### A General Framework for Systemic Risk

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Joint work with Chen Chen and Garud Iyengar.

# Systemic Risk

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### Examples:

- firms in an economy
- business units in a company
- suppliers, sub-contractors, etc. in a supply chain network
- generating stations, transmission facilities, etc. in a power network
- flood walls, pumping stations, etc. in a levee system

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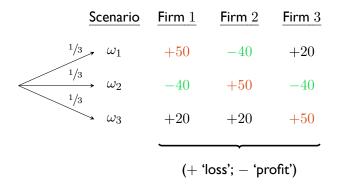
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Systemic risk refers to the risk of catastrophic collapse of the entire system. Involves:

- the simultaneous analysis of outcomes across all entities in a system
- the possibility of complex interactions between components

### **Joint Distribution of Outcomes**

- 3 firms in 3 future scenarios (equally likely)
- Loss matrix:



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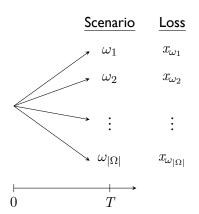
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- Transmission of illiquidity, 'bank runs' (e.g., Lehman)
- Fire sales, asset price contagion (e.g., CDOs)

## Contributions

- A general, axiomatic framework for coherent systemic risk analyzes joint distribution of outcomes allows for *some* endogenous mechanisms of contagion subsumes many recently proposed systemic risk measures
- A structural decomposition of systemic risk
- A dual representation for systemic risk measures 'shadow price of risk'
- A mechanism for systemic risk attribution & decentralization
- Methodology extends to a much broader class of risk functions

## **Literature Review**

- Axiomatic theory of single-firm risk measures: Artzner et al., (2000); see survey of Schied (2006)
- Systemic risk measures: portfolio approach Gauthier et al., (2010); Tarashev et al., (2010); Acharya et al., (2010); Brownlees & Engle (2010); Adrian & Brunnermeier (2009)
- Systemic risk measures: deposit insurance / credit approach Lehar (2005); Huang et al., (2009); Giesecke & Kim (2011)
- Structural models of contagion & systemic risk: Acharya et al., (2010); Staum (2011); Liu & Staum (2010); Cont et al., (2011); Bimpikis & Tahbaz-Salehi (2012)
- Portfolio attribution: Denault (2001); Buch & Dorfleitner (2008)



 $\Omega = \mathsf{set} \mathsf{ of} \mathsf{ scenarios}$ 

$$x \in \mathbb{R}^{\Omega}$$

 $x_{\!\omega} = \mathrm{loss}$  in scenario  $\omega$ 

**Definition.** A coherent single-firm risk measure is a function  $\rho \colon \mathbb{R}^{\Omega} \to \mathbb{R}$  that satisfies, for all  $x, y \in \mathbb{R}^{\Omega}$ :

- (i) Monotonicity: if  $x \ge y$ , then  $\rho(x) \ge \rho(y)$
- (ii) Positive homogeneity: for all  $\alpha \ge 0$ ,  $\rho(\alpha x) = \alpha \rho(x)$

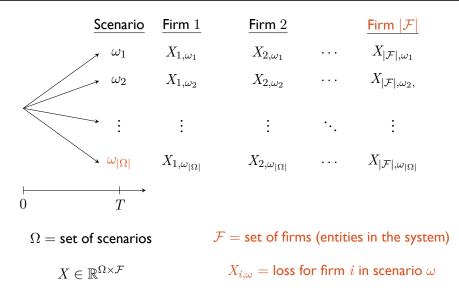
(iii) Convexity: for all 
$$0 \le \alpha \le 1$$
,  
 $\rho(\alpha x + (1 - \alpha y)) \le \alpha \rho(x) + (1 - \alpha)\rho(y)$ 

(iv) Cash-invariance: for all  $\alpha \in \mathbb{R}$ ,

$$\rho(x + \alpha \mathbf{1}_{\Omega}) = \rho(x) + \alpha$$

[Artzner et al., 2000]

## Systemic Risk Measures



- $\Omega = \text{set of scenarios}, \ \mathcal{F} = \text{set of entities in the system}, \ X \in \mathbb{R}^{\Omega \times \mathcal{F}}$
- $X_{i,\omega} =$ loss for firm i in scenario  $\omega$ ,  $X_{\omega} =$ loss vector in scenario  $\omega$

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**Definition.** A systemic risk measure is a function  $\rho \colon \mathbb{R}^{\Omega \times \mathcal{F}} \to \mathbb{R}$  that satisfies, for all economies  $X, Y, Z \in \mathbb{R}^{\Omega \times \mathcal{F}}$ :

(i) Monotonicity: if  $X \ge Y$ , then

$$\rho(X) \ge \rho(Y)$$

(ii) Positive homogeneity: for all  $\alpha \geq 0$ ,

$$\rho(\alpha X) = \alpha \rho(X)$$

(iii) Normalization:  $\rho(\mathbf{1}_{\mathcal{E}}) = |\mathcal{F}|$ 

**Definition.** (con't.) Given  $x, y \in \mathbb{R}^{\mathcal{F}}$ , define the ordering  $x \succeq_{\rho} y$  by  $x \succeq_{\rho} y \iff \rho(x, \dots, x) \ge \rho(y, \dots, y)$ 

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(iv) Preference consistency: if  $X_{\omega} \succeq_{\rho} Y_{\omega}$  for all scenarios  $\omega$ , then  $\rho(X) \ge \rho(Y)$ 

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### Definition. (con't.)

- (v) Convexity: for all  $0 \le \alpha \le 1$ ,  $\bar{\alpha} = 1 \alpha$ 
  - (a) Outcome convexity: if

 $Z = \alpha X + \bar{\alpha} \, Y$ 

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(b) Risk convexity: if for all scenarios  $\omega \in \Omega$ ,

 $\rho(Z_{\omega}, \dots, Z_{\omega}) = \alpha \rho(X_{\omega}, \dots, X_{\omega}) + \bar{\alpha} \rho(Y_{\omega}, \dots, Y_{\omega})$ then,  $\rho(Z) \le \alpha \rho(X) + \bar{\alpha} \rho(Y)$ 

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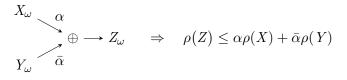
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Two different notions of **diversity** 

#### Definition. (con't.)

1. Outcome convexity: Increasing diversification reduces risk



2. Risk convexity: Removing randomness reduces risk

$$\rho(Z_{\omega}\mathbf{1}_{\Omega}^{\top}) \propto \rho(X_{\omega}\mathbf{1}_{\Omega}^{\top}) \Rightarrow \rho(Z) \leq \alpha\rho(X) + \bar{\alpha}\rho(Y)$$
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**Definition.** An aggregation function is a function  $\Lambda \colon \mathbb{R}^{\mathcal{F}} \to \mathbb{R}$  that is monotonic, positively homogeneous, convex, and normalized so that  $\Lambda(\mathbf{1}_{\mathcal{F}}) = |\mathcal{F}|$ .

Aggregation function: aggregates risk across firms in a given scenario

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Aggregation function: aggregates risk across firms in a given scenario

**Theorem.** A function  $\rho \colon \mathbb{R}^{\Omega \times \mathcal{F}} \to \mathbb{R}$  is a systemic risk measure with  $\rho(-\mathbf{1}_{\mathcal{E}}) < 0$  iff there exists

- $\bullet\,$  an aggregation function  $\Lambda\,$
- coherent single-firm base risk measure  $\rho_0$  such that

$$\rho(X) = (\rho_0 \circ \Lambda)(X) \triangleq \rho_0\left(\Lambda(X_1), \Lambda(X_2), \dots, \Lambda(X_{|\Omega|})\right)$$

### **Example: Economic Systemic Risk Measures**

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**Example.** (Systemic Expected Shortfall)

$$\Lambda_{\text{total}}(x) \triangleq \sum_{i \in \mathcal{F}} x_i, \qquad \rho_{\text{SES}}(X) \triangleq (\text{CVaR}_{\alpha} \circ \Lambda_{\text{total}})(X)$$

[Acharya et al., 2010; Brownlees, Engle 2010]

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**Example.** (Deposit Insurance)

$$\Lambda_{\mathsf{loss}}(x) \triangleq \sum_{i \in \mathcal{F}} x_i^+, \qquad \rho_{\mathsf{DI}}(X) \triangleq \mathsf{E}\left[\Lambda_{\mathsf{loss}}(X_\omega)\right] = \mathsf{E}\left[\sum_{i \in \mathcal{F}} X_{i,\omega}^+\right]$$

[e.g., Lehar, 2005; Huang et al., 2009]

## **Example: Investing with Performance Fees**

- $\mathcal{F} = a$  collection of hedge funds or portfolio managers
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$$\Lambda_{\mathsf{HF}}(x) \triangleq \sum_{i \in \mathcal{F}} \left( x_i + \gamma_i x_i^- \right)$$

+  $\gamma_{\rm FoF} \in [0,1]$  is the performance fee of the fund-of-funds manager

Example. (Fund-of-Funds Investor)

$$\Lambda_{\mathsf{FoF}}(x) \triangleq \sum_{i \in \mathcal{F}} \left( x_i + \gamma_i x_i^- \right) + \gamma_{\mathsf{FoF}} \left( \sum_{i \in \mathcal{F}} \left( x_i + \gamma_i x_i^- \right) \right)^-$$

## **Example: Resource Allocation**

- $\mathcal{A} = a$  set of activities
- $\mathcal{F} = a$  set of capacitated resources
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Consider the aggregation function:

$$\begin{split} \Lambda_{\mathsf{RA}}(x) &\triangleq & \min_{u} \text{inimize} & \sum_{a \in \mathcal{A}} c_a u_a \\ & \text{subject to} & \sum_{\substack{a \in \mathcal{A} \\ u \in \mathbb{R}^{\mathcal{A}}}} b_{ia} u_a \geq x_i, \quad \forall \ i \in \mathcal{F} \end{split}$$

where

- $u_a =$  reduction in level of activity a (decision variable)
- $c_a = per-unit cost of reductions in activity a$
- $b_{ia} = per-unit$  consumption of resource i by activity a

### **Example: Interbank Contagion Model**

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Consider the aggregation function:

$$\begin{split} \Lambda_{\mathsf{CM}}(x) &\triangleq & \underset{y \in \mathbb{R}_{+}^{\mathcal{F}}, \ b \in \mathbb{R}_{+}^{\mathcal{F}}}{\text{subject to}} & & \sum_{i \in \mathcal{F}} y_i + \gamma \sum_{i \in \mathcal{F}} b_i \\ & \text{subject to} & & b_i + y_i \geq x_i + \sum_{j \in \mathcal{F}} \Pi_{ji} y_j, \quad \forall \ i \in \mathcal{F} \end{split}$$

where

- loss x<sub>i</sub> is covered by firm i reducing the payments by an amount y<sub>i</sub>, or relying on an injection from the regulator in the amount b<sub>i</sub>
- reminiscent of Eisenberg & Noe (2001)

# "General" Aggregation Function

Given:

• 
$$c \in \mathbb{R}^N_+$$
,  $A \in \mathbb{R}^{K \times \mathcal{F}}_+$ ,  $B \in \mathbb{R}^{K \times N}$ 

•  $\mathcal{K} \subset \mathbb{R}^N$  a convex cone, such that  $\exists \ \bar{y} \in \mathcal{K}$  with  $B\bar{y} > \mathbf{0}$ 

Define:

$$\begin{array}{rll} \Lambda_{\mathsf{OPT}}(x) \triangleq & \underset{y}{\minimize} & c^\top y \\ & \text{subject to} & Ax \leq By \\ & y \in \mathcal{K} \end{array}$$

- $\Lambda_{\text{OPT}}$  is monotonic, positively homogeneous, and convex
- if  $\Lambda_{\mathsf{OPT}}(\mathbf{1}_\mathcal{F})>0,$  it can also be normalized
- allows for general endogenous mechanisms for 'co-operative' contagion

### **Structural Decomposition: Proof Sketch**

'If' part is not hard. 'Only if' part:

• Define 
$$\Lambda(x) \triangleq \rho(x \mathbf{1}_{\Omega}^{\top}), \ \forall \ x \in \mathbb{R}^{\mathcal{F}}$$

• Define  $\rho_0(z) \triangleq \rho(X), \ \forall \ z \in \mathcal{Q}^{\Omega}$ , for some  $X: \Lambda(X_{\omega}) = z_{\omega}, \ \forall \ \omega$ 

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- Step I:  $\rho_0$  is well-defined. Suppose X, Y have  $\Lambda(X_{\omega}) = \Lambda(Y_{\omega}), \forall \omega \in \Omega$ . Preference consistency of  $\rho$  implies  $\Lambda(X_{\omega}) > \Lambda(Y_{\omega}), \forall \omega \in \Omega \implies \rho(X) > \rho(Y),$

$$\Lambda(X_{\omega}) \leq \Lambda(Y_{\omega}), \ \forall \ \omega \in \Omega \quad \Longrightarrow \quad \rho(X) \leq \rho(Y).$$

Thus,  $\rho(X)=\rho(\,Y)$ 

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$$\begin{split} \Lambda(X_{\omega}) &\geq \Lambda(Y_{\omega}), \ \forall \ \omega \in \Omega \quad \Longrightarrow \quad \rho(X) \geq \rho(Y), \\ \Lambda(X_{\omega}) &\leq \Lambda(Y_{\omega}), \ \forall \ \omega \in \Omega \quad \Longrightarrow \quad \rho(X) \leq \rho(Y). \end{split}$$
  
Thus, 
$$\rho(X) = \rho(Y)$$

• Step 2: Derive other properties: monotonicity, convexity, homogeneity of  $\Lambda$  and  $\rho_0$ .

$$\rho = (\rho_0 \circ \Lambda)(X) \triangleq \rho_0 \left( \Lambda(X_1), \Lambda(X_2), \dots, \Lambda(X_{|\Omega|}) \right)$$

## **Acceptance Sets and Primal Representation**

**Theorem.** Any systemic risk measure  $\rho = (\rho_0 \circ \Lambda)$  can be expressed as

$$\begin{array}{ll} \rho(X) = & \underset{m,\ell}{\text{minimize}} & m\\ \text{(PRIMAL)} & & \text{subject to} & (m,\ell) \in \mathcal{A},\\ & & (\ell_{\omega}, X_{\omega}) \in \mathcal{B},\\ & & m \in \mathbb{R}, \ \ell \in \mathbb{R}^{\Omega}. \end{array} \quad \forall \ \omega \in \Omega, \end{array}$$

where acceptance sets  ${\cal A}$  and  ${\cal B}$  can be taken as the epigraphs of  $\rho_0$  and  $\Lambda,$  i.e.,

$$\mathcal{A} \triangleq \left\{ (m, z) \in \mathbb{R} \times \mathbb{R}^{\Omega} : m \ge \rho_0(z) \right\},$$
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- $\ell$  = regulator's position required to support the cross-sectional profile
- $m = \operatorname{cash} position required to support \ell$

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### **Dual Representation**

**Theorem.** Any systemic risk measure  $\rho$  can be expressed as

$$\rho(X) = \underset{\bar{\pi},\Xi}{\operatorname{maximize}} \sum_{i \in \mathcal{F}} \sum_{\omega \in \Omega} \Xi_{i,\omega} X_{i,\omega}$$
  
subject to  $(1,\bar{\pi}) \in \mathcal{A}^*$   
 $(\bar{\pi}_{\omega}, \Xi_{\omega}) \in \mathcal{B}^*, \ \forall \ \omega \in \Omega$   
 $\bar{\pi} \in \mathbb{R}^{\Omega}, \ \Xi \in \mathbb{R}^{\mathcal{F} \times \Omega}$ 

where  $\mathcal{A}^* \subset \mathbb{R} \times \mathbb{R}^{\Omega}$ ,  $\mathcal{B}^* \subset \mathbb{R} \times \mathbb{R}^{\mathcal{F}}$  are (up to a sign change) the dual cones to epi( $\rho_0$ ), epi( $\Lambda$ ). Further,  $(\bar{\pi}, \Xi)$  must satisfy  $\bar{\pi} \ge \mathbf{0}_{\Omega}, \quad \mathbf{1}_{\Omega}^{\top} \bar{\pi} \le 1, \quad \Xi \ge \mathbf{0}_{\mathcal{E}}, \quad \mathbf{1}_{\mathcal{F}}^{\top} \Xi \le |\mathcal{F}| \bar{\pi}^{\top}$ 

**Robust optimization interpretation:**  $\rho(X)$  is worst-case expected loss of a rescaled economy over a set of probability distributions and scaling functions

$$\begin{split} \rho(X) &= \max_{\bar{\pi}, \Xi} \max_{i \in \mathcal{F}} \sum_{\omega \in \Omega} \Xi_{i, \omega} X_{i, \omega} \\ \text{(DUAL)} \qquad \qquad \text{subject to} \quad \begin{array}{l} (1, \bar{\pi}) \in \mathcal{A}^* \\ (\bar{\pi}_{\omega}, \Xi_{\omega}) \in \mathcal{B}^*, \ \forall \ \omega \in \Omega \\ \bar{\pi} \in \mathbb{R}^{\Omega}, \ \Xi \in \mathbb{R}^{\mathcal{F} \times \Omega} \end{array} \end{split}$$

**Corollary.** If  $(\bar{\pi}^*, \Xi^*)$  is an optimal solution to (DUAL), then  $\Xi^*$  is a subgradient of  $\rho$  at X.

 $\Xi^*_{i,\omega}$  is the shadow price of risk  $\equiv$  the minimum marginal rate of increase of systemic risk given an increase of losses for firm i in scenario  $\omega$ 

## **Risk Attribution**

Suppose  $\rho$  is a systemic risk measure, and  $\Xi^*$  is an optimal solution to (DUAL) at X. Define the risk attribution of firm i as

$$y_i^* = \sum_{\omega \in \Omega} \Xi_{i,\omega}^* X_{i,\omega}$$

By strong duality,

$$\rho(X) = \sum_{i \in \mathcal{F}} y_i^*$$

## **Risk Attribution**

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Theorem. (No Undercut) Given  $\alpha \in \mathbb{R}_+^{\mathcal{F}}$ , define  $r(\alpha) \triangleq \rho(\alpha_1 x_1; \dots; \alpha_{|\mathcal{F}|} x_{|\mathcal{F}|})$ Then,  $\alpha^\top y^* \leq r(\alpha)$ 

Generalization of attribution scheme of Aumann & Shapley (1974) or Denault (2001); Buch & Dorfleitner (2008).

- $X^{(i)} =$ outcomes of firm  $i, X \triangleq (X^{(1)}; X^{(2)}; \dots; X^{(|\mathcal{F}|)})$
- $\mathcal{T}_i = \mathsf{set} \mathsf{ of outcomes of firm } i, \mathcal{T} = \mathcal{T}_1 imes \mathcal{T}_2 \ldots imes \mathcal{T}_{|\mathcal{F}|}$

### **Definition.** (Social Optimality) An economy $\bar{X} \in \mathcal{T}$ is socially optimal if it maximizes risk-adjusted welfare

$$\underset{X \in \mathcal{T}}{\text{maximize}} \sum_{i \in \mathcal{F}} U_i(X^{(i)}) - \tau \rho(X)$$

Here,  $\tau > 0$  captures the systemic risk externality.

## Decentralization

**Theorem.** Suppose that  $\bar{X} \in \mathcal{T}$  is a socially optimal economy. There exists  $\Xi^*$  that is an optimal solution to the dual problem for  $\rho(\bar{X})$  so that if we define, for each firm i, the tax function

$$t_i(X^{(i)}) \triangleq \tau \sum_{\omega \in \Omega} \Xi^*_{i,\omega} X_{i,\omega}$$

then,  $\bar{X}^{(i)}$  is an optimal outcome for firm i, i.e.,

$$\bar{X}^{(i)} \in \operatorname*{argmax}_{X^{(i)} \in \mathcal{T}^{(i)}} \ U_i(X^{(i)}) - \tau \sum_{\omega \in \Omega} \Xi^*_{i,\omega} X_{i,\omega}$$

Decentralized computation of optimal taxes possible

### **Planner's problem:**

$$\underset{X \in \mathcal{T}}{\text{maximize}} \sum_{i \in \mathcal{F}} U_i(X^{(i)}) - \rho(X).$$

Decentralization scheme: apply proximal gradient method

$$\underset{X \in \mathcal{T}}{\text{maximize}} \sum_{i} U_{i}(X^{(i)}) - \left(\rho(\bar{X}) + \sum_{i} \left(X^{(i)} - \bar{X}^{(i)}\right)^{\top} \frac{\partial \rho(\bar{X})}{\partial X^{(i)}} + \frac{t}{2} \|X - \bar{X}\|_{2}^{2}\right)$$

### **Planner's problem:**

$$\underset{X \in \mathcal{T}}{\text{maximize}} \sum_{i \in \mathcal{F}} U_i(X^{(i)}) - \rho(X).$$

Decentralization scheme: apply proximal gradient method

$$\underset{X \in \mathcal{T}}{\text{maximize}} \sum_{i} U_i(X^{(i)}) - \left(\rho(\bar{X}) + \sum_{i} \left(X^{(i)} - \bar{X}^{(i)}\right)^\top \frac{\partial \rho(\bar{X})}{\partial X^{(i)}} + \frac{t}{2} \|X - \bar{X}\|_2^2\right)$$

Individual firm's problem:

$$X_{i}^{*} = \underset{X^{(i)} \in \mathcal{T}_{i}}{\operatorname{argmax}} \left\{ U_{i}(X^{(i)}) - (X^{(i)} - \bar{X}^{i})^{\top} \frac{\partial \rho(\bar{X})}{\partial X^{(i)}} - \frac{t}{2} \|X^{(i)} - \bar{X}^{(i)}\|_{2}^{2} \right\}$$

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$$\underset{X \in \mathcal{T}}{\text{maximize}} \sum_{i \in \mathcal{F}} U_i(X^{(i)}) - \rho(X).$$

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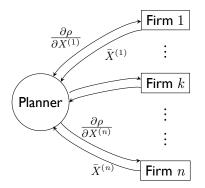
$$\underset{X \in \mathcal{T}}{\text{maximize}} \sum_{i} U_i(X^{(i)}) - \left(\rho(\bar{X}) + \sum_{i} \left(X^{(i)} - \bar{X}^{(i)}\right)^\top \frac{\partial \rho(\bar{X})}{\partial X^{(i)}} + \frac{t}{2} \|X - \bar{X}\|_2^2\right)$$

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Information sent by the planner:  $\frac{\partial \rho(X)}{\partial X^{(i)}}$ Information sent by the firm:  $\bar{X}^{(i)}$ 

### **Communication between the planner and firms**



### Homogeneous Systemic Risk Measures:

- monotone, +vely homogeneous, preference consistent, **not** convex
- structural decomposition exists homogeneous single-firm base risk measure homogeneous aggregation function

### **Convex Systemic Risk Measures:**

- monotone, convex, preference consistent, **not** +vely homogeneous
- structural decomposition exists convex single-firm base risk measure convex aggregation function

Key idea: Preference consistency allows for the structural decomposition

# **Conclusions / Future Directions**

- A general, axiomatic framework for coherent systemic risk analyzes joint distribution of outcomes allows for 'cooperative' endogenous forms of contagion potential applications in a broad array of engineering & economic systems
- A structural decomposition of systemic risk
- Mechanisms for systemic risk attribution & decentralization

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### **Future Directions:**

- Statistical estimation of systemic risk
- Strategic mechanisms of contagion
- Is network structure important for systemic risk in financial markets?