

Competitive Contagion Scoring Review

- Let P be the population distribution of seed choices on graph G
- For every seed set s that appears with non-zero probability in P , we will compute its *expected payoff with respect to P* :
 - average of $\text{pay}(s,s')$ over many trials and many draws of s' from P
 - enough draws/trials to distinguish/rank expected payoffs accurately
- We will then rank the s that appear in P by their expected payoffs
- If you played s on G , you will receive a number of points equal to the *number of other players* you *strictly beat* in expected payoff
- Example: Suppose s_1 , s_2 and s_3 appear in P , and have expected payoffs and population counts as follows:
 - s_1 : payoff 0.57, count 11; s_2 : payoff 0.48, count 71; s_3 : payoff 0.31, count 18
 - if you play s_1 , your score is $71+18=89$; if s_2 , your score is 18; if s_3 , your score is 0
- If everyone plays the same thing, nobody receives any points
- You must submit seeds for *all* graphs in order to receive any credit
- Your overall score/grade for the assignment is the sum of your scores over all graphs, which will then be curved
- In general, there is no right/best choice for seeds: depends on P !

Questions Worth Pondering

- What does it mean for the population distribution P to be an equilibrium?
- If P is an equilibrium what can we say about different players' payoffs?
- If P is an equilibrium and G is connected, what can we say about payoffs?
- What if G is not connected?

Networked Games: Coloring, Consensus and Voting

Prof. Michael Kearns

Networked Life

NETS 112

Fall 2014

Experimental Agenda

- Human-subject experiments at the intersection of CS, economics, sociology, “network science”
- Subjects simultaneously participate in groups of ~ 36 people
- Subjects sit at networked workstations
- Each subject controls some simple property of a single vertex in some underlying network
- Subjects have only *local* views of the activity: state of their own and neighboring vertices
- Subjects have (real) financial incentive to solve their “piece” of a collective (global) task
- Simple example: graph coloring (social differentiation)
 - choose a color for your vertex from fixed set
 - paid iff your color differs from all neighbors when time expires
 - max welfare solutions = proper colorings
- Across many experiments, have deliberately varied *network structure* and *task/game*
 - networks: inspired by models from network science (small worlds, preferential attachment, etc.)
 - tasks: chosen for diversity (cooperative vs. competitive) and (centralized) computational difficulty
- Goals:
 - structure/tasks → performance/behavior
 - individual & collective modeling → prediction
 - computational and equilibrium theories

Experiments to Date

- **Graph Coloring**
 - **player controls:** color of vertex; number of choices = chromatic number **payoffs:** \$2 if different color from all neighbors, else 0 **max welfare states:** optimal colorings **centralized computation:** hard even if approximations are allowed
- **Consensus**
 - **player controls:** color of vertex from 9 choices **payoffs:** \$2 if same color as all neighbors, else 0 **max welfare states:** global consensus of color **centralized computation:** trivial
- **Independent Set**
 - **player controls:** decision to be a “King” or a “Pawn”; variant with King side payments allowed **payoffs:** \$1/minute for Solo King; \$0.50/minute for Pawn; 0 for Conflicted King; continuous accumulation **max welfare states:** maximum independent sets **centralized computation:** hard even if approximations are allowed
- **Exchange Economy**
 - **player controls:** limit orders offering to exchange goods **payoffs:** proportional to the amount of the other good obtained **max welfare states:** market clearing equilibrium **centralized computation:** at the limit of tractability (LP used as a subroutine)
- **Biased Voting**
 - **player controls:** choice of one of two colors **payoffs:** only under global agreement; different players prefer different colors **max welfare states:** all red and all blue **centralized computation:** trivial
- **Networked Bargaining**
 - **player controls:** offers on each edge to split a cash amount; may have hidden deal limits and “transaction costs” **payoffs:** on each edge, a bargaining game --- payoffs only if agreement **max welfare states:** all deals/edges closed **centralized computation:** nontrivial, possibly difficult
- **Voting with Network Formation**
 - **player controls:** edge purchases and choice of one of two colors **payoffs:** only under global agreement; different players prefer different colors **max welfare states:** ??? **centralized computation:** ???

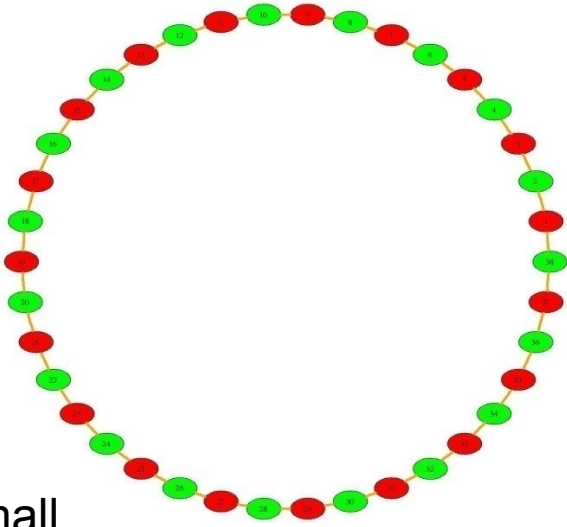
Coloring and Consensus

game progress: 80%
game status: ColoringGame in progress
elapsed time:
your current payoff: **\$2.00**
(payoff is \$2.00 if your color is DIFFERENT from all your neighbours, otherwise \$0.00)

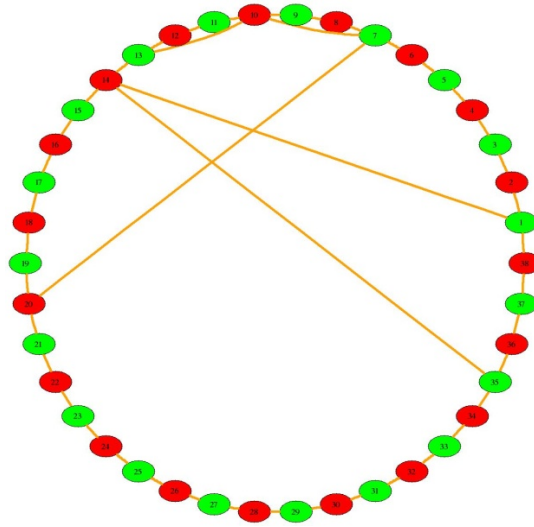
your color: yellow red green

“first neighborhood” view

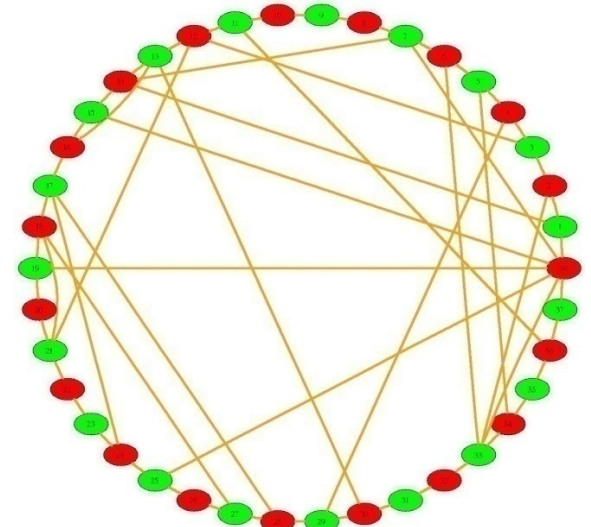
[\[demo\]](#)



Simple Cycle

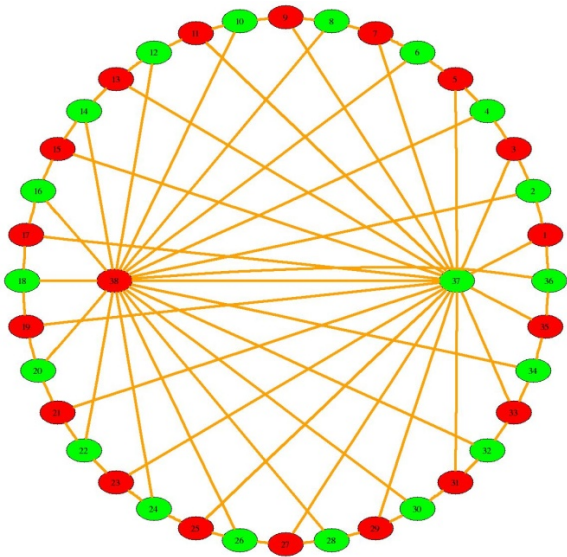


5-Chord Cycle

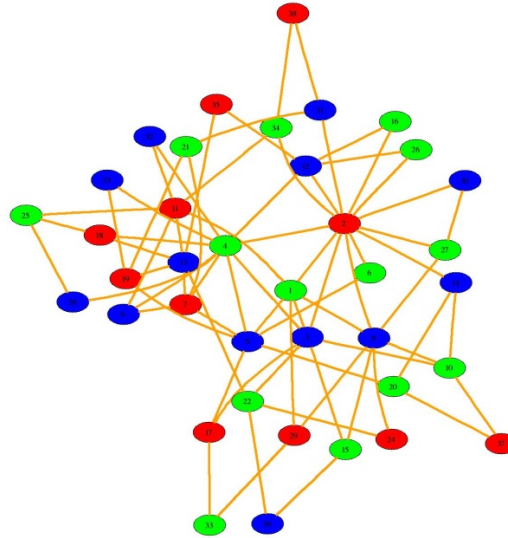


20-Chord Cycle

Small
Worlds
Family

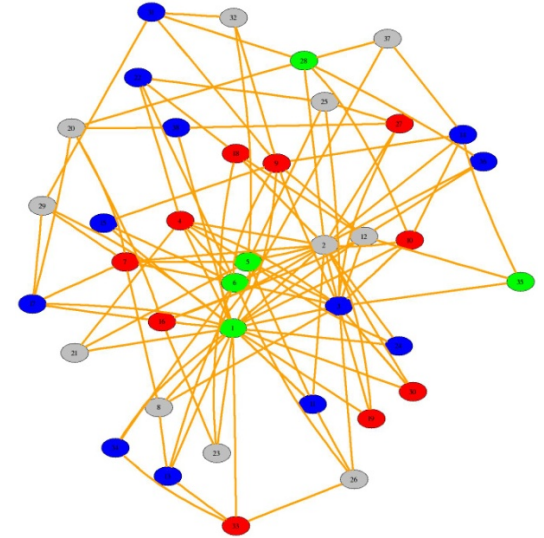


Leader Cycle



Preferential Attachment,

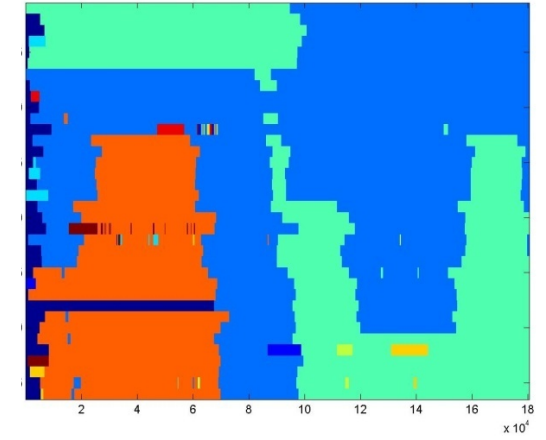
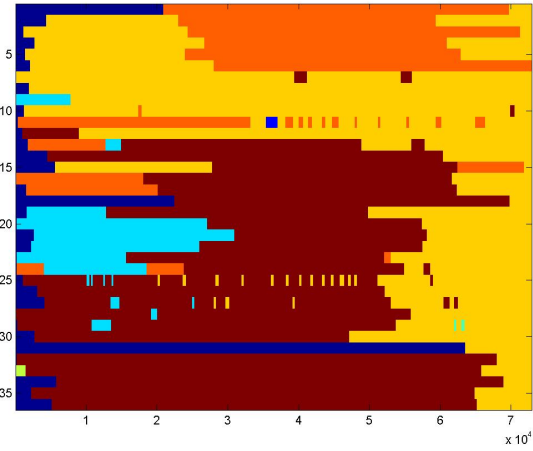
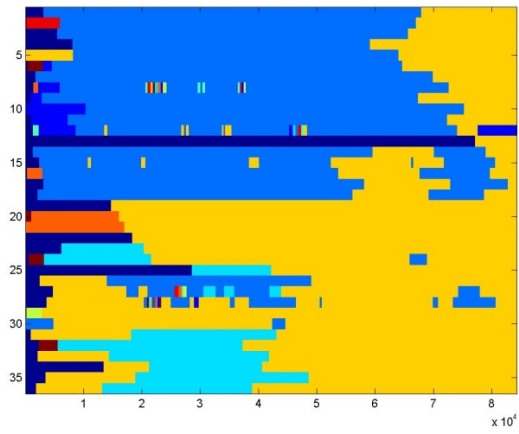
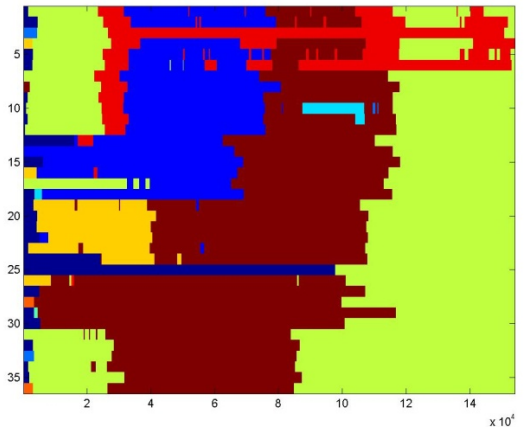
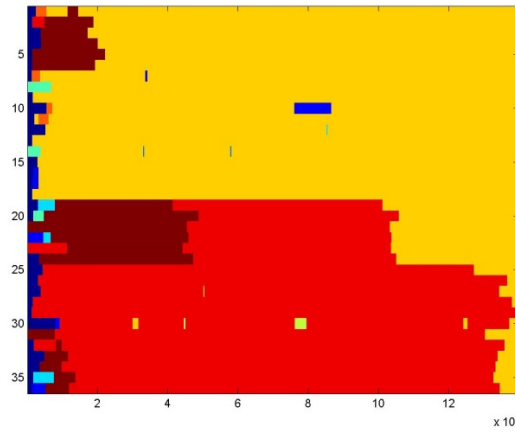
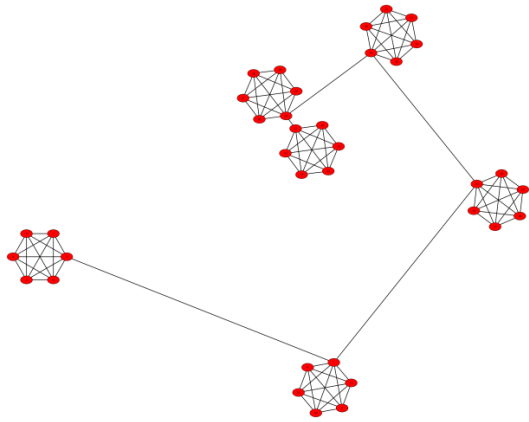
$$\nu = 2$$



Preferential Attachment,

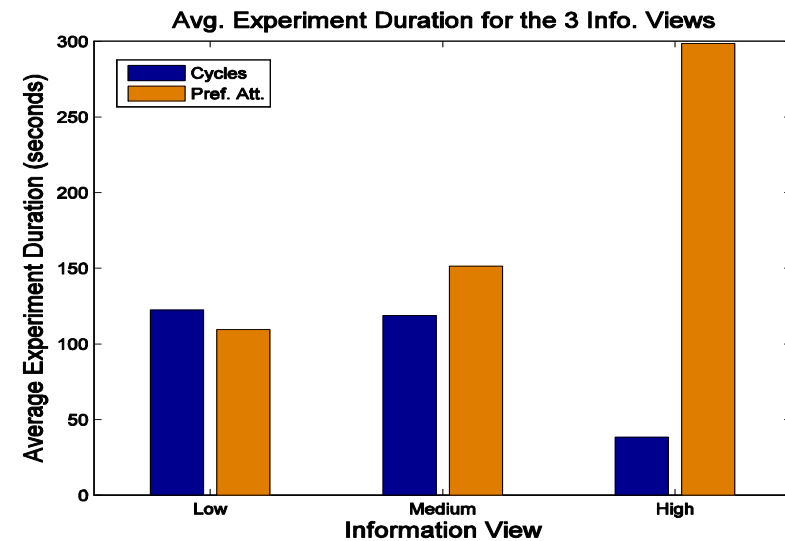
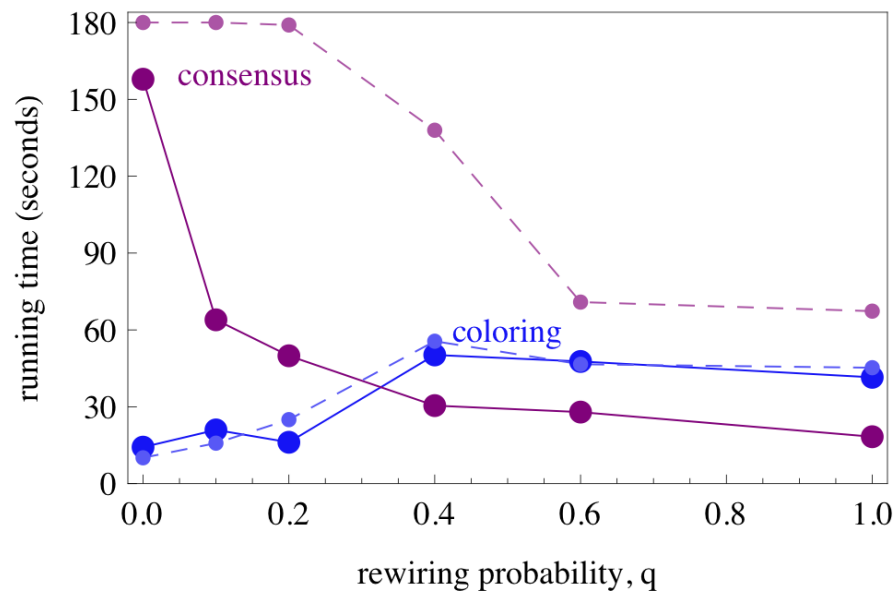
$$\nu = 3$$

Art by Consensus



Sample Findings

- Generally strong collective performance
 - nearly all problems globally solved in a couple minutes or less
- Systematic effects of structure on performance and behavior:
 - *rewiring harms* coloring performance in “clique chain” family
 - *rewiring helps* consensus performance in clique chain family
- Preferential attachment much harder than small worlds for coloring
 - natural heuristics can give reverse order of difficulty
- Providing more global views of activity:
 - helps coloring performance in small world family
 - harms coloring performance in preferential attachment
- Coloring problems solved more rapidly than consensus
 - easier to get people to disagree than agree



Biased Voting in Networks

Biased Voting in Networks

- Cosmetically similar to consensus, with a crucial strategic difference
- Deliberately introduce a tension between:
 - individual preferences
 - desire for collective unity
- Only two color choices; challenge comes from competing incentives
- If everyone converges to same color, everyone gets some payoff
- But different players have different preferences
 - each player has payoffs for their preferred and non-preferred color
 - e.g. \$1.50 red/\$0.50 blue vs. \$0.50 red/\$1.50 blue
 - can have symmetric and asymmetric payoffs
- High-level experimental design:
 - choice of network structures
 - arrangement of types (red/blue prefs) & strengths of incentives
 - most interesting to coordinate network structure and types

Behavioural Network Science Games

http://kearnsgroup.cis.upenn.edu/BNS/games.php?playername=Zak+Xavier

travel JCR Portal ABN AMRO - ...al Banking BofA Penn.Pay Login BlueCross ACHE.org PPM Members Gmail Adventuring Abroad...

Democratic Primary Games

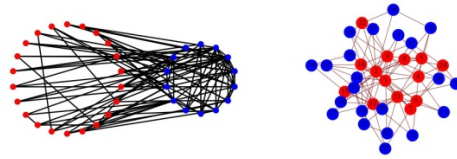
Zak Xavier

game progress: 67%
game status: Voter Game in progress
elapsed time:

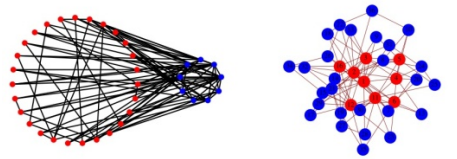
If unanimity is reached, your payoff will be
\$0.75 for red, \$1.25 for blue

your color: blue red

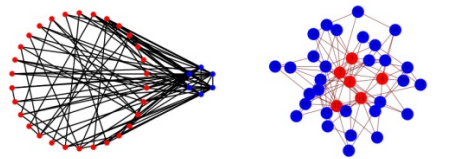
power22



power27



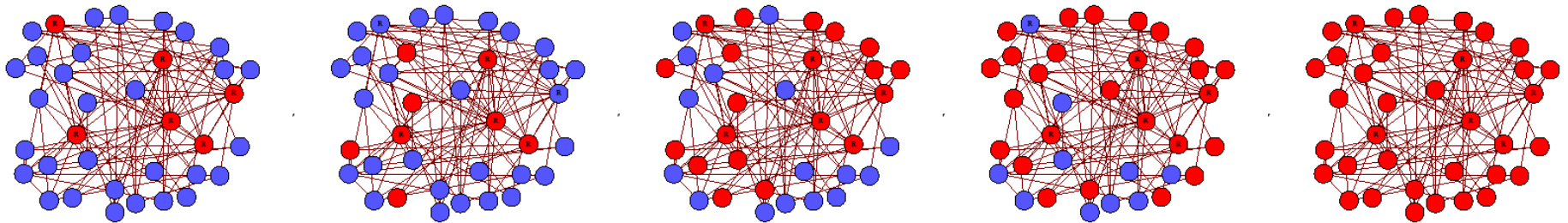
power30



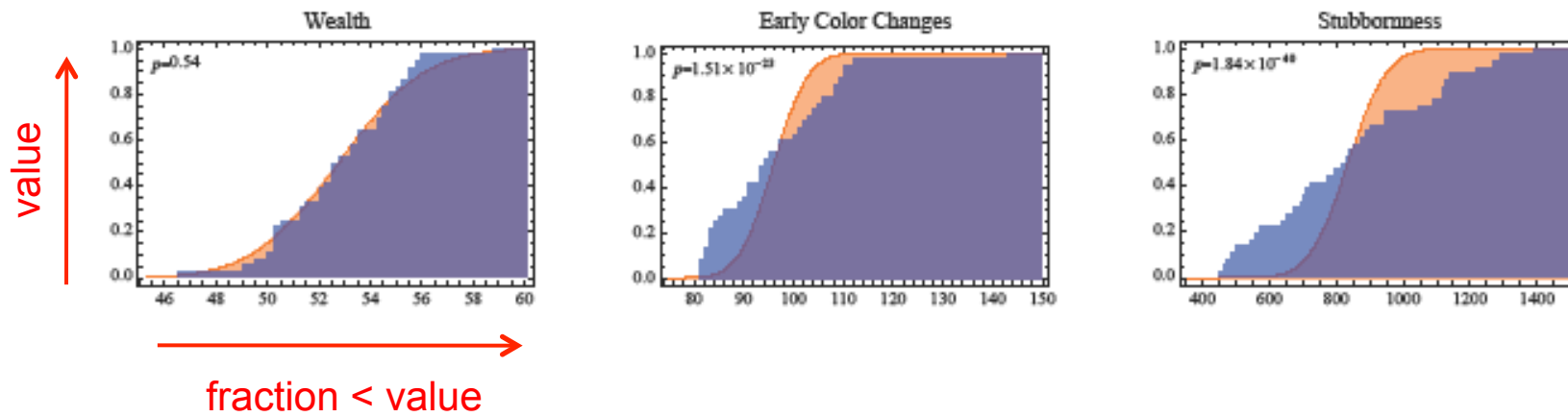
Minority Power: Preferential Attachment

Summary of Findings

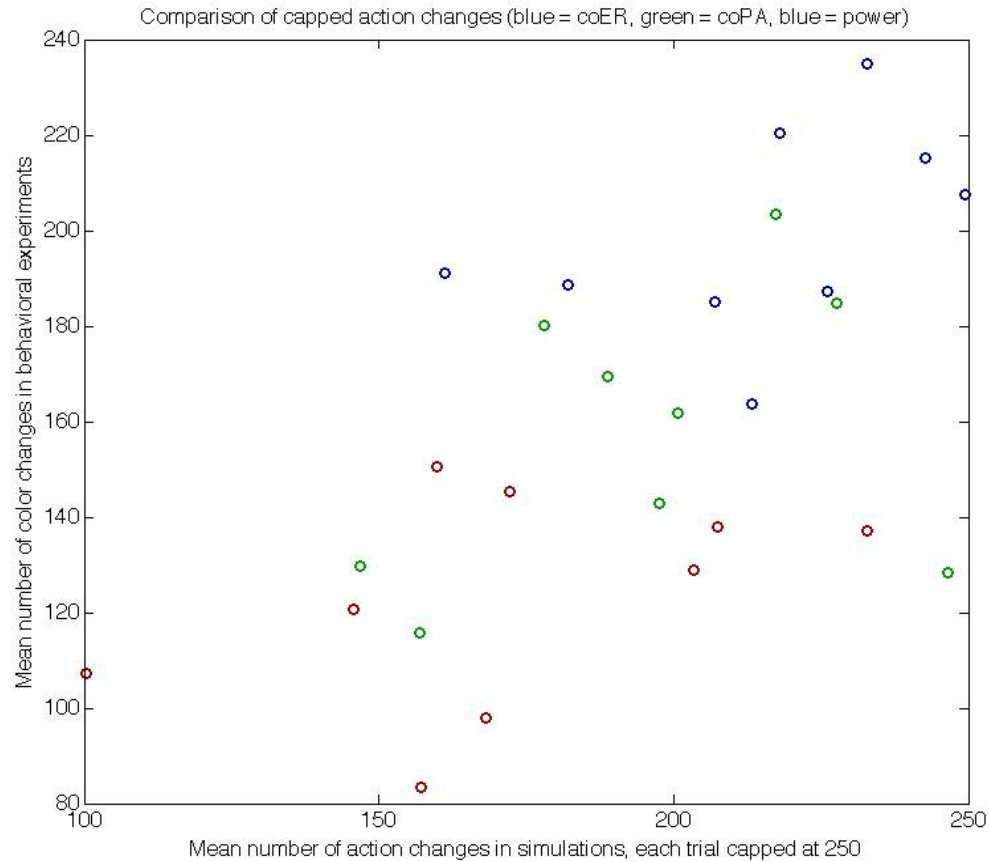
- 55/81 experiments reached global consensus in 1 minute allowed
 - mean of successful ~ 44s
- Effects of network structure:
 - Cohesion harder than Minority Power: 31/54 Cohesion, 24/27 Minority Power
 - all 24 successful Minority Powers converge to minority preference!
 - Cohesion P.A. (20/27) easier than Cohesion E-R
 - overall, P.A. easier than E-R (contrast w/coloring)
 - within Cohesion, increased inter-group communication helps
 - some notable exceptions...
- Effects of incentives:
 - asymmetric beats weak symmetric beats strong symmetric
 - the value of “extremists”



Effects of “Personality”



Behavioral Modeling



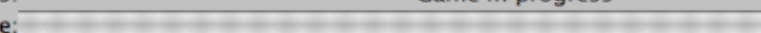
model: play color c with probability $\sim \text{payoff}(c) \times \text{fraction in neighborhood playing } c$

Lessons Learned, 2005-2011

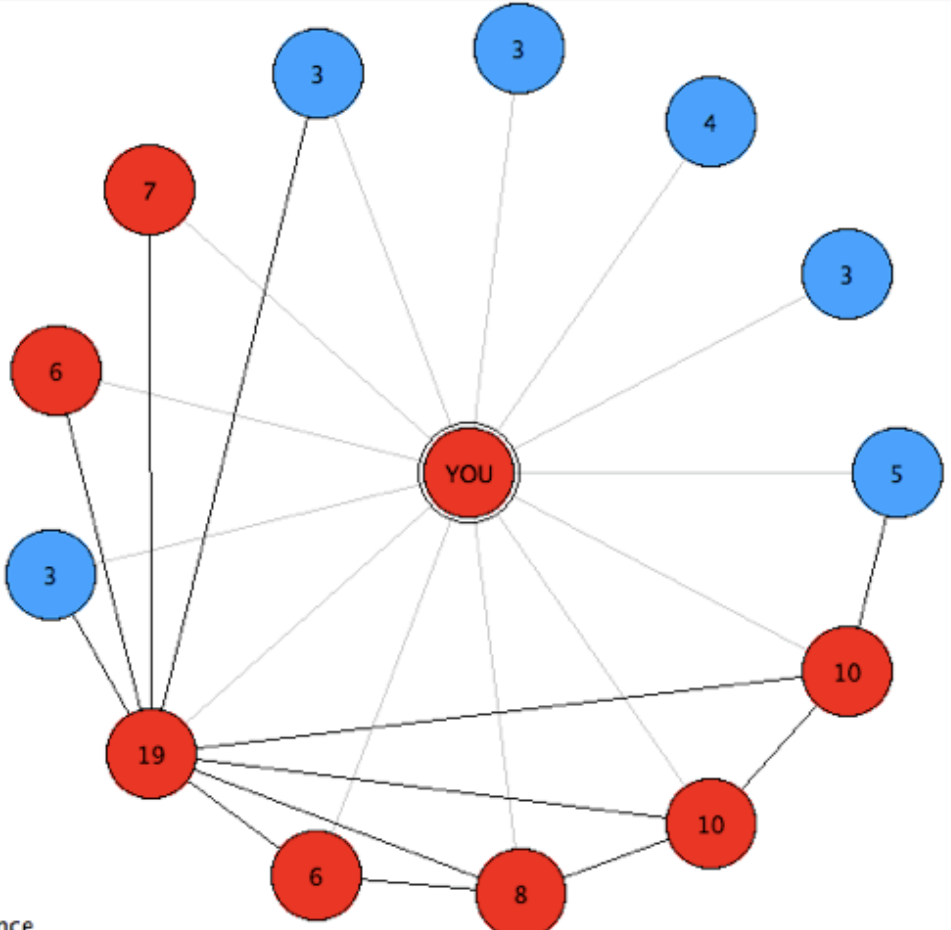
- At least for $n=36$, human subjects remarkably good
 - diverse set of collective tasks
 - diverse set of network topologies
 - efficiency $\sim 90\%$ across all tasks/topologies
- Network structure matters; interaction with **task**
 - contrast with emphasis on topology alone
- Importance of subject variability and style/personality
- Most recently: **endogenized** creation of the network
 - network formation games
 - challenging computationally (best response) and analytically

BNS Network Games

Behavioural Network Games Baffle Blythe

game status: Game in progress
elapsed time: 

payoffs for unanimity: \$3.00 \$2.00
edges cost \$0.03 You may purchase 35 edges.



degree

- 12 ○
- 11 ○
- 9 ○
- 8 ⊕
- 7 ○
- 6 ○
- 5 ⊕
- 4 ⊕
- 3 ⊕ ○
- 2 ○
- 1 ○
- 0 ○

2 3 4 5
distance

your color:

cycle time 200; work time 2, 2, 1

Edge Purchases: Strategic Tensions

- Buy edges or not?
- For information or influence?
- Early in the game or late?
- To high degree or low degree players?
- Nearby or far away?

Experimental Design

- Session A: 99 experiments
 - 63 “unseeded” with varying payoffs, imbalances, asymmetries
 - 36 seeded with Minority Power settings
- Session B: 72 experiments
 - mixture of unseeded and variety of seeded (cliques, torus)
- A: 47/99 solved (47%): 25/63 unseeded, MP 22/36
- B: 27/72 solved (38%)
- Session C: 72 experiments
 - **final networks** from “hard” settings in Session A
 - permitted 0 or 1 edge purchases per player
 - started with both **initial and final incentives** from Session A
- C: 25/72 (35%); All: 99/243 (41%)
- Subjects seem to build difficult networks!

[tape]