

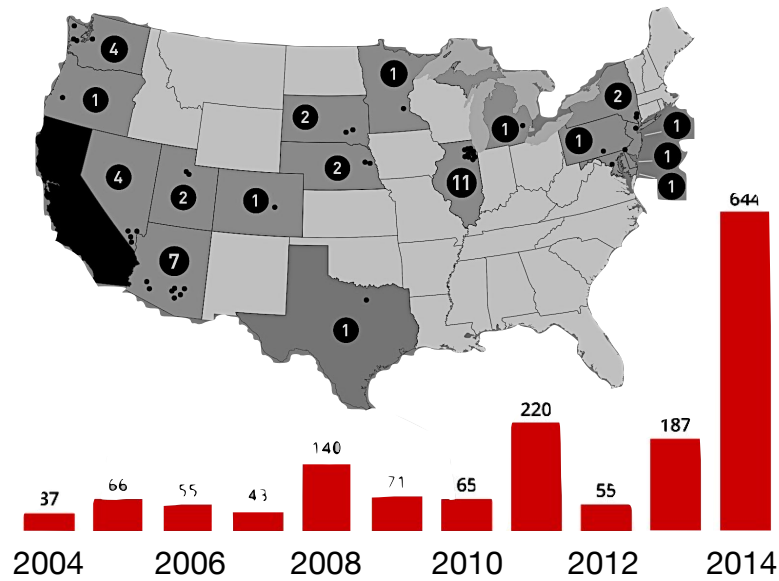
Strategic Investment in Protection in Networked Systems

*Matt V. Leduc (IIASA, Stanford) and **Ruslan Momot** (INSEAD)*



EXAMPLES OF NETWORKED SYSTEMS IN WHICH INDIVIDUAL INCENTIVES MATTER

Measles outbreak in US 2014-2015



*“While I think it’s a good idea to take the vaccine, I think that’s a **personal decision** for individuals”*

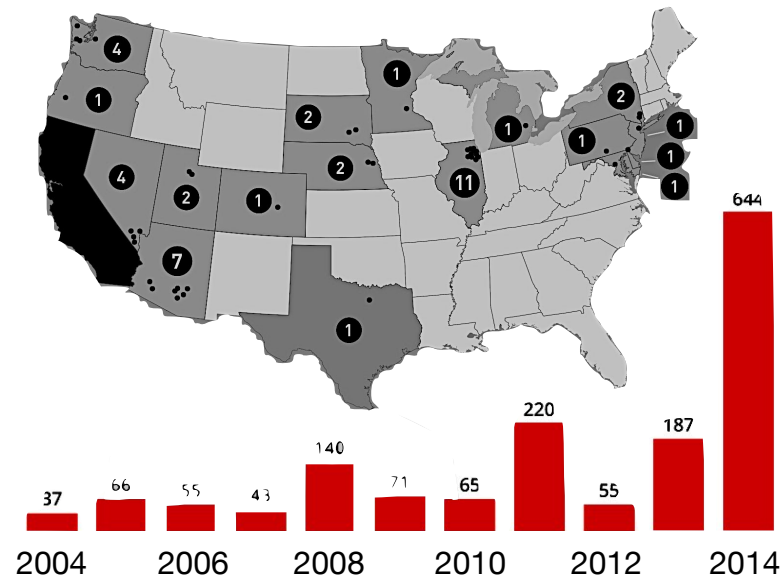
Senator Rand Paul of Kentucky

*“There is absolutely **no reason to get the shot**. I said, ‘I’d rather you miss an entire semester than you get the shot.’ “*

Mother of a 16-year-old student

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Paris Attacks, Nov 2015



“The European Union will step up checks on its citizens traveling abroad, tighten gun control and collect more data on airline passengers”

“David Cameron is to respond to the escalation in terror attacks around the world by making provisions for 1,900 extra security and intelligence staff and doubling funds for aviation security.”

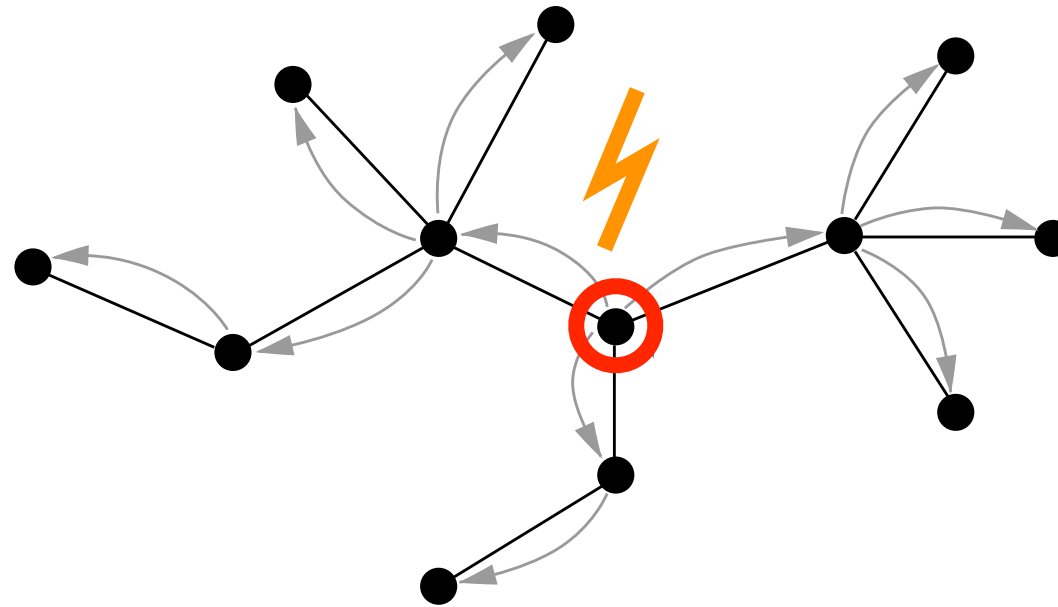
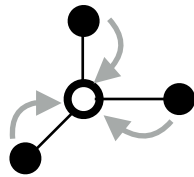
RESEARCH QUESTION

2 ways to fail:

intrinsic failure



cascade of failures



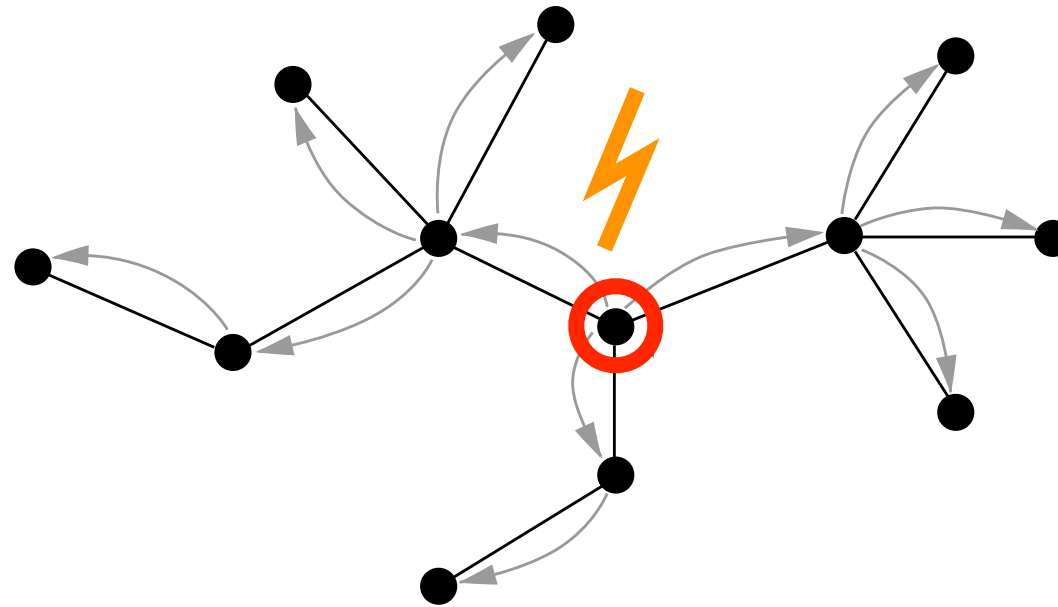
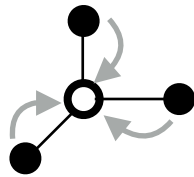
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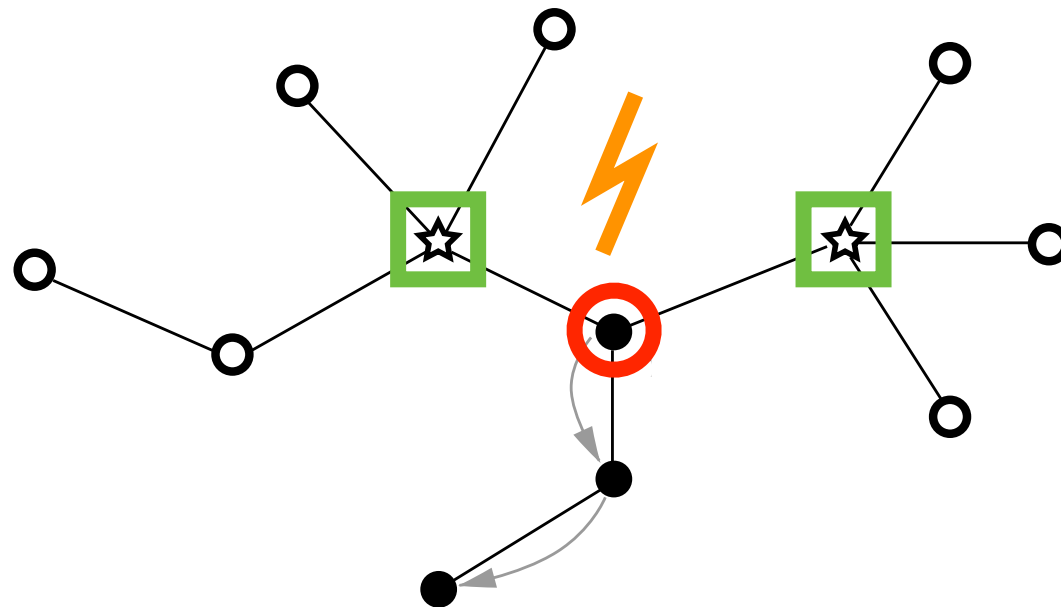
intrinsic failure



cascade of failures



Agents can invest in costly protection



*What are the strategic incentives of **agents** to invest in **costly protection**?
How does the **network structure** influence these decisions?*

LITERATURE

Network Games

- *Galeotti et al., 2010*
- *Jackson and Yariv, 2007*
- *Kearns, 2007*
- *Jackson and Zenou, 2014*

Interdependent Security (IDS)

- *Heal and Kunreuther, 2005*
- *Heal et al., 2006*
- *Johnson et al., 2011*

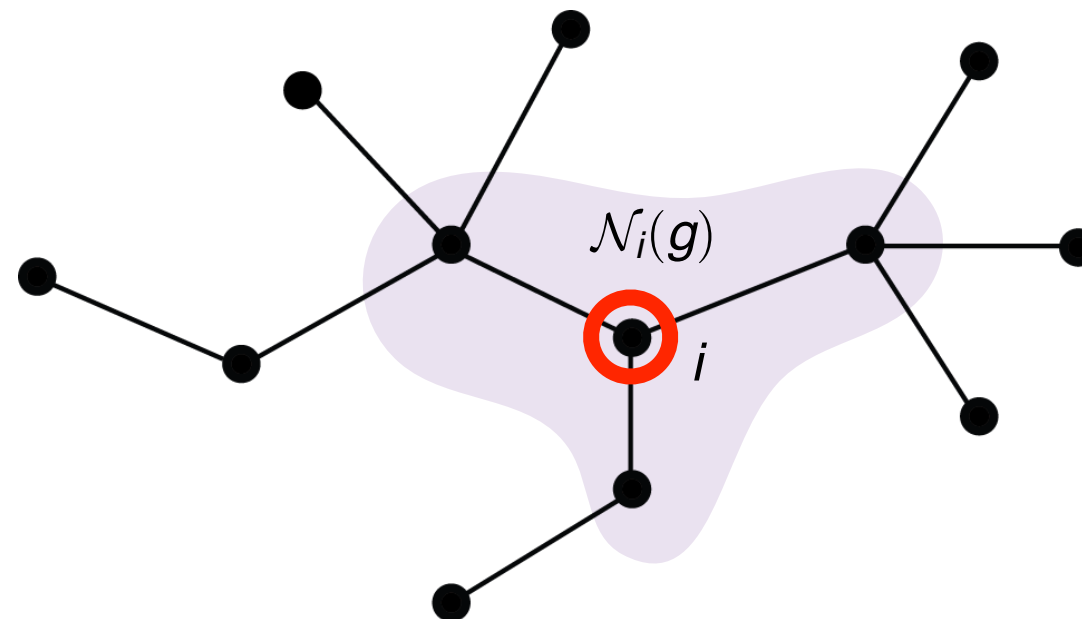
Cascade Risk in Networks

- *Lelarge, Bolot, 2008, 2009*
- *Galeotti, Rogers, 2013*
- *Dziubinski, Goyal, 2014*
- *Goyal, Vigier, 2014*
- *Blume et al., 2011*

Contribute to the literature on strategic investments in protection in complex interconnected systems.

MODEL OVERVIEW

Network - **nodes** (agents) and **edges** (interconnections)



$\mathcal{N}_i(g)$ - neighborhood of agent i
 $d_i(g) = |\mathcal{N}_i(g)|$ - degree of agent i

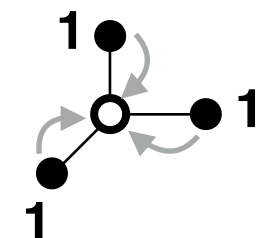
Network - nodes (agents) and edges (interconnections)

agents can fail:

intrinsically (ext)



cascade of failures



probabilities:

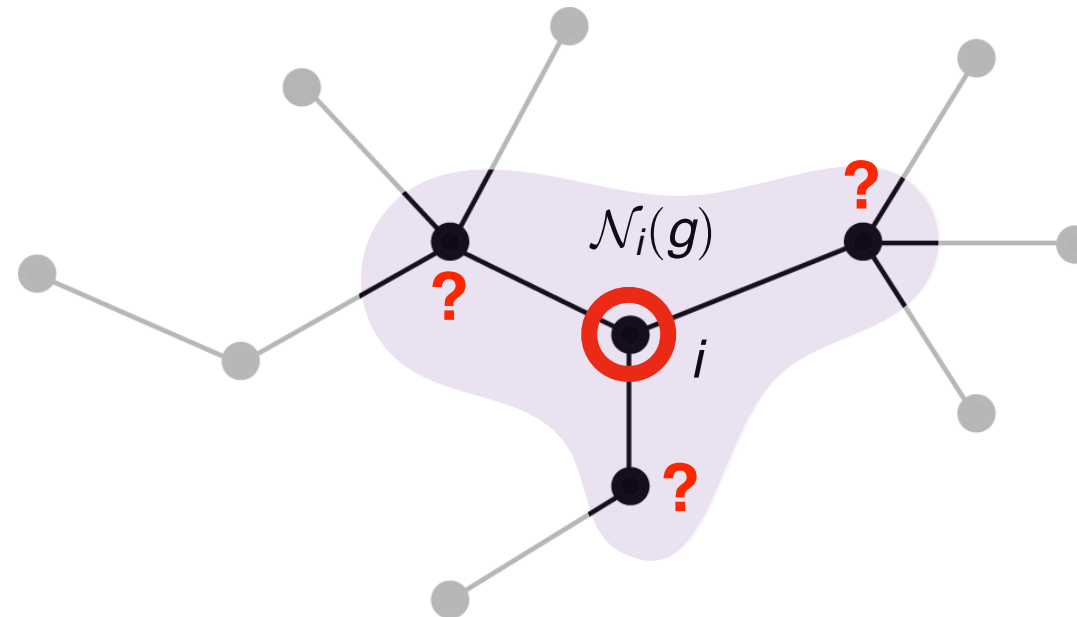
p

$\mathcal{H}_d : \{0, 1\}^d \rightarrow [0, 1]$

cascading failure function:
 vector of binary variables (**friend** failed/didn't fail) \rightarrow
 agent's probability to fail

This model leads to BNE - hard to work with it. Can only prove existence of eq.

MEAN-FIELD MODEL

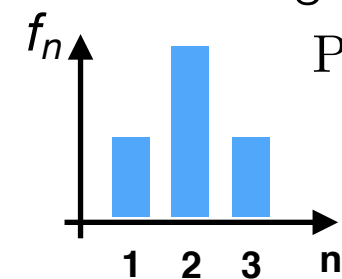


Agent i knows his **own** degree and:

doesn't know full network structure

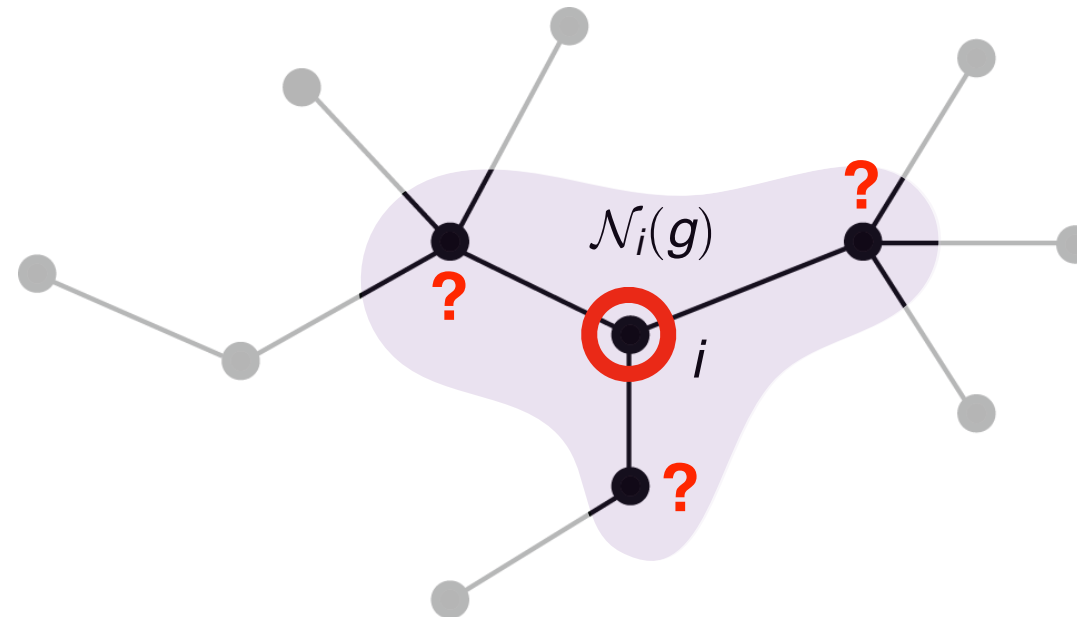
but

knows the degree distribution it is drawn from



$\Pr[\text{an agent has degree } d] = f_d$
 $\{f_1, f_2, \dots\}$

MEAN-FIELD MODEL

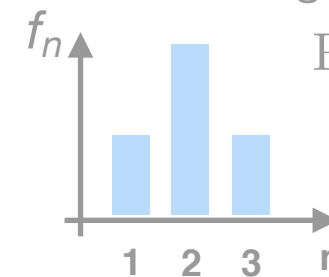


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doesn't know each friend's degree

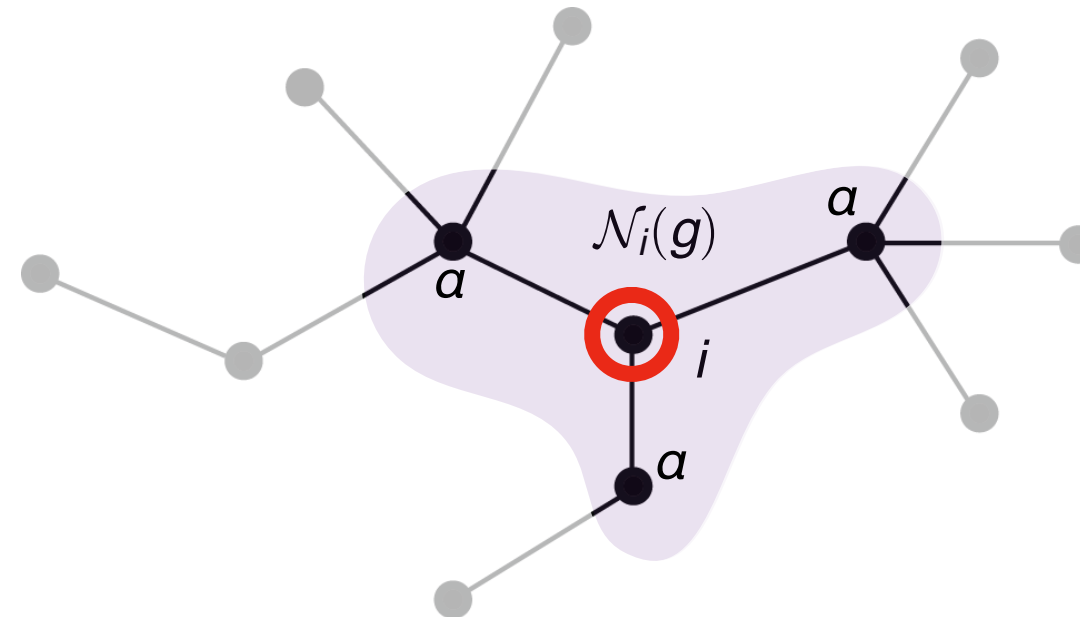
but

knows edge-perspective degree distribution

$$\Pr[\text{a **friend** has degree } d] = \tilde{f}_d$$

$$\{\tilde{f}_1, \tilde{f}_2, \dots\}$$

MEAN-FIELD MODEL

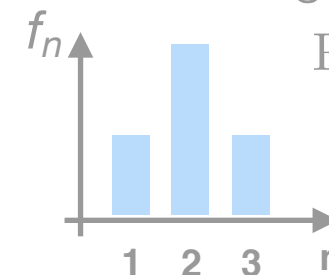


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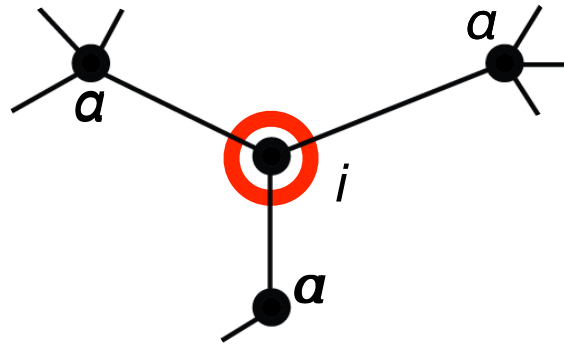
doesn't know each friend's failure probability

but

conjectures that each friend fails with the same probability (*bounded rationality*)

$$\Pr[\text{a **friend** fails}] = a$$

TOTAL PROBABILITY TO FAIL



Model of accumulative risk:

agent's cascading failure probability: $q_d : [0, 1] \rightarrow [0, 1]$

$$q_{d'}(a) > q_d(a), \forall d' > d$$

more connections - **higher** risk

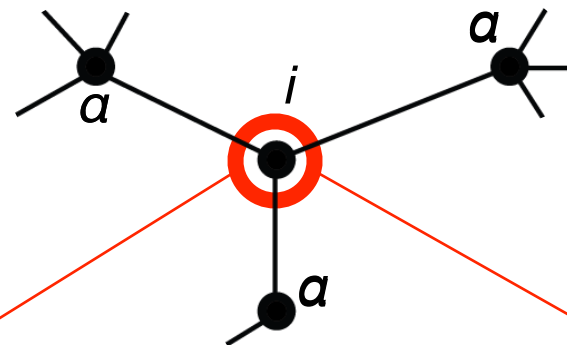
example: malware or virus spread

$$q_d(a) = 1 - (1 - ra)^d$$

virus is transmitted with r probability

$$\text{Total probability to fail: } \beta_d = p + (1 - p)q_d$$

DECISIONS



invest in protection

VS

don't invest in protection

$$a_i = 1$$

$$U(a_i = 1, a) = -V \cdot \mathcal{B}(p, q_d(a), a_i = 1) - C$$

cost of protection

$$a_i = 0$$

$$U(a_i = 0, a) = -V \cdot \mathcal{B}(p, q_d(a), a_i = 0)$$

effective probability to fail

mean-field strategy

for each degree-type specifies probability to invest in protection

$$\mu : \mathbf{N}^+ \rightarrow [0, 1]$$

MEAN-FIELD EQUILIBRIUM

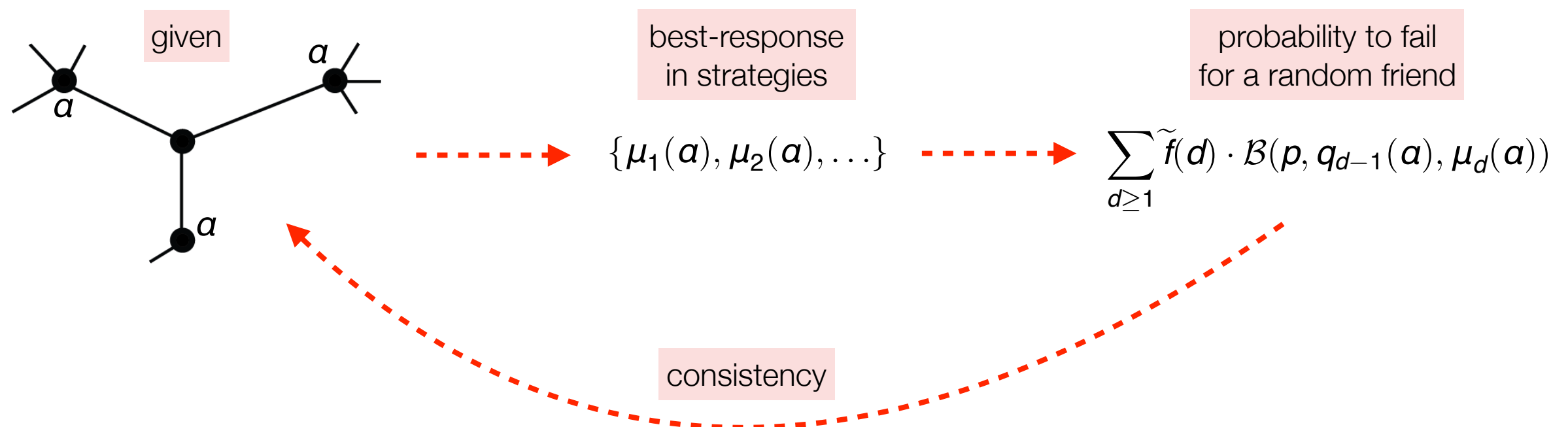
We are searching for:

- mean-field local probabilities to fail
- set of strategies for each degree-type

$$(\alpha^*, \mu^* = \{\mu_1^*, \mu_2^*, \dots\})$$

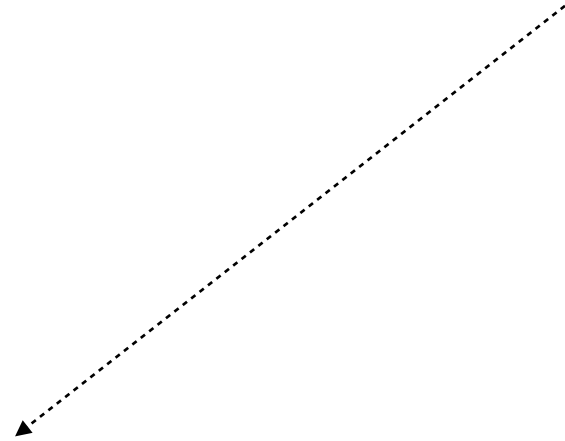
Fixed point argument:

α^* must be induced by the mean-field strategies μ^* that are BR to α^*



*Th: there **exists** a mean-field equilibrium*

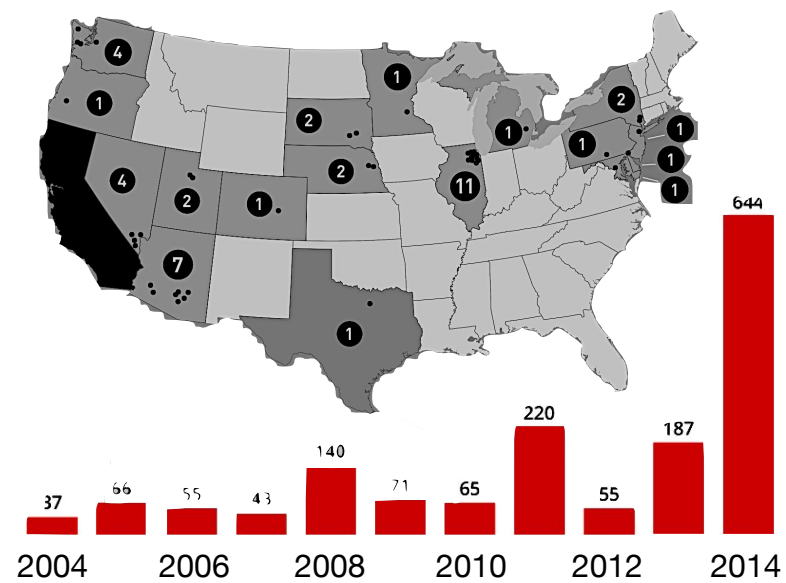
WHAT DOES PROTECTION DO?



insulates against **total** risk

games of total protection

$$\mathcal{B}(p, q_d(a), a) = \left(p + (1 - p)q_d(a) \right) \cdot (1 - ka)$$



Examples:

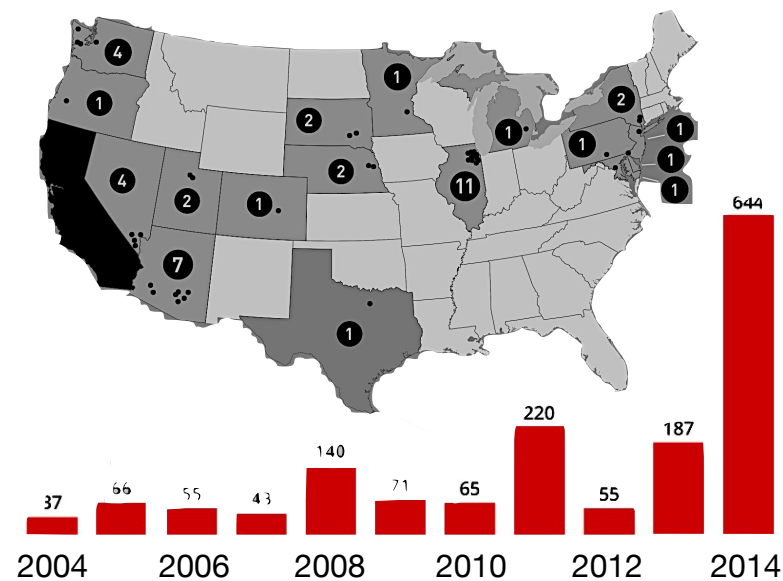
- computer antivirus software
- vaccination against measles

WHAT DOES PROTECTION DO?

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$$\mathcal{B}(p, q_d(a), a) = \left(p + (1 - p)q_d(a) \right) \cdot (1 - ka)$$



Examples:

- computer antivirus software
- vaccination against measles

insulates against **intrinsic** risk **only**

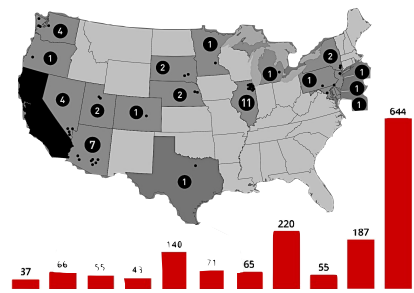
games of self protection

$$\mathcal{B}(p, q_d(a), a) = p \cdot (1 - ka) + (1 - p \cdot (1 - ka))q_d(a)$$



Examples:

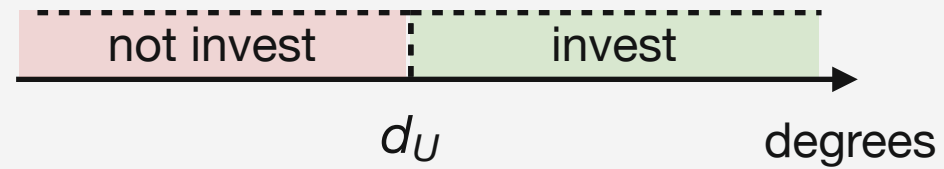
- investing in airport security
- investing in national security within EU

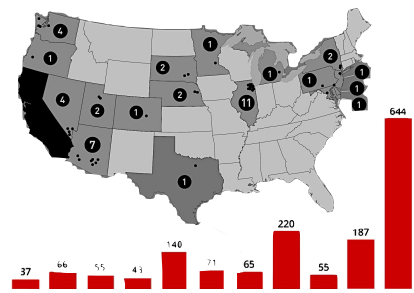


EQUILIBRIUM: TOTAL PROTECTION

submodular game (**strategic substitutes**)

Th: equilibrium is **unique** and only sufficiently connected agents invest in protection (**upper-threshold strategy**).

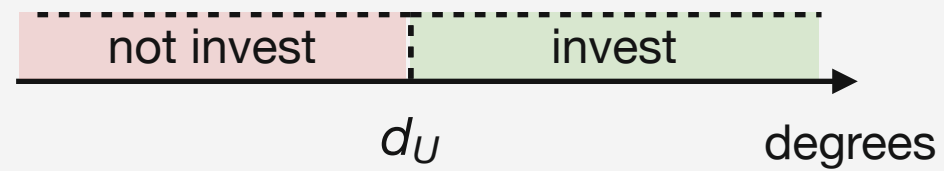




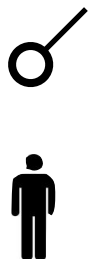
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low connected agent

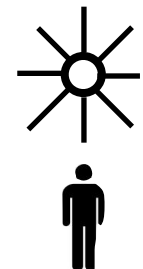


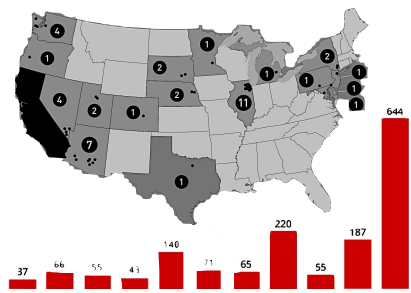
$$d_L < d_H$$

intrinsic risk = intrinsic risk

cascading failure risk < cascading failure risk

high connected agent

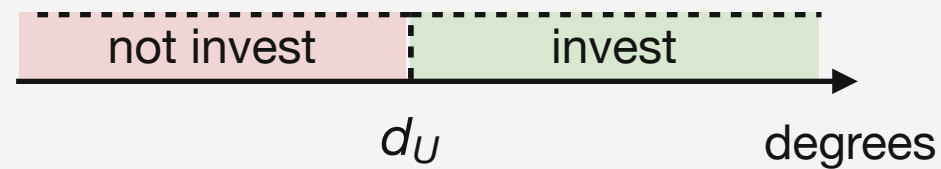




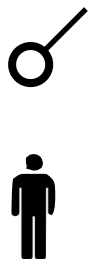
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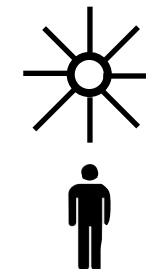


low connected agent



$d_L < d_H$

high connected agent



intrinsic risk

=

intrinsic risk

cascading failure risk

<

cascading failure risk

Total Protection

incentive to invest in total protection

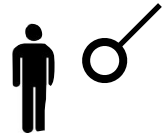
<

incentive to invest in total protection



EQUILIBRIUM: SELF PROTECTION

supermodular game (**strategic complements**)



Self Protection



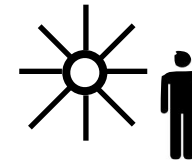
intrinsic risk
cascading failure risk



intrinsic risk



cascading failure risk



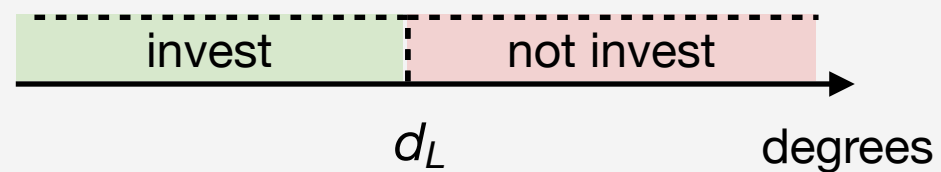


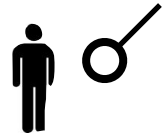
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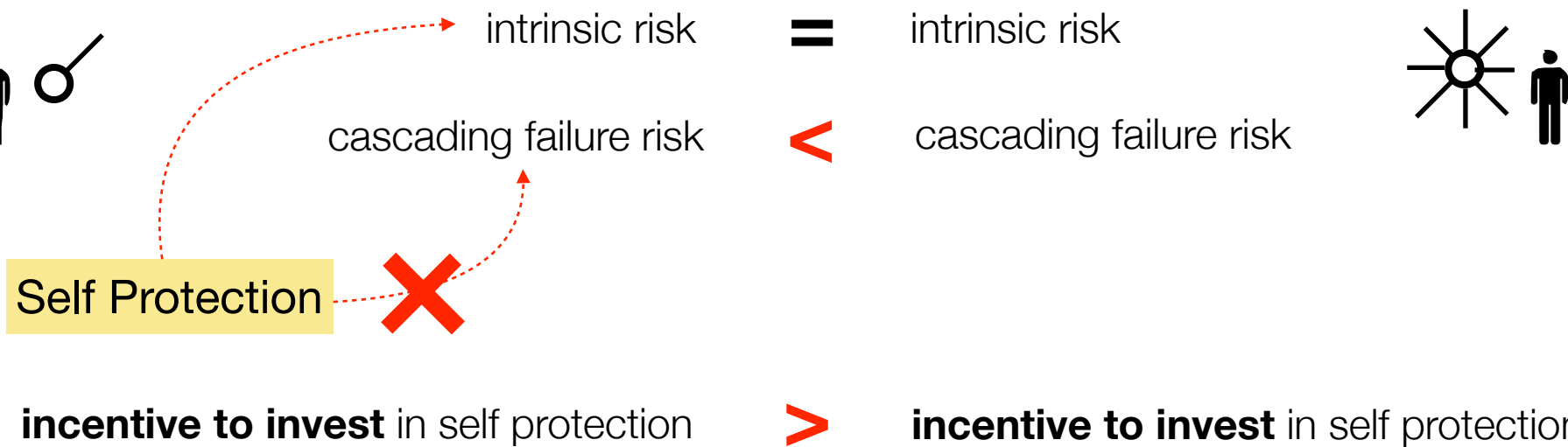
Th: in equilibrium only low connected agents invest in protection (**lower-threshold strategy**).



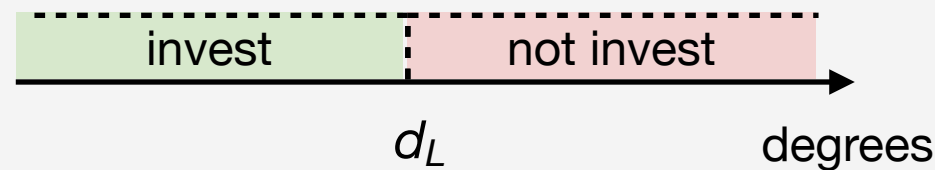


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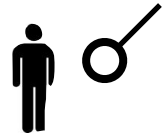


Proposition:

equilibrium effective failure probability $\mathcal{B}(p, q_d(\alpha^*), \mu^*(d))$ < equilibrium effective failure probability $\mathcal{B}(p, q_d(\alpha^*), \mu^*(d))$

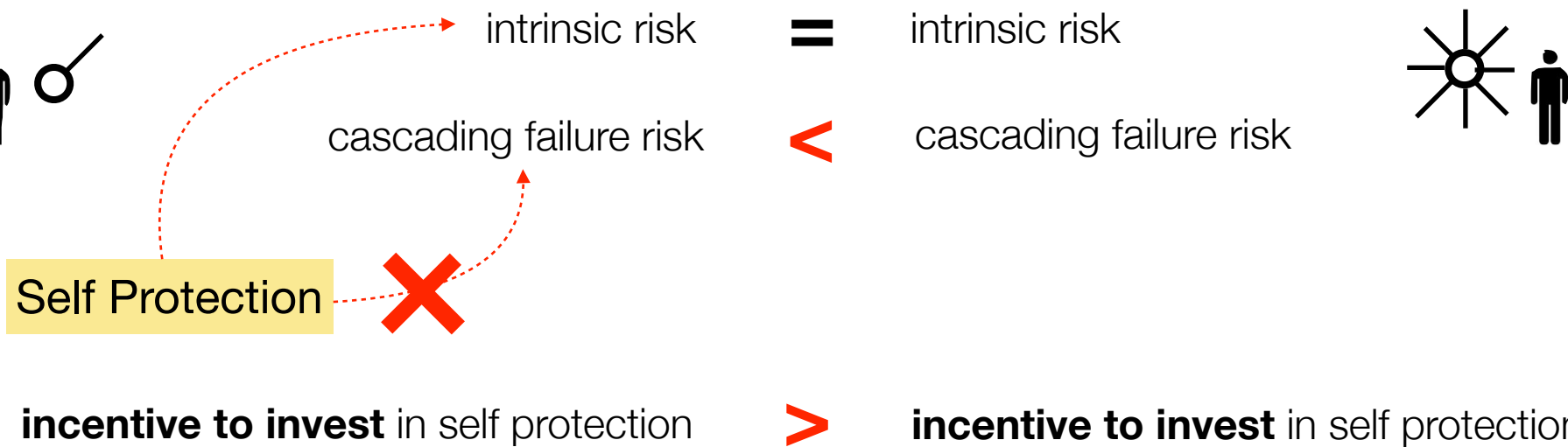
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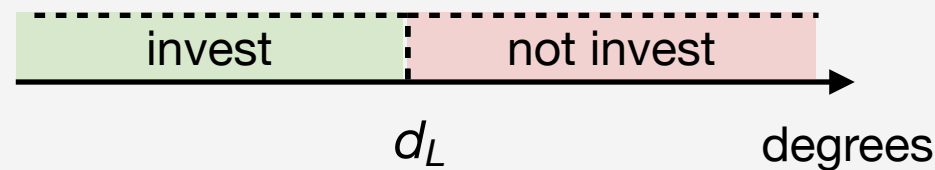


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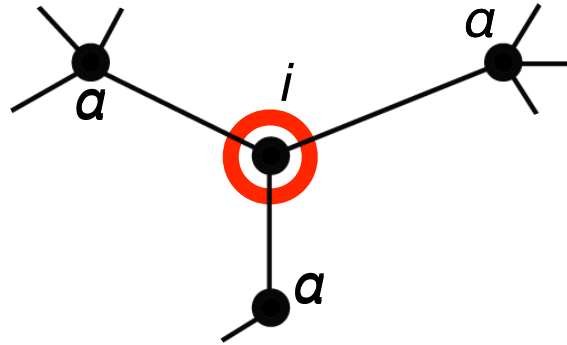
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Proposition:

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Proposition: under FOSD increase in \tilde{f}_d incentives to invest in costly protection are lower.

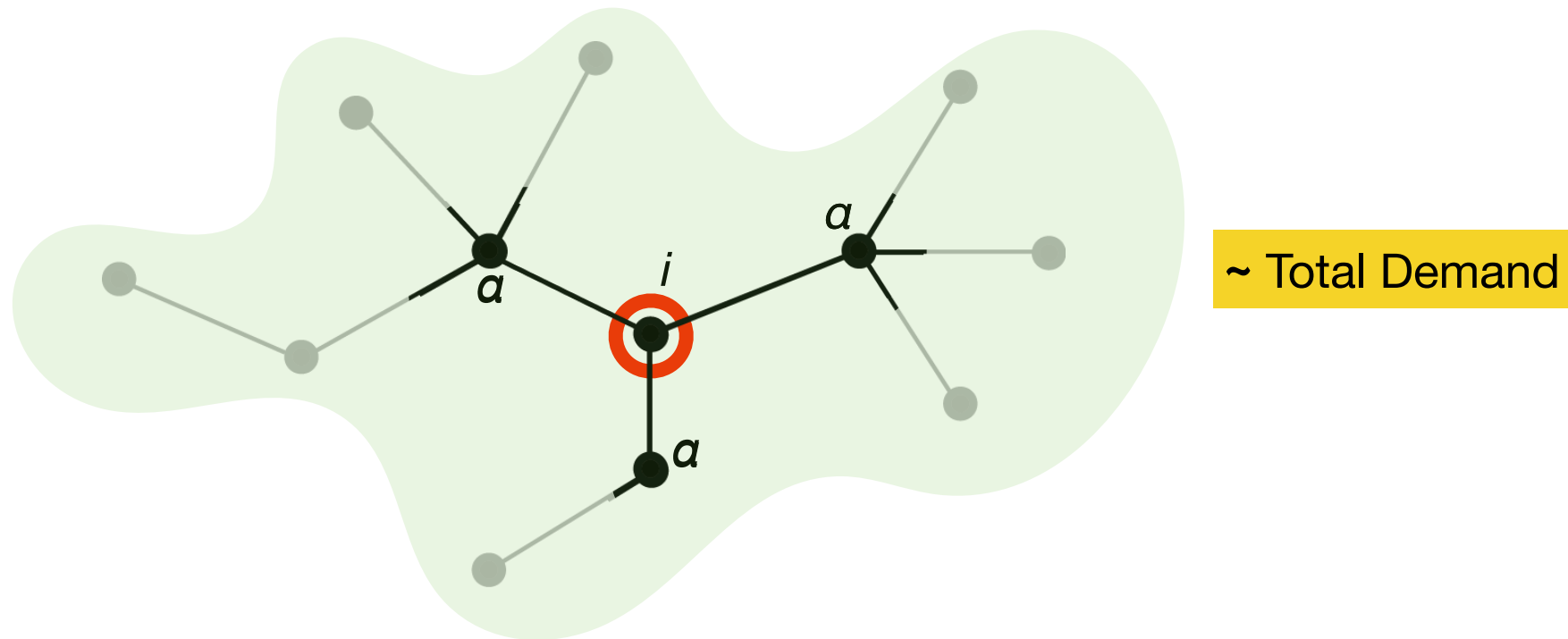
GLOBAL & LOCAL EXTERNALITIES



$$U(a_i = 1, a) = -V \cdot \mathcal{B}(p, q_d(a), a_i = 1) - C$$

cost of protection

GLOBAL & LOCAL EXTERNALITIES



$$U(a_i = 1, \mathbf{a}) = -V \cdot \mathcal{B}(p, q_d(\mathbf{a}), a_i = 1) - \mathcal{C}(\text{Demand})$$

cost of protection

Th: The threshold characterization of equilibria is robust to the introduction of a global price feedback

Th: In a game of total protection with global price feedback, the mean-field equilibrium is unique if C is an increasing function.

CONCLUSIONS

- Incentives to protect depend on **both** the type of protection **and** network structure.
- Market failure is more severe in case of self-protection (EU security, airport security) than in case of total protection (vaccination, malware)
 - ➔ Incentives of the agents are aligned with the system's efficient outcome
- We employ a mean-field equilibrium concept that places a reasonable cognitive burden on the agents.
- Model is flexible and allows for:
 - comparative statics in the structure of the network
 - introduction of global externalities.