# Collaboration in social networks: Incentives and topology

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# The puzzle of cooperation

Why do we see so much cooperation around? 

Failed states, why do societies collapse?  $\mathbf{a}$ 

Will Euro collapse if Greece drops out? 







## Much has been written on the emergence of cooperation on networks

- Repeated games, reputation and trust (Myerson 1991)  $\blacksquare$
- Endogenous network games (Vega-Redondo 2007, Jackson 2008, Goyal 2009)
- Repeated games on evolving networks (Ellison 1994, Haag Lagunoff  $\blacksquare$ 2006, Vega-Redondo 2006).
- Cooperation in evolutionary games without mutation (Boyd 1999,  $\bullet$ Hofbauer Sigmund 2003, Poncela et al 2010)
- Repeated games and punishment on specific structures (Eshel et al  $\mathbf{H}$ 1998, Haag Lagunoff 2007, Fainmesser 2009, Karlan et al 2009)
- Focus here: social network  $=$  pattern of repeated interactions E. repeated interaction = forward looking behavior collaboration  $=$  incentives  $+$  credibility of threats How difficult is this in large games on complex structures?

# **Outline**

- The prisoners dilemma
- Collaboration in repeated interaction: 2 players ×
	- Collaboration is supported by credible threats of punishment  $\blacksquare$
- Collaboration in N players games on a network: Local contribution game  $\bullet$ 
	- Conditional collaboration has to be reciprocal and limited to a subset of neighbors  $\bullet$
- How does collaboration depend on incentives and topology?  $\hat{\mathbf{z}}$ 
	- Collaborative equilibria are subgraphs of the social network
- The complexity of collaboration:  $\blacksquare$ 
	- Counting collaborative equilibria with message passing ×
- **Conclusions**  $\blacksquare$

## Defection is the only possible outcome in one shot prisoner's dilemma



Each player either cooperates (C) or defects (D) with all neighbors

Payoff: 1 for each neighbor that collaborates minus  $X_i$  (=cost of collaboration)

All D (s<sub>i</sub>=0) is the only Nash equilibrium  $s_i = 0, 1$ 



 $u_i(s_i,s_{-i}) = -X_is_i + \sum$ 

 $j \in \partial_i$ 

*sj*

## N=2: When the game is played many times cooperation is possible, among other things

Strategies become plans of actions, decided at time 0, to optimize future payoffs  $\blacksquare$ 

#### $U_i = (1-d)$ ⇤  $\infty$ *t*=0  $d^t u_i$  $\overline{1}$ *s*  $\binom{(t)}{i}, s\binom{(t)}{i}$ ⇥  $d \in [0, 1]$

- Cooperation under trigger strategies T:  $\blacksquare$  $T = \{$ start with C;
	- C as long as opponent plays C,
	- D forever, if opponent plays D}

If d is large enough, (T, T) is a Nash equilibrium

- Folk's theorem: many other outcomes can be  $\blacksquare$ supported as a Nash equilibrium
- ! d=1 in what follows  $\mathbf{r}$



# But threats should be credible

#### $N=3$

- Is it credible that 1 and 2 punish 3?
- $\bullet$  Not if  $u_1(C, C, D) > u_1(D, D, D)$ !
- Players need to condition C only to a subset of their neighbors
- **If i conditions on j, j should condition on i**
- Emergent heterogeneity





## On trees, Nash equilibria are subtrees

- Given an undirected tree G=(N,L)  $k_i = |\partial_i|$  = degree of node i  $m_i$  = smallest integer larger than  $X_i$  $c_i$  = number of collaborators in  $\partial_i$
- Any collection of disjoint undirected subgraphs Γ=(V,Λ) of G is a collaborative equilibrium where all i∈V cooperate conditionally to neighbors in Γ and |∂i ∩ Λ|=mi
- Incentives:  $i \in V$  c<sub>i</sub> X<sub>i</sub>  $\geq$  c<sub>i</sub>-m<sub>i</sub>  $\Rightarrow$  m<sub>i</sub>  $\geq$  X<sub>i</sub>  $\blacksquare$
- Reciprocity: i,j∈V, if j does not punish i  $\blacksquare$  $\Rightarrow$  i should not punish j when j defects
- Credibility:  $\blacksquare$ i,k∈V, (i,k)∈ $\Lambda$  if k defects c<sub>i</sub> - 1 - X<sub>i</sub> < c<sub>i</sub> - m<sub>i</sub>  $\Rightarrow$  m<sub>i</sub> < X<sub>i</sub> +1

### $u_i(s_i, s_{-i}) = c_i - X_i s_i$



## On generic graphs cascades of defection make things more complex

- Indirect defections: As a result of the defection of j∈∂i other neighbors K k∈∂i may also defect because of loops
- A collection of disjoint undirected subgraphs  $\Gamma = (V, \Lambda)$  of G is a  $\blacksquare$ collaborative equilibrium where all i∈V cooperate conditionally to neighbors in  $\Gamma$  and  $|\partial_i \cap \Lambda| = m_i$  provided i) the indirect effects caused by the defection of all j∈ $\partial_i \cap \Lambda$  have the same consequence of the defection of i itself.
- $\bullet$  i) holds provided removing i from V does not disconnect Γ
	- Works on trees, for dimers and loops, ig. for the complete graph
	- Likely works on random graphs and on  $\blacksquare$ dense graphs



Fig. s1 A counter example

# Nash equilibria on random graphs



### Do NE exist? How many? How hard is it to find them?

Circuits: Marinari, Monasson, Semerjian 2006 q-regular subgraphs: Pretti, Weigt 2006

# Counting NE by message passing

i

j

#### **■**  $x_{i\rightarrow i} = 1$  if i conditions C on j,  $x_{i\rightarrow i} = 0$  otherwise

- there are m<sub>i</sub>-1 k  $\in$  ∂i/j with  $x_{k\rightarrow i}=1$   $\Rightarrow$   $x_{i\rightarrow j}=1$
- $-m_i$  k  $\in \partial_i$  with  $x_{k\rightarrow i}=1 \Rightarrow x_{i\rightarrow j}=0$
- $-$  no  $k \in \partial_i / j$  with  $x_{k\rightarrow i} = 1 \Rightarrow x_{i\rightarrow j} = 0$

### **\*** Marginals:

## $\mu_{i \to j} = P\{i \in V, i \text{ punishes } j\}$

Circuits: Marinari, Monasson, Semerjian 2006 q-regular subgraphs: Pretti, Weigt 2006

## Message passing equations:

$$
\mu_{i \to j} = \frac{e^{-\epsilon} Z_{N_i \backslash j \to i}^{m_i - 1}}{Z_{N_i \backslash j \to i}^0 + e^{-\epsilon} Z_{N_i \backslash j \to i}^{m_i - 1} + e^{-\epsilon} Z_{N_i \backslash j \to i}^{m_i}} \\
Z_{V \to i}^q = \sum_{U \subseteq V} \mathbb{I}_{|U| = q} \prod_{j \in U} \mu_{j \to i} \prod_{k \in V/U} (1 - \mu_{k \to i}) \\
P\{i \in C\} = \frac{e^{-\epsilon} Z_{N_i \to i}^{\gamma_i}}{Z_{N_i \to i}^0 + e^{-\epsilon} Z_{N_i \to i}^{\gamma_i}} \\
P\{i \in \Gamma_j\} = \frac{\mu_{i \to j} \mu_{j \to i}}{\mu_{i \to j} \mu_{j \to i} + (1 - \mu_{i \to j})(1 - \mu_{j \to i})}.
$$

Fixed point  $\Rightarrow$  number of subgraphs (entropy)

Circuits: Marinari, Monasson, Semerjian 2006 q-regular subgraphs: Pretti, Weigt 2006

k

#### Regular random graphs: dimers *0 -4 -2 0 2 4* ε *0.2*  $(m<sub>i</sub>=1)$

as a function of K for m = 1.

#### **Exponentially many NE's**  $D$  , and consideration instability  $\mathcal{C}$  , and  $\mathcal{C}$  , and  $\mathcal{C}$  and

Figure 5: Sprectrum of possible equilibria (m = 1) in a RRG with K = 4.





larger values of K we have a similar behavior (see Figs.5-6 for K = 4).

 $\bullet$   $\rho$  = fraction of cooperators  $s(\rho) = log(number of NE|\rho)/N$ 

 $p(A|A)$  $\mathsf{D}$ )/N NE ∃ ∀ρ∈[0,1] edges in the graph. This is at the origin of the degeneracy s(ρ = 1) > 0. For

#### Regular random graphs: circuits  $(m<sub>i</sub>=2)$ *-4 -2 0 2 4* ε *0.2*

as a function of K for m = 2.

Exponentially many NE's Density of cooperators ρ("), entropy s("), modulation instability λ1(") and s Exnonentially

Figure 8: Sprecht et possible equilibria (m = 2) in a RRG with K = 3. In a RRG with K = 3. In a RRG with K = 3





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colation and kinetically-constrained models). An explicit analytical solution

are no-more essential for the collective behavior (Marinari, Monasson, Semerjian 2006)

 $\bullet$   $\rho$  = fraction of cooperators  $s(\rho) = log(number of NE|\rho)/N$ 

 $\mathsf{I}\rho$ )/N NE ∃  $\forall \rho \in [0,1]$ 

#### Regular random graphs: mi=3  $D_{\Omega}$  possible companies in the cooperators in spin-glass instability λ2(") are displayed.

! <sup>3</sup>

<sup>K</sup> <sup>−</sup> <sup>3</sup>

closes only at K  $\approx$ 

<sup>K</sup> and ρ ≈ 1 − <sup>√</sup>



*-4 -2 0 2 4*



Figure 10: Typical behavior of the density of cooperators and entropy of CE

non-zero solution, the Lyapunov exponent becomes smaller than 1 exactly at the transition point). Increasing K the transition point moves to larger

values of !, i.e. ρc(K) decreases. Now, it is not clear numerically that the

ad an the limit K and some limit K expansion we see that we se

**Exponentially many NE's** as a function of K for m = 3.

20

Figure 12: Diagram entropy vs. density (for m = 3) in a RRG with K = 4.

 $\bullet$  NE  $\sharp$   $\forall \rho < \rho_c$ 

NE are non-local and fragile µ!,<sup>3</sup> ≈ with the numerical simulations). Computing the large  $\alpha$ stability condition, inserting the expression for µ and keeping only the first



## Heterogeneous random graphs

*x = 0.025*

FIG. 3 (Left) Density *typ* of collaborators in a typical equilibrium in Erd¨os-R´enyi graphs as function of

*0 10 20 30 40 50 60 70 k*



*0,5 0,6 0,7 0,8 0,9* likely than spokes *C(k) x = 0.1* Scale free: hubs collaborate more

*0,5*

*0,4*

#### Erdös-Rényi: E[k]=4

### $X_i = xk_i$



# Assortative networks are more conducive to collaboration



Scale free network  $P(k)$ ~ $k^{-2.5}$  $X_i = xk_i, x=0.1$ 

# Conclusions: Theory

Collaboration in repeated prisoners dilemma as a graph theoretical problem:  $\blacksquare$ 

- 1- make sure enough neighbors collaborate
- 2- not credible to monitor more neighbors
- 3- checks should be reciprocal
- If incentives to defect (x)  $\blacksquare$ 
	- is small then cooperation is easy  $\blacksquare$
	- **x** is large

i) collaboration requires critical mass ii) Nash equilibria are fragile

- iii) effect of defection are non-local
- Topology: Collaboration is easier on  $\blacksquare$ i) trees

ii) densely connected graphs

Collaboration is harder on networks which can be disconnected (e.g. quasi 1d graphs)

# Conclusion: Empirical evidence

- Individuals condition collaboration on that of others  $\blacksquare$ (Fishbacher, Gachter, Fehr 2001)
- Weak and strong ties (Granovetter, 1973)  $\mathbf{x}$ Individuals do not condition collaboration to all contacts, not even to all those who collaborate, only to a subset of them
- Critical mass theory of collective action (Oliver, Marwell 1993) If the cost of collaboration is large enough, collaborative equilibria only arise if a finite fraction of agents participate
- Collaboration easier in dense networks (Kirchkamp, Nagel 2007; ×. Cassar 2007)
- More connected agents are more likely to collaborate (Cassar 2007)
- Collaboration is not contagious (Suri, Watts 2011)  $\blacksquare$ The more of my contacts are engaged in conditional collaboration with others, the less likely I am to find neighbors with whom to collaborate conditionally