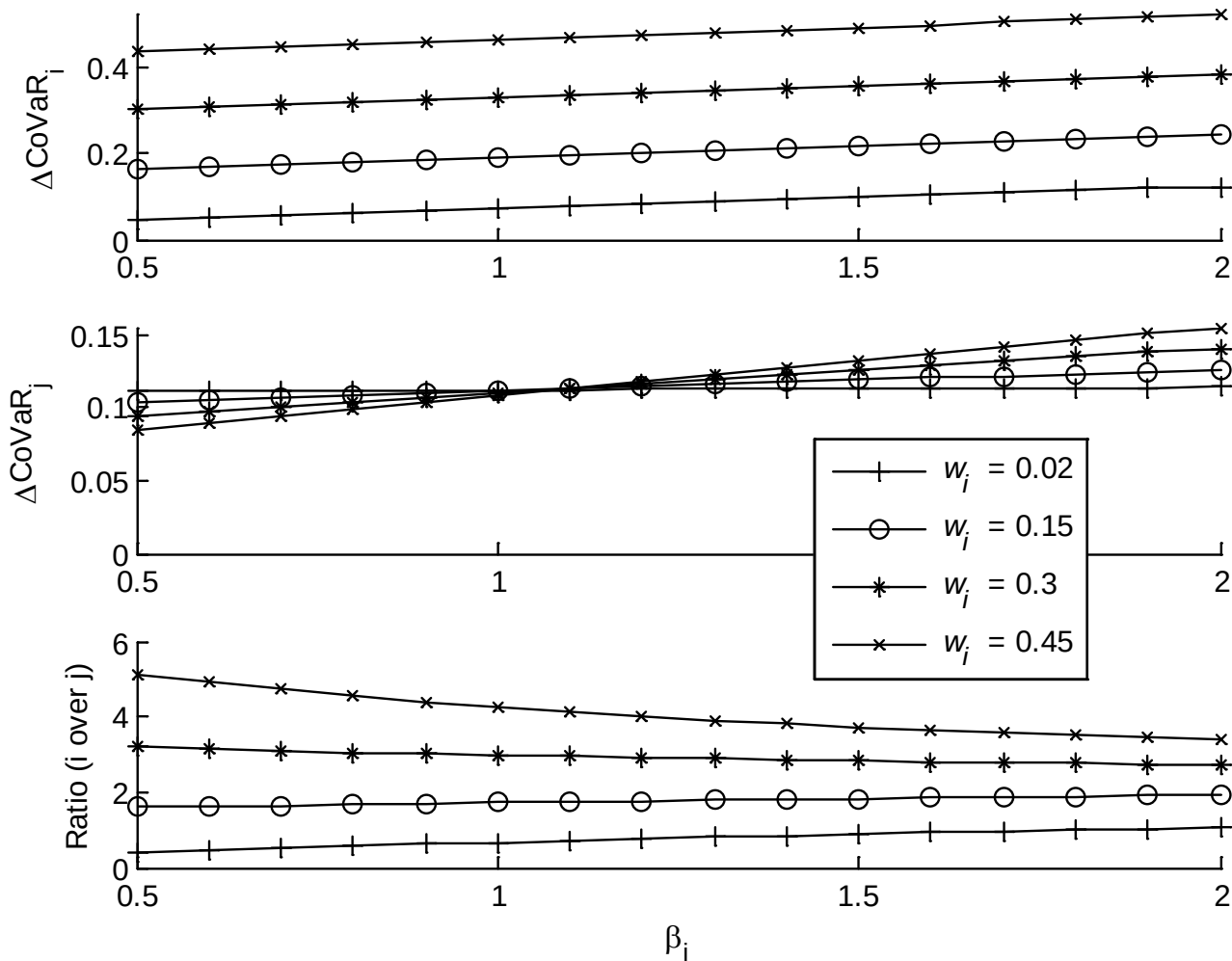
Example: sensitivity of ΔCoVaR to idiosyncratic risk σ_i



$$\Delta\text{CoVaR}_{\alpha}^{S|i} = \frac{\text{cov}(R_S, R_i)}{\sigma(R_i)} \Phi^{-1}(1 - \alpha)$$

Do SRM set the right incentives?

Example: sensitivity of ΔCoVaR to systematic risk β_i



Direct effect

Similar graphs
for the other
SRMs

(Side effect)

Relative effect

Do SRM set the right incentives?

Robustness to distributional assumptions

- Multivariate t -distributed returns, increasing tail thickness
- Dynamic structural model; multivariate extension of Collin-Dufresne / Goldstein (2001)
 - Lognormal asset returns, 1 systematic factor
 - Stationary equity returns with thicker-than-normal tails

Do SRM set the right incentives?

dynamic structural model; sensitivity to risk parameters

Parameter	Effect type	Return type	ΔCoVaR	Exposure ΔCoVaR	MES	Beta
Idiosyncratic risk σ_i	direct	assets	—	+	+	+
		equity	—	+	+/-	+
	relative	assets	—	+	+	+
		equity	—	+	+/-	+
Systematic risk ρ_t	direct	assets	+	+	+	+
		equity	+	+	+	+
	relative	assets	+/-	+/-	+/-	+/-
		equity	+/-	+/-	+/-	+/-
Size w_i	direct	assets	+	+	+	+/-
		equity	+	+/-	+	+/-
	relative	assets	+	+	+	+
		equity	+/-	+/-	+	+

Σ : New negative effects, only for equity → tail thickness matters.

Previous negative effects confirmed.

negative sensitivities that did not appear in the linear normal model

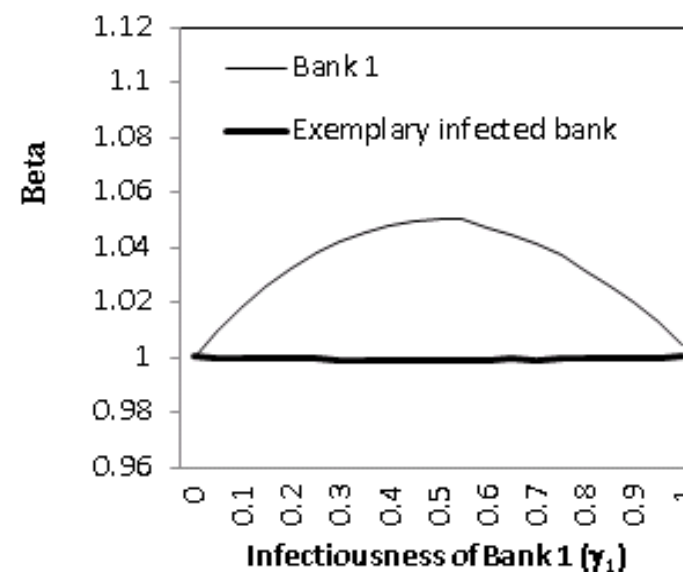
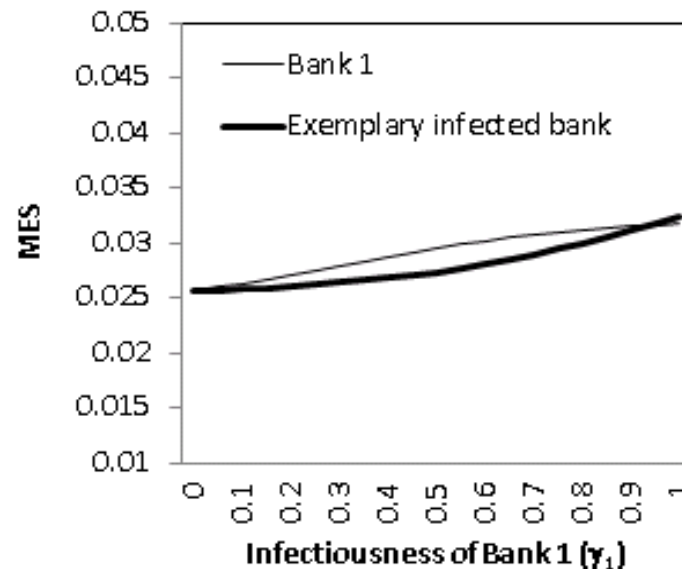
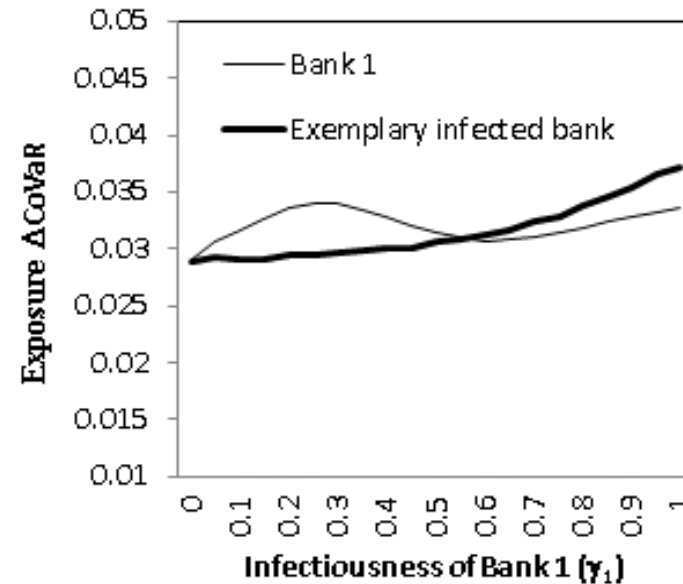
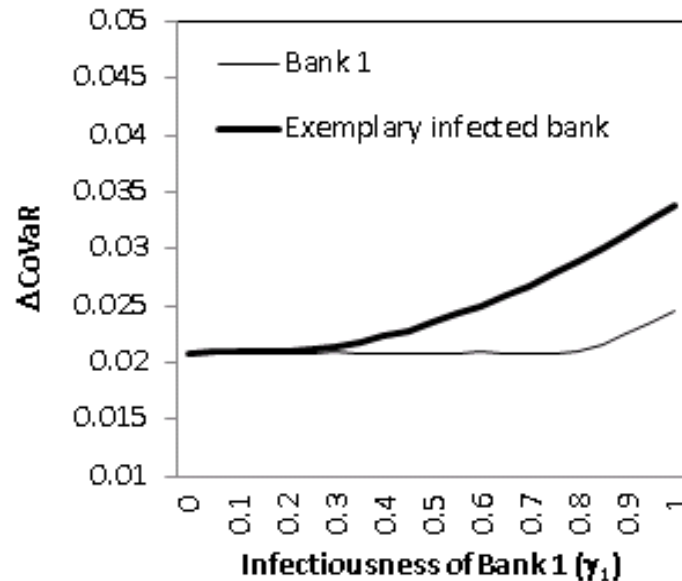
Can SRM identify an “infectious” bank?

Model setup

- A single infectious bank: $R_1 = \beta_1 F + \varepsilon_1$
- Infected banks: $R_j = \beta_j F + \varepsilon_j + \gamma_1 I_{\{\varepsilon_1 < \kappa\}} \varepsilon_1, \quad j = 2, \dots, N$
- Bank sector index $R_S = \frac{1}{N} \sum_j R_j$
- Same beta and idiosyncratic risk for all banks
- Monte Carlo simulation
 - varying impact parameter γ_1 and „infection threshold“ κ
 - $N = 50$

Can SRM identify an “infectious” bank?

Varying the impact parameter γ_1



Can SRM identify an “infectious” bank?

Robustness

Tests

- Calibrating volatility and expectation of infected banks to that of the infectious bank
- Varying the contagion threshold κ (quantiles at 1% , 0.1%)
- Making the infectious bank big (25% weight in the index return)
- Raising the loading to systematic risk β_1 to 1.25
- Five infectious banks
- Systematic factor as GARCH(1,1), same unconditional volatility as before
- All factors t -distributed, 4 degrees of freedom
- Volatility spillover: $R_j = \beta_j F + \varepsilon_j \times \left(I_{\{\varepsilon_1 \geq \kappa\}} + m I_{\{\varepsilon_1 < \kappa\}} \right)$
- Time delay in the spillover:

$$R_{jt} = \beta_j F_t + \varepsilon_{1t} + 0.5\gamma_1 I_{\{\varepsilon_{1t} < \kappa\}} \varepsilon_{1t} + 0.5\gamma_1 I_{\{\varepsilon_{1t-1} < \kappa\}} \varepsilon_{1t-1}$$

Result

- Base case **confirmed** in all cases, except for delayed spillover

Conclusion

- If banks benefit from low SRMs, some SRMs set **strange incentives** w.r.t. idiosyncratic risk, systematic risk and size, even in a plain linear **well-behaved model** with normal returns.
- **Contagion** model: **no clear picture** whether, when and by which SRM an infectious banks would be identified.
- Results are robust to various changes in the model.
- ➔ **A direct application of the proposed measures to regulatory capital surcharges for systemic risk could create a lot of noise and wrong incentives to banks.**