

Cooperation Networks

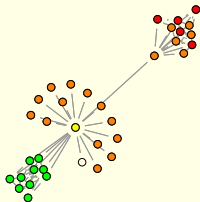
Endogeneity & Complexity

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Research initiated at the SFI Computational Grad Workshop, 2004*

ICCS, Boston, June 2006



Agenda

1. The problem of endogeneity;
2. A simple modification: properties and results;
3. Endogenous networks and complexity;
4. Areas for future interaction/work.

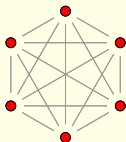
Cooperation: how does it work?

- ▶ Maynard Smith & Price: The logic of animal conflict (*Nature*, 1973);
- ▶ Axelrod & Hamilton: The evolution of cooperation (*Science*, 1981);
- ▶ Nowak & May: Evolutionary games and spatial chaos (*Nature*, 1992);
- ▶ Nowak & Sigmund: Evolution of indirect reciprocity by image scoring (*Nature*, 1998);
- ▶ Riolo, Cohen & Axelrod: Evolution of cooperation without reciprocity (*Nature*, 2001);
- ▶ Burtsev & Turchin: Evolution of cooperative strategies from first principles (*Nature*, 2006);

Analytical approaches

Uniform

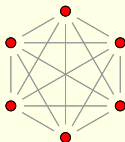
- ▶ 'Trembling towards equilibrium'
(best-response with mistake-making)
- ▶ Risk-dominant eq.
- ▶ e.g. KMR (1993)



Analytical approaches

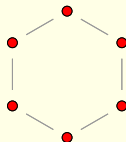
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Circle, Line, Grid

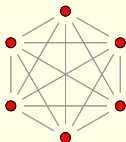
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Analytical approaches

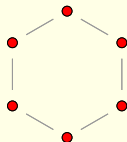
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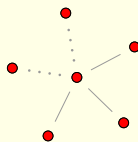
Circle, Line, Grid

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Dynamic

- ▶ Best-response graph-formation
- ▶ Inefficient and non-risk-dominant eq. possible
- ▶ e.g. Jackson & Watts (2002)



Limitations of Analytic Framework

- ▶ Strategies other than the Best-response (utility maximizing) hard to model analytically;
- ▶ Non-uniform (and non-regular) interaction spaces very challenging;
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- ▶ **boundedly rational behaviour + strategic network formation + dynamic interaction space = limits of analysis!**
- ▶ **But, computational, agent-based approaches well suited!**

Computational Approaches

Many models of boundedly rational play, but endogeneity of interaction?

- ▶ Ising models (incl. social-influence on small-world, random);
- ▶ Computation on a grid (e.g. 2D);
- ▶ Diffusion of technologies (again, structure-defined);
- ▶ IPD/CR (choice-refusal), endogenous, but network not strategic;

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Desirable Computational Model Qualities:

- ▶ 'Simple' set-up – relationship to previous literature;
- ▶ Truly *endogenous* (strategy-based, rather than observer based) interaction-space dynamics;
- ▶ Equilibria? Dynamics? Complexity?

Model Overview

Game

- ▶ Reward for cooperative, but risky play (modified IPD);
- ▶ Signal (#) play: fore-go payoff, establish link;
- ▶ Re-establishment each interaction.

	$\#_w$	C	D	$\#_s$
$\#_w$	(0,0)	...		(0,0)
C	.	(3,3)	(0,5)	.
D	:	(5,0)	(1,1)	:
$\#_s$	(0,0)	...		(0,0)

Model Overview

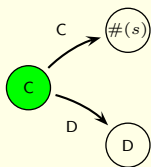
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Agents

Finite State Automata (FSA),
GA updating



Control: length of interactions τ ;
number of ints/prd m

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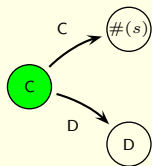
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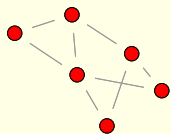
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Mixing

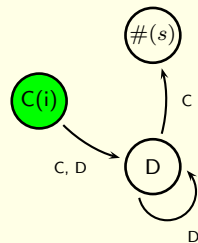
Uniform initially;
then endogenous
 \sim like, dislike,
untried



Control: impact of 'like' η

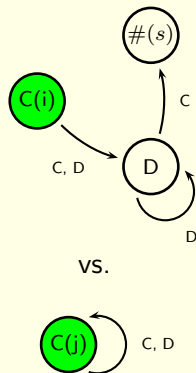
Example Interaction

1. Agent i addressed;



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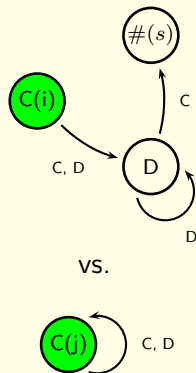
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Example Interaction

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3. IPD: interaction stops if $\#$ played, or τ iterations reached;

Iteration	s_i	s_j	π_i	π_j
1	C	C	3	3
2	D	C	5	0
3	$\#(s)$	C	0	0
$\sum \pi_x$			8	3

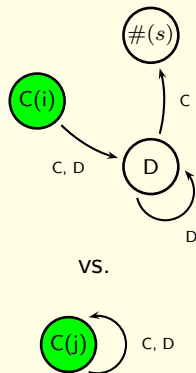


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$\sum \pi_x$			8	3

4. $\sum \pi_x$ added to period payoffs;
5. Update interaction structure (here, $i \leftrightarrow j$).



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- ▶ Due to enhanced *agency*, network that arise are due to strategic play of individuals (not externally applied);

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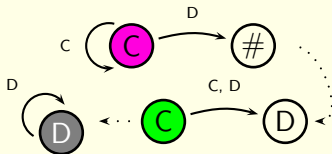
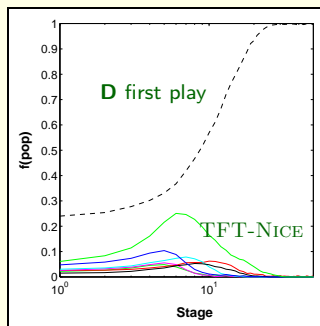
- ▶ Due to enhanced *agency*, networks that arise are due to strategic play of individuals (not externally applied);
- ▶ Networks can be 'good' *and* 'bad' for agents (not just arbitrary decision of inquirer);
- ▶ Capacity to deal with multiple networks at same time (not single component or list);
- ▶ FSA allows for large strategic space (e.g. for $\tau = 3 \longrightarrow 34$ *distinct* strategies);
- ▶ FSA encoding provides facile method of *learning* and *innovation/mistake-making* for agents;

Without network formation, $\eta = 0$

- ▶ Is system still similar to standard IPD set-up?
- ▶ ... does the playing of # affect things?

Without network formation, $\eta = 0$

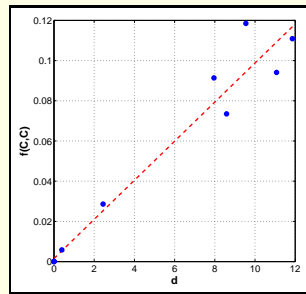
- ▶ Is system still similar to standard IPD set-up?
- ▶ ... does the playing of # affect things?
- ▶ Can show analytically that **D** play inevitable;
- ▶ Seen computationally (20 trials; 100 agents; $m = 20$).



$\eta > 0$: Network formation & Cooperation

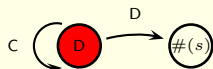
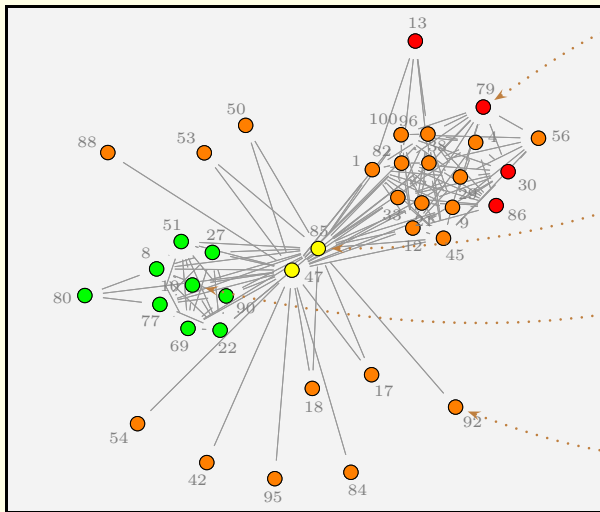
'Frequency' & 'Choice'

- ▶ Cooperation and average degree strongly related;
- ▶ Frequency of interaction AND 'impact' of edges necessary for sustainable cooperation-networks.

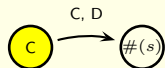


$m \setminus \eta$	\bar{d}			$f(C, C)$		
	0.80	0.90	0.95	0.80	0.90	0.95
10	0.000	0.000	0.000	0.000	0.000	0.000
14	0.004	0.001	0.391	0.000	0.000	0.006
18	2.441	11.859	8.587	0.029	0.111	0.074
20	7.959	11.073	9.548	0.091	0.094	0.119

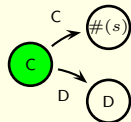
Network formation I: usual suspects



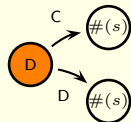
ID 79 'robust-D'



ID 85 'sucker'

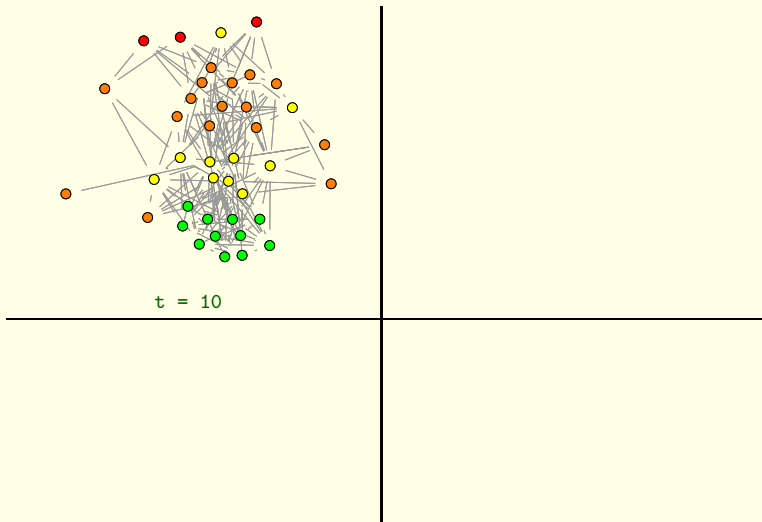


ID 10 'robust-C'

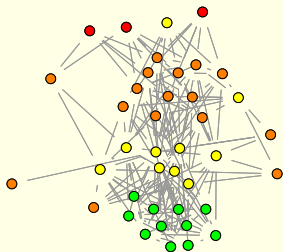


ID 92 'opportunist'

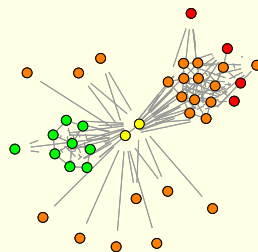
Network formation II: a character tour ...



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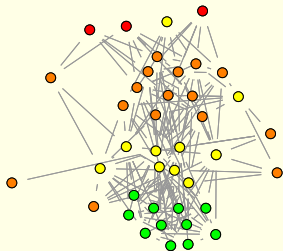


$t = 10$

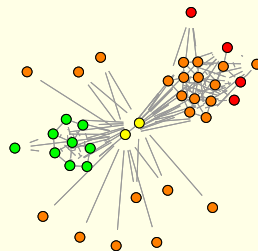


$t = 13$

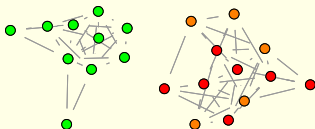
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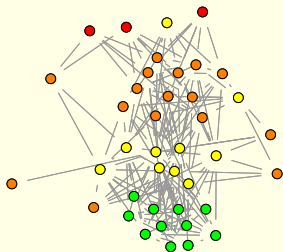
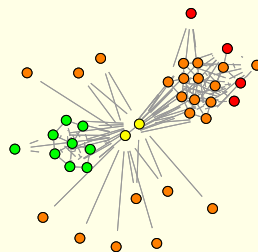
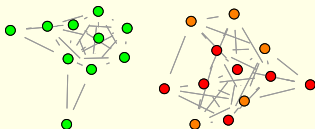
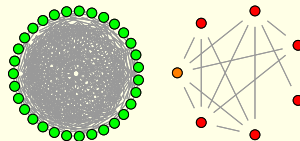


$t = 13$



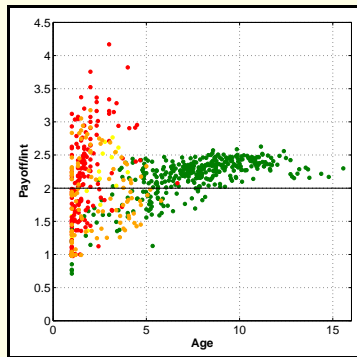
$t = 17$

Network formation II: a character tour ...

 $t = 10$  $t = 13$  $t = 17$  $t = 28$

Network Purity & Stability

- ▶ High payoffs in mixed networks visible;
- ▶ Assortative (preferential) mixing leads to long-gevity;
- ▶ All-D payoff cut-point.



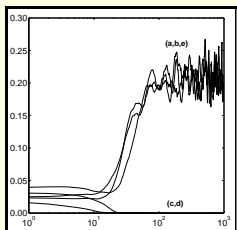
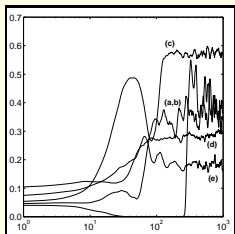
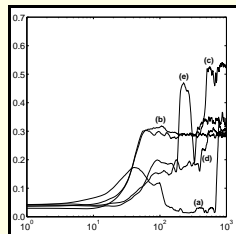
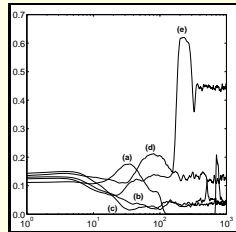
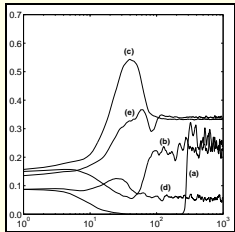
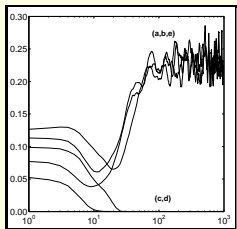
Network payoffs vs. mean age by dominant (> 50%) type

What about dynamics?

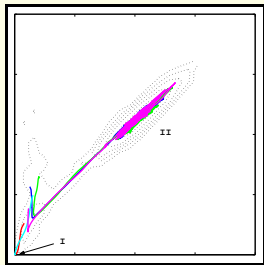
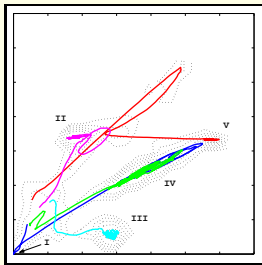
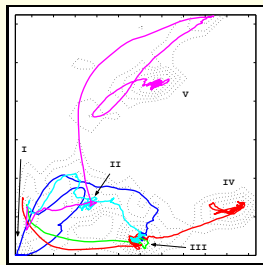
What is a state?

- ▶ State description is enormous (network + automata);
- ▶ Alternative, capture descriptive statistics that give aggregate description of state:
 - ▶ **Strategy measure** fraction of mutual cooperative plays, out of all plays ($f(C, C)$);
 - ▶ **Network measure** average agent link sponsorships ($\langle d \rangle$);

From simplicity to complexity...

 $\tau = 2$  $\tau = 3$  $\tau = 4$  $\langle d \rangle$  $f(C, C)$

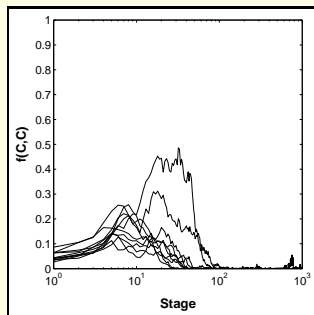
In 'Network:Strategy' space

 $\tau = 2$  $\tau = 3$  $\tau = 4$ 

Sources of complexity?

- ▶ Interactions?
- ▶ Strategies?
- ▶ Network dynamics?

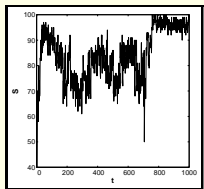
$\tau = 4$, network off



Endogeneity & self-organized criticality

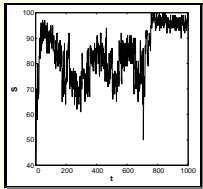
- ▶ Network changes could be source of complexity;
- ▶ Does the network show scaling over time and space?
- ▶ What is an **event**?
 - ▶ **Space** Frequency distribution of changes in network size (nodes, principle component);
 - ▶ **Time** Power spectra of size changes over time;
- ▶ Power-law scaling would indicate system criticality.

Analysing self-organized criticality on networks



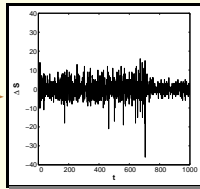
Network size series

Analysing self-organized criticality on networks



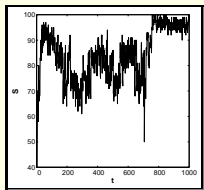
Network size series

$$x_t - x_{t-1}$$



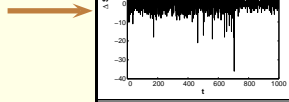
'Event' series: change in size

Analysing self-organized criticality on networks



Network size series

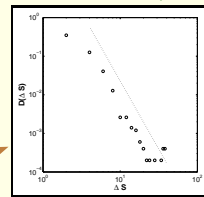
$$x_t - x_{t-1}$$



'Event' series: change in size

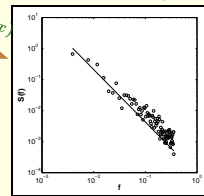
$$F(x)$$

Fractal scaling (space)

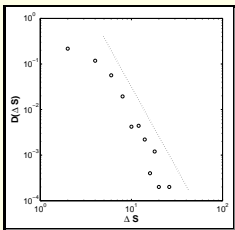
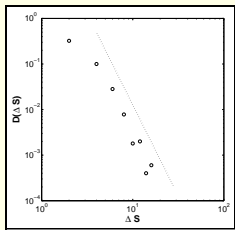
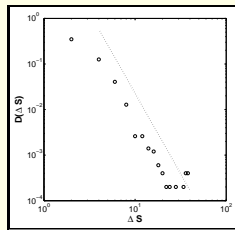
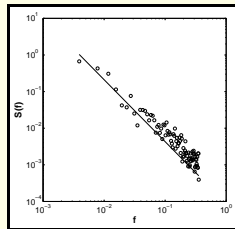
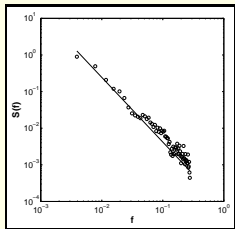
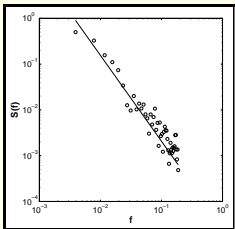


$$DFFT(x)$$

Power spectra (time)



Power-law scaling (again...)

 $\tau = 2$

 $\tau = 3$

 $\tau = 4$

 $D(\Delta S)$

 $S(f)$

What does it mean?

- ▶ White noise would give slope of 0, here, slope: -1.8 ± 0.1 ;
- ▶ The system (analysed in these measures) displays critical behaviour (i.e. at/near a phase change);
- ▶ Impact of events propagate through spatial and temporal dimensions – connectivity;

What does it mean?

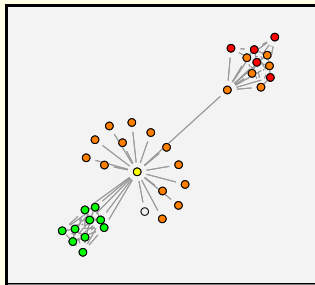
- ▶ White noise would give slope of 0, here, slope: -1.8 ± 0.1 ;
- ▶ The system (analysed in these measures) displays critical behaviour (i.e. at/near a phase change);
- ▶ Impact of events propagate through spatial and temporal dimensions – connectivity;
- ▶ Not surprising, although...
- ▶ Implies the SOC outcomes:
 - ▶ Tracking 'equilibrium' (resting) points becomes a *statistical* task (rather than by explicit prediction);
 - ▶ Seemingly small events can cause system-wide effects (although rarely) .. don't expect proportionality;
 - ▶ 'Simple' modification can upset canonical behaviour.

Future questions

- ▶ Coordination games on networks?
- ▶ Biological: self-replication with fitness?
- ▶ Economic: communication? reputation? signalling in network?
- ▶ Implications of SOC/complexity in these models – are we comfortable with disequilibrium? Long-run data on these effects?
- ▶ Social/government policy: dynamic control? What are the instruments? How costly are they? Where do they apply?

Thanks

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$\eta > 0$: Mean Population Behaviours

Establishing the Network

($m = 20, \eta = 0.8$)

- ▶ Periodic behaviours observed: 'sucker' types; 'opportunists'; cooperation network builders; and defection network builders;
- ▶ 'Shake-out' period as before, but cooperation network resilient;
- ▶ In network forming trials, cooperative network grows to encompass $\sim 60\%$ of population

