



POPULATION WIDE ATTITUDE DIFFUSION IN COMMUNITY STRUCTURED GRAPHS

Cognitive Science and Applications

Sandia National Laboratories

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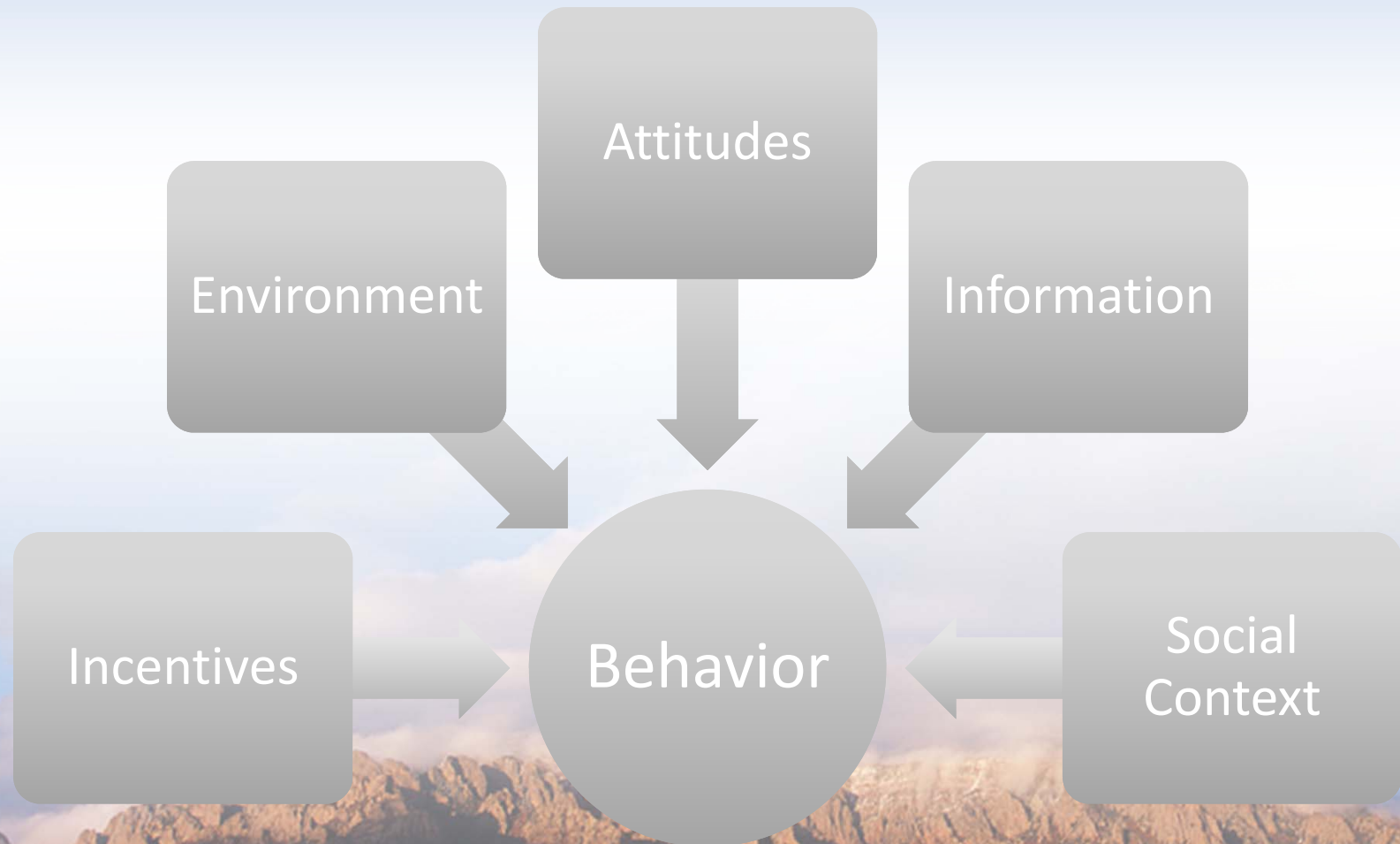


Research Objective

- How can we model population wide changes in behavior?
 - Linguistic behavior
 - Technology adoption (e.g., Smart Grid)
 - Violent actions etc.
 - Climate Change
- Building computational models that help in understanding population wide attitude change.



Factors in behavior change





Current Focus

- Attitude change.
- Leads to behavior change, but that is out of scope right now.
- How can we model the **diffusion** of **information** and subsequent **attitude change** within a population?



Attitude Definition

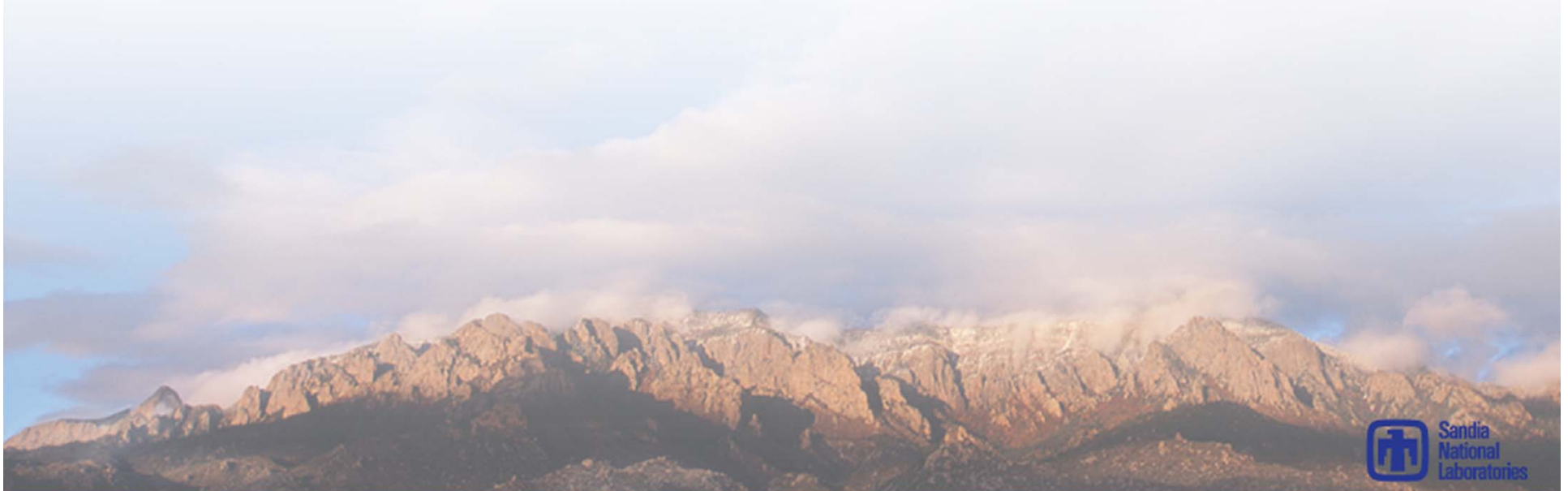
- “General, relatively enduring evaluation of an object” (Visser, 2003)
 - Object: Person, group, concept, issues
 - Valence: Positive or negative evaluation
 - Strength: Intensity of evaluation





Attitude properties

- **Continuous:** Attitudes can range from strongly negative to strongly positive.
- **Malleable:** Attitudes can change.
- **Linked:** Attitudes can influence each other





Factors in Attitude Change

- **Cognitive Effort:** Level of effort individuals place in evaluating arguments etc.
- **Social Network:** The impact of friends and family etc.



Drive for consistency

- Cognitive Consistency Theories: Originated in the 1950's from work on Gestalt Psychology.
- Key idea: Maintain consistency among “elements of thought”
- Example:
 - “positive attitude towards environmentalism” == “positive attitude towards recycling”
 - “positive attitude towards environmentalism” != “positive attitude towards littering”



Bi-directional Reasoning

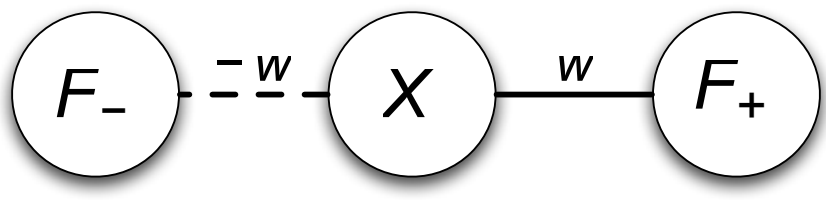
- Bidirectional reasoning
 - Evidence influences conclusion; conclusion influences belief in evidence.
 - “motivated reasoning” (Mooney, 2010)
 - Numerous studies from different domains:
 - Legal reasoning
 - Climate issues
 - Political affiliations

Parallel Constraint Satisfaction

- Nodes represent *concepts*: propositions, beliefs, information, traits, actions, goals.
- Links represent positive/negative influences between concepts: entailment, explanation, deduction, similarity, association.
- Nodes have a value (i.e. $-1 \dots +1$) and a valence (+/-)
- Attitude towards a concept is determined by the value and the valence of the concept:
 - Positive value = positive feeling.
 - Negative value = negative feeling.

Preliminary Experiment

- Two satisfying attitude structures:
 - $F_1=1, X=1, F_2=-1$
 - $F_1=-1, X=-1, F_2=1$





Three node

- X represents attitude to a particular object.
- F+ : Agents attitude towards positive information about X.
- F-: Agents attitude towards negative information about X.

Impression Formation (IMP) model



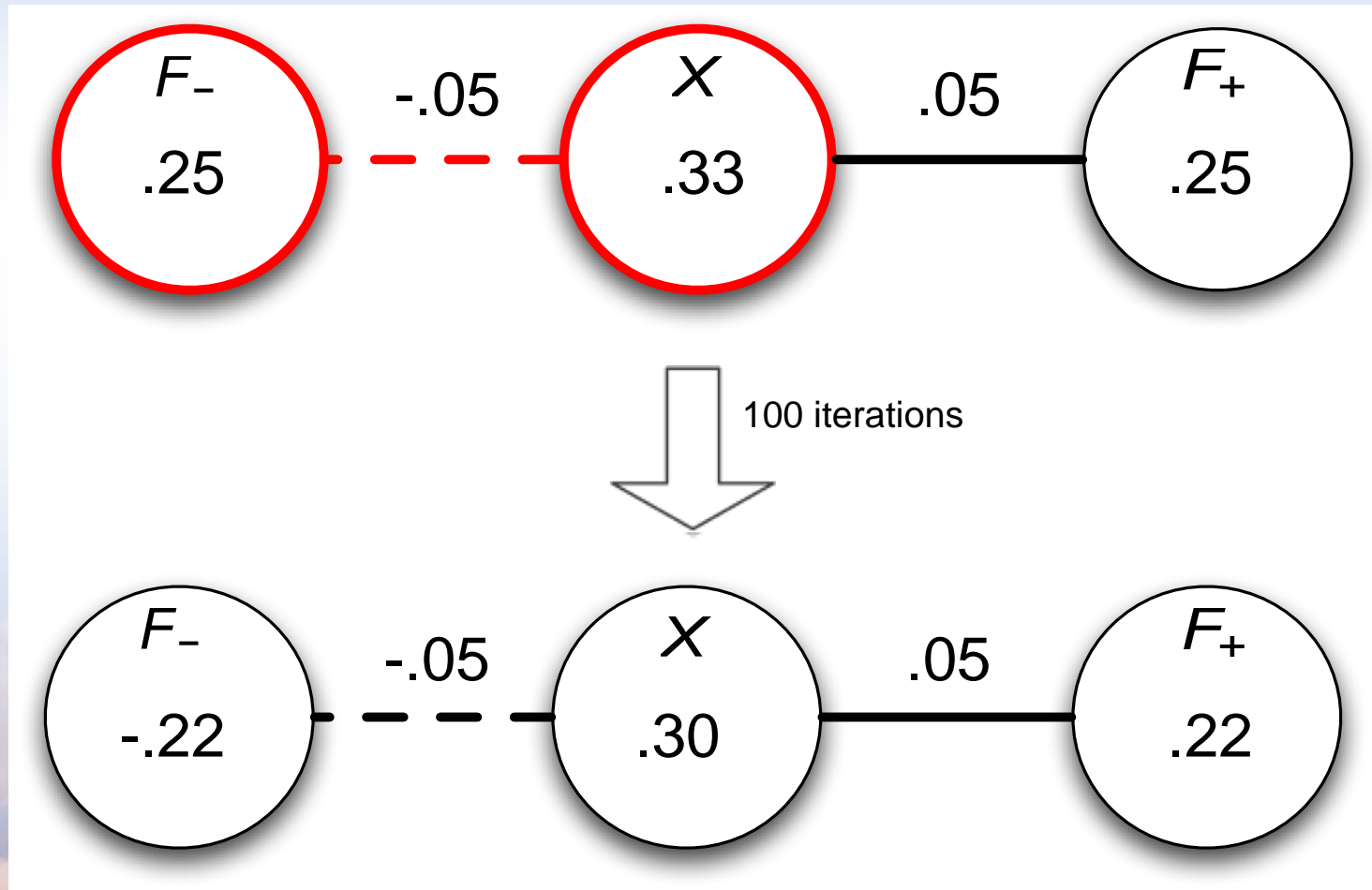
$$a_j(t+1) = a_j(t)(1-d) + \begin{cases} net_j(\max - a_j(t)) & \text{if } net_j > 0 \\ net_j(a_j(t) - \min) & \text{if } net_j \leq 0 \end{cases} \quad (1)$$

where:

$$net_j = \sum_i w_{ij} a_i(t) \quad (2)$$

From: (Kunda and Thagard, 1996)

Example Dynamics





Key aspects

- Cognitive Effort.
- Bidirectional reasoning.



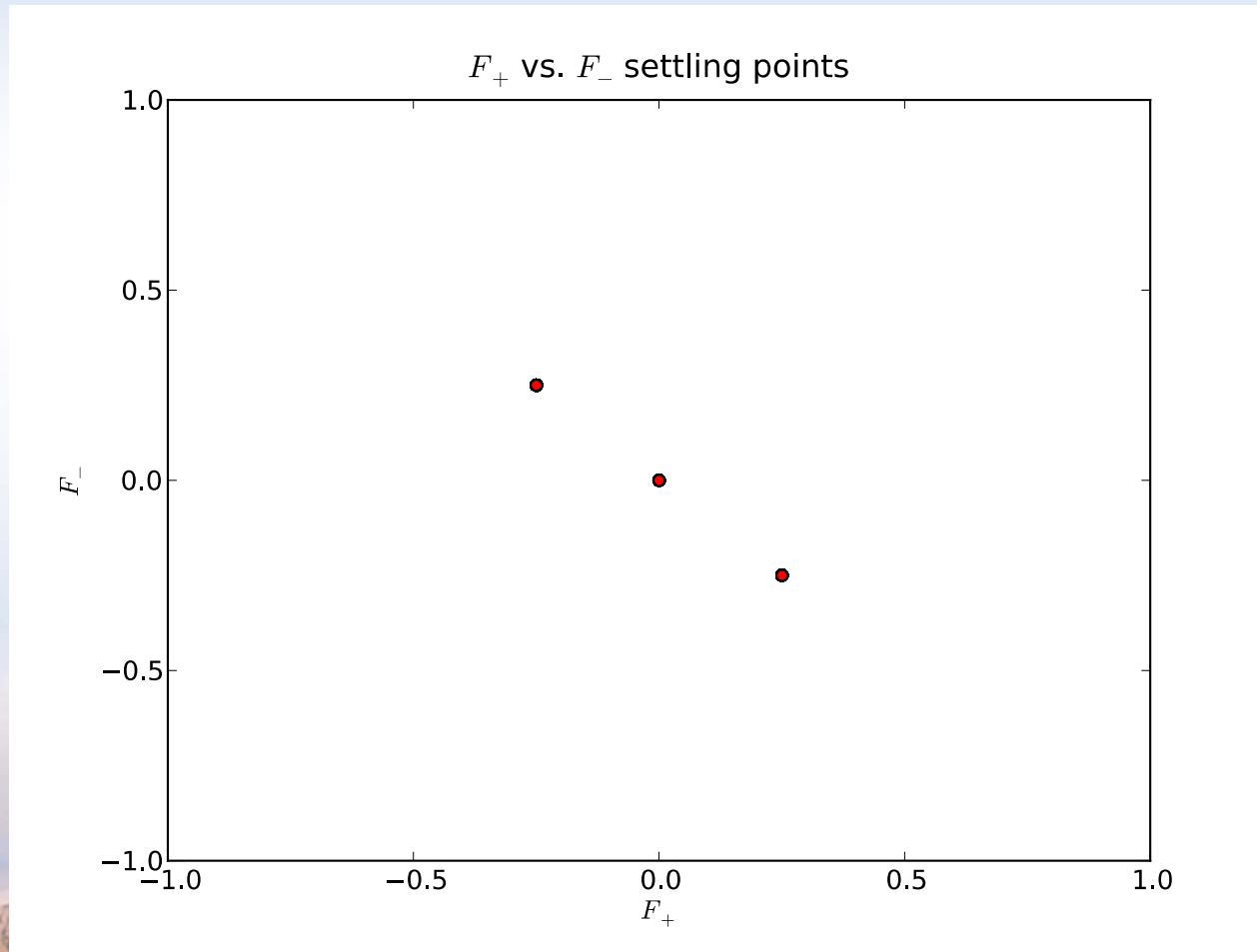


Cognitive Effort

- Cognitive effort captured by the *number* of iterations of the update Eq.
- Higher cognitive effort – more iterations.
- Lower cognitive effort – fewer iterations.

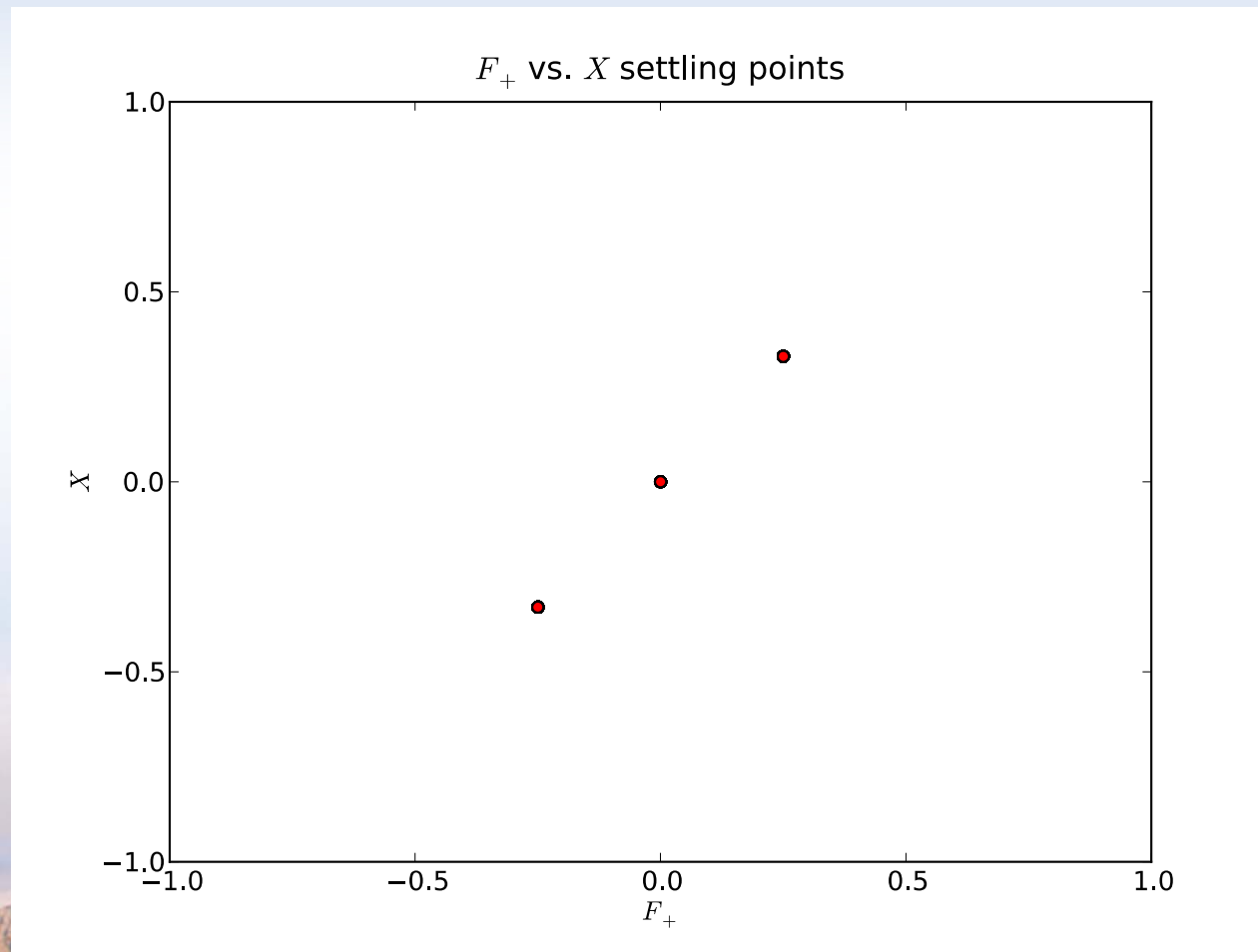
Settling Points

F+ vs. F-

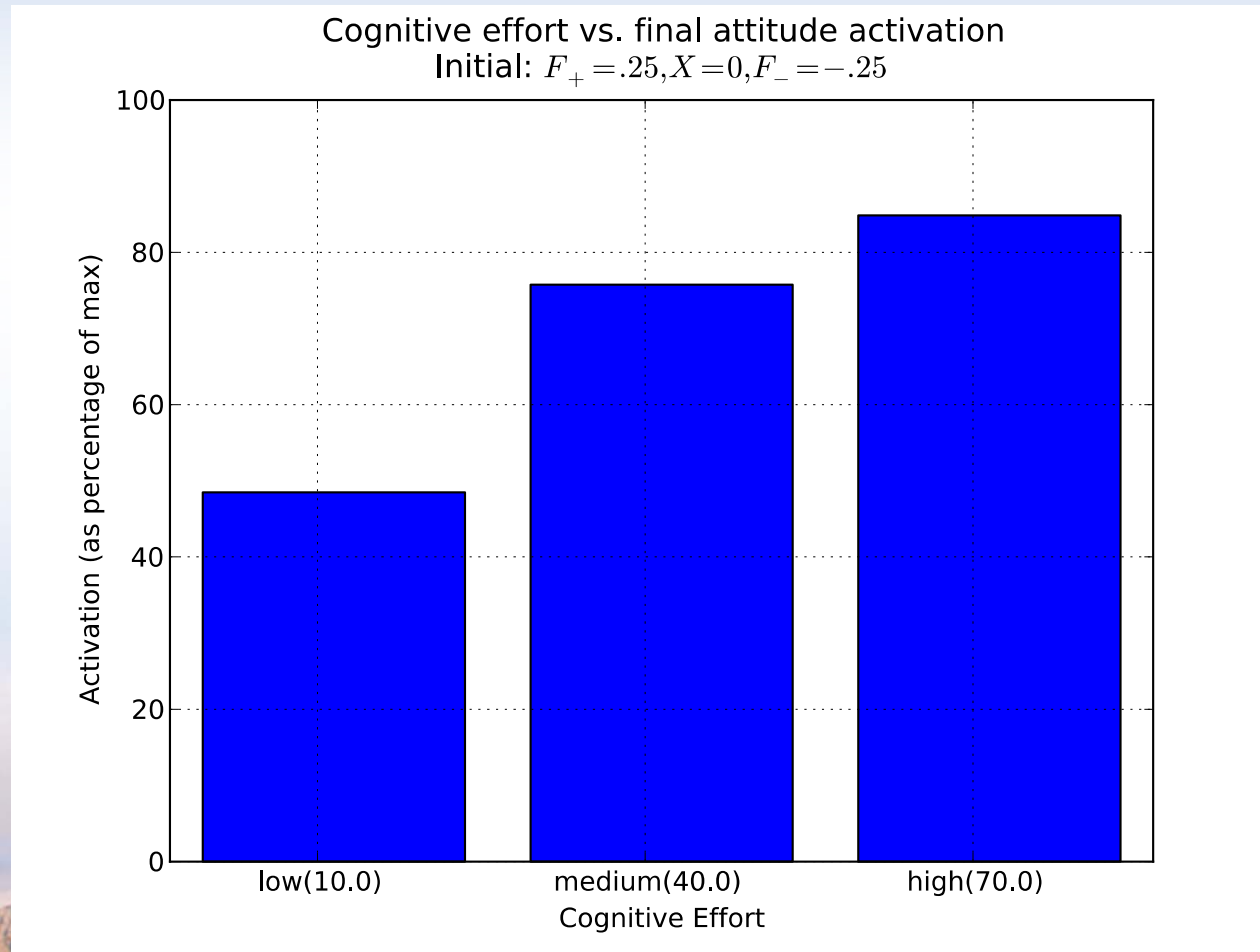


Settling Points

F_+ vs. X



Non-linear increase in attitude strength





Bi-directional change



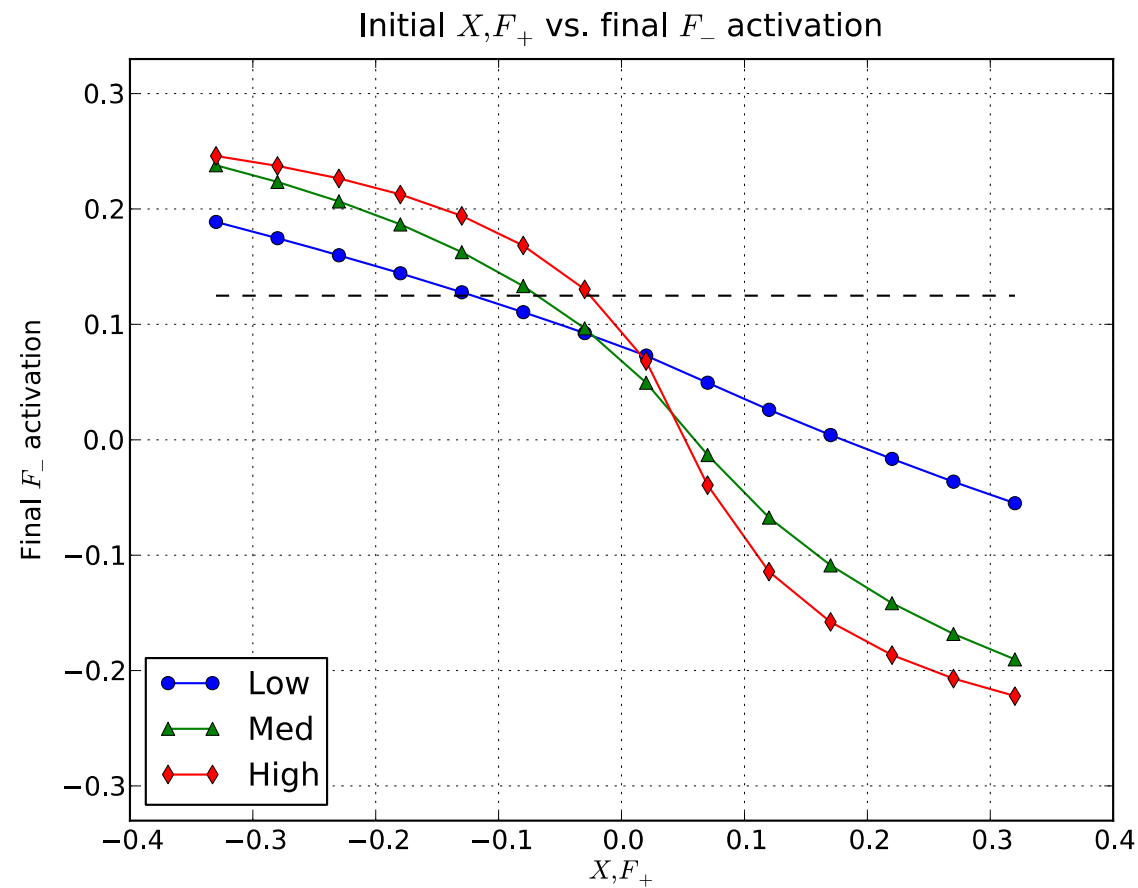
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- Eqs. (1) and (2) clearly show the impact of X on F^- and F^+ .
- What is the exact impact?
- Vary X, F^+ from $-.125$ to $.125$
- Set $F^- = .125$



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Bi-Directional Change



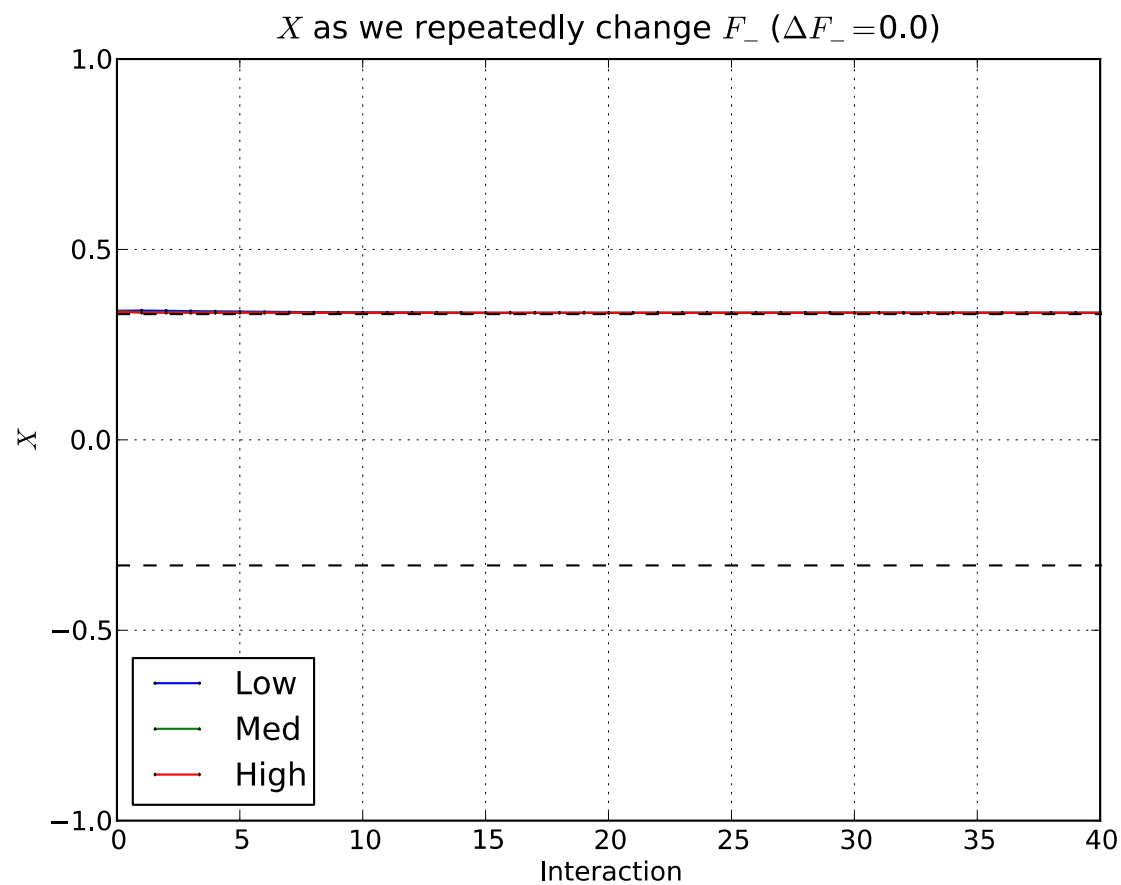


Attitude change

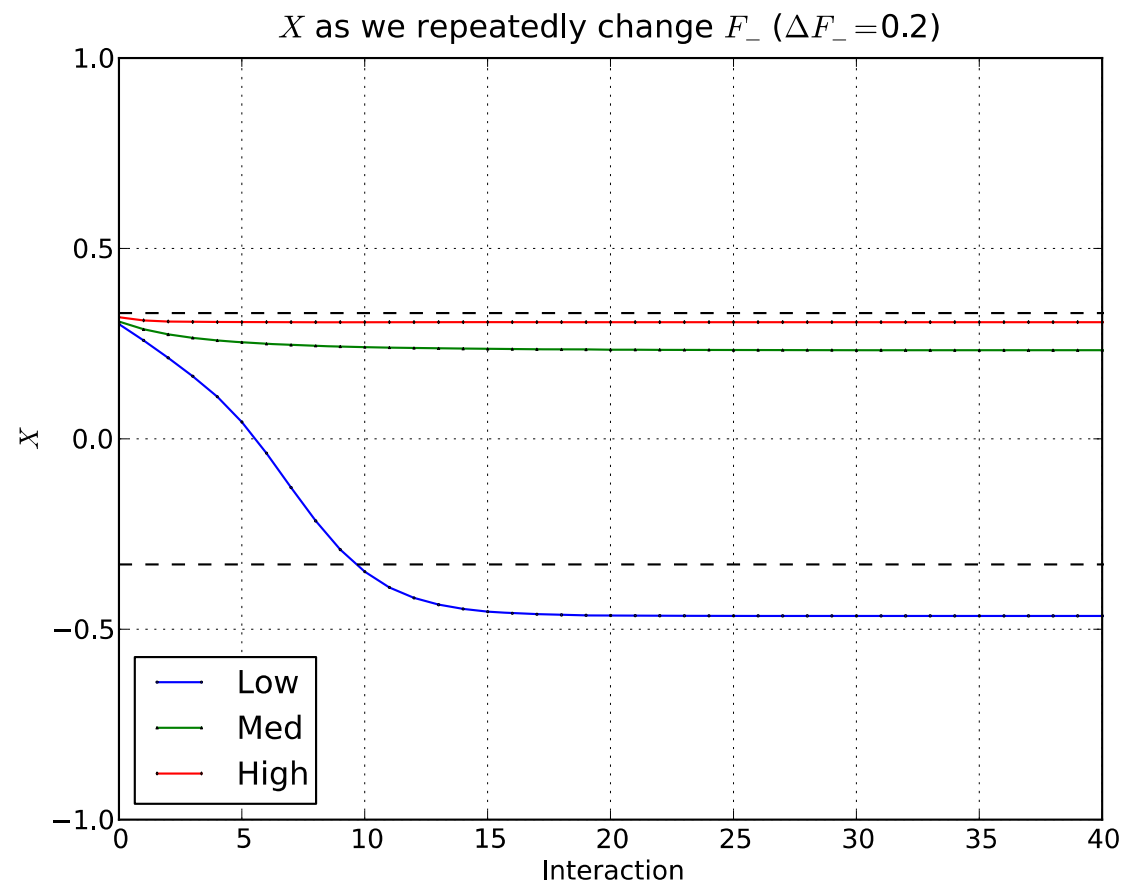
- How many *interaction* till a positive attitude turns into a negative attitude?
- Start with a positive attitude agents $<.25, .33, -.25>$
- Increase F- by fixed amount each turn.
- Run update eqs.
- How many turns till change?



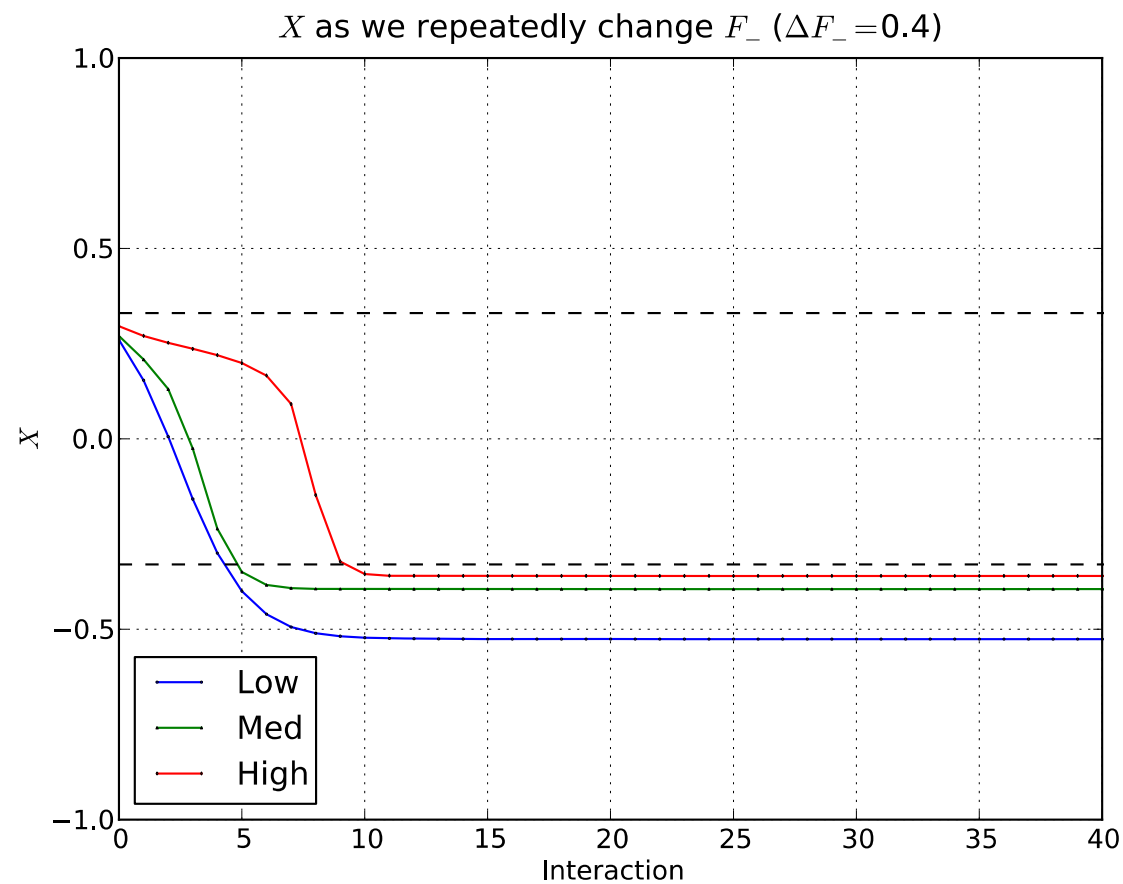
$\Delta F_- = 0.0$



\Delta F- = .2



\Delta F- = .4

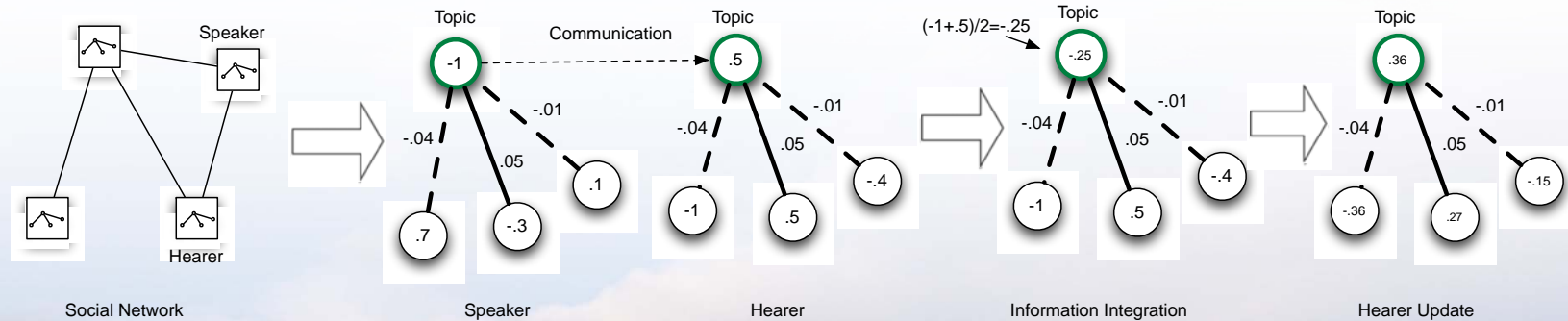




A Socio-Cognitive Model

- So we have attitudes and information within individual, how do we capture social influence?
- Use a social network to capture the topology of interactions between individuals.

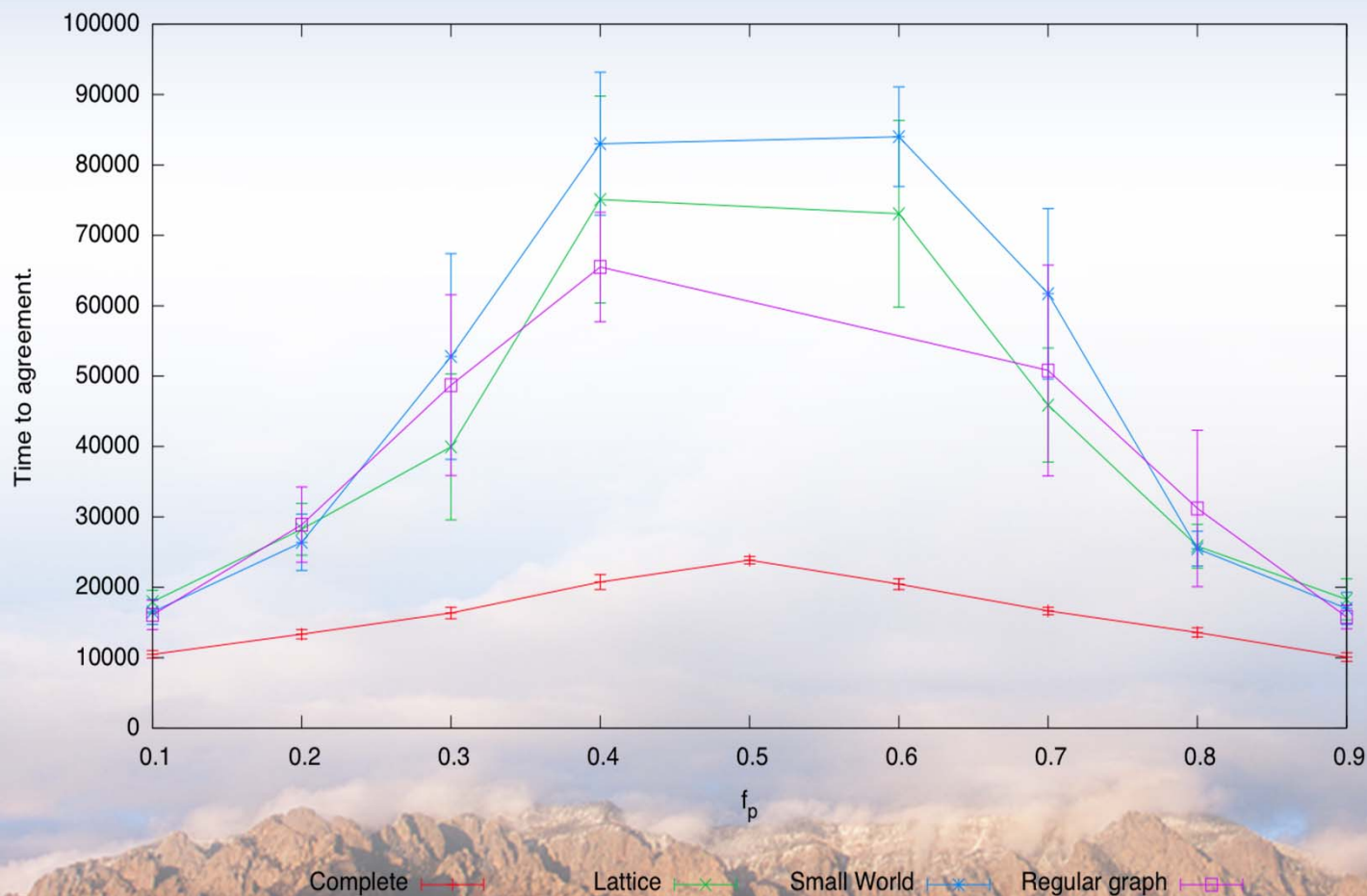
Socio-Cognitive Model



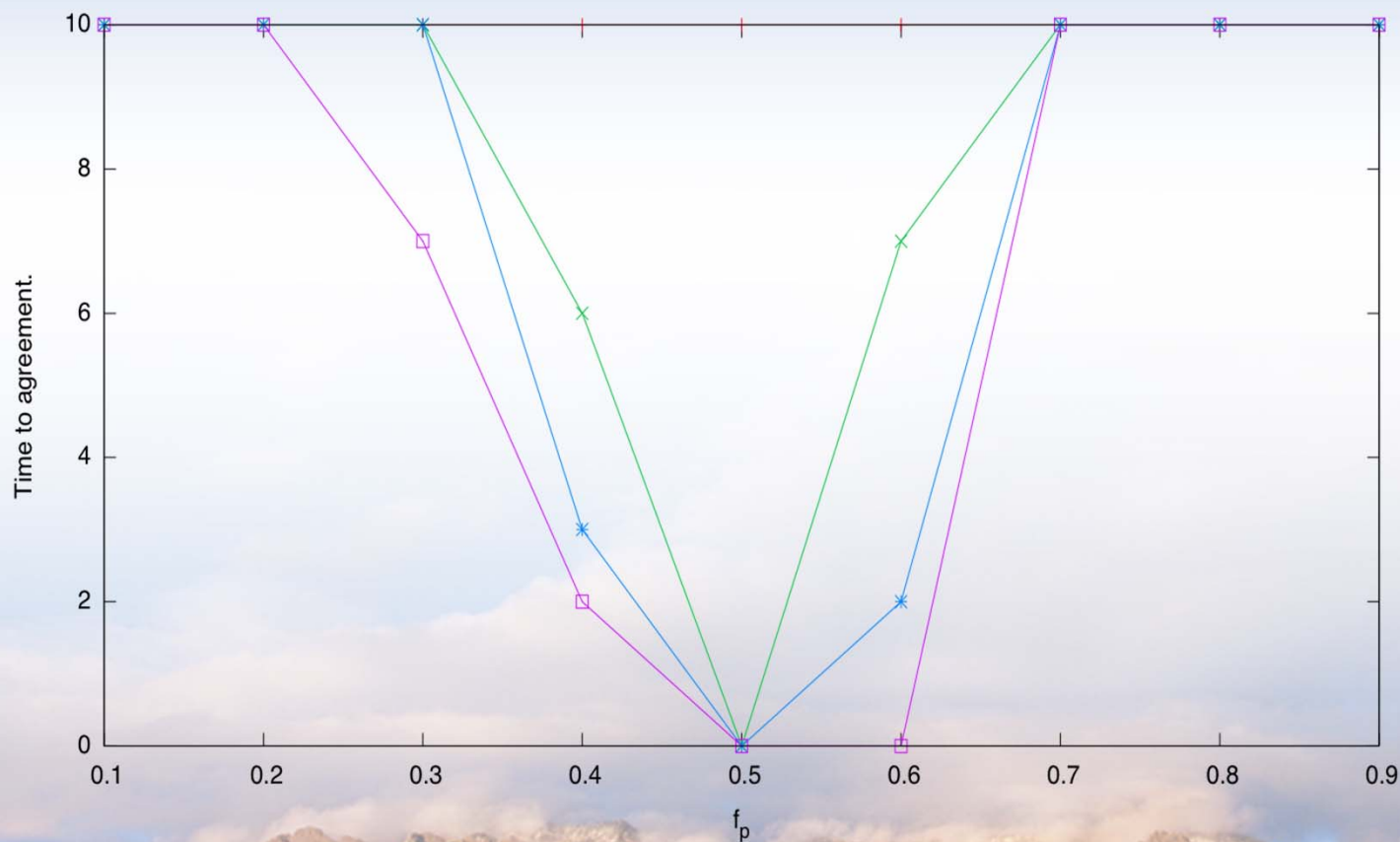
Experiment

- Preliminary experiment: What happens when two differing populations meet?
- Simple cognitive network.
- Initialize f_p fraction of population with positive state.
- Low cognitive effort (using different measurement of effort in these experiments).
- Execute turn dynamics.
- Dependent variables:
 - Does the system reach consensus?
 - How long till consensus?
 - On what state is agreement reached?

Agreement time vs. initial distribution of attitudes



Number of runs that reached agreement vs. initial distribution of attitudes.



Complete —+

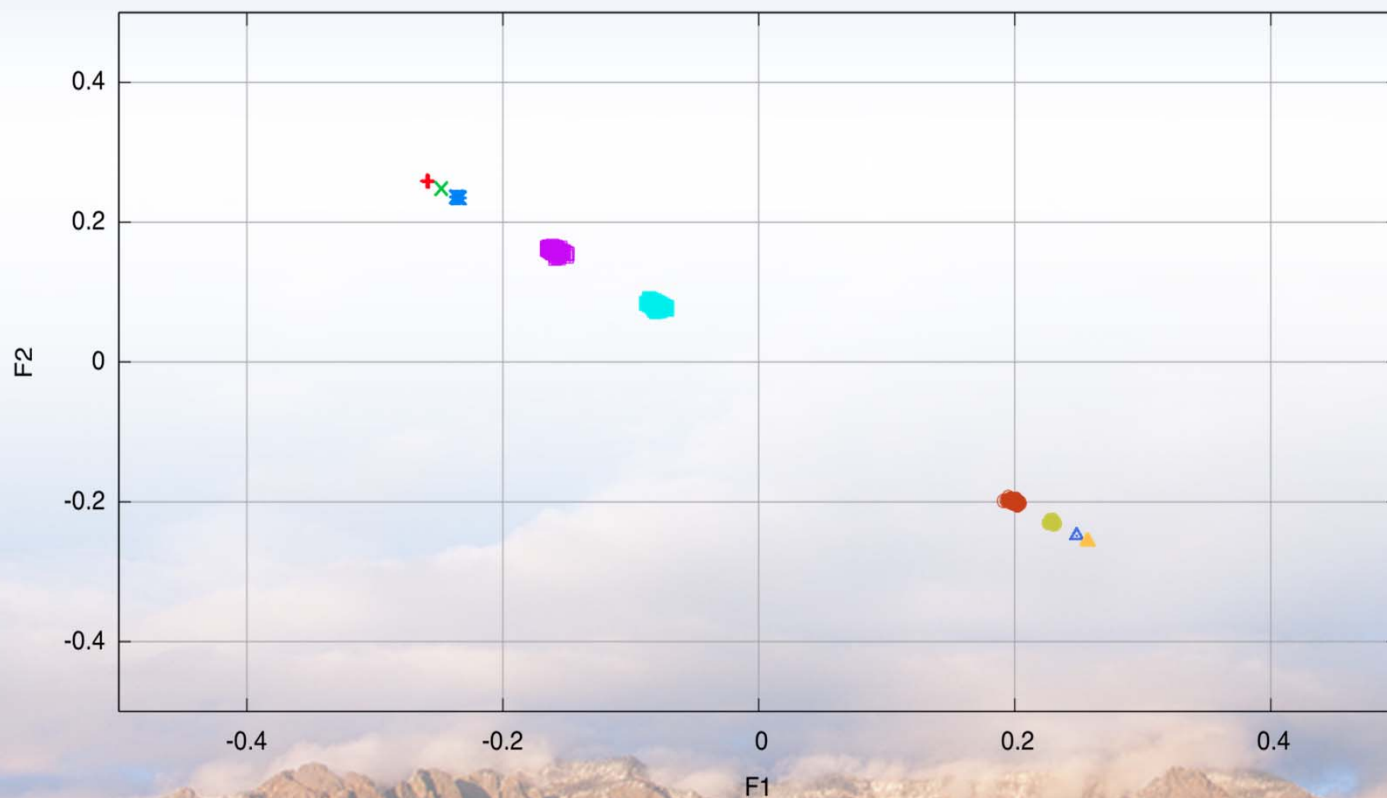
Lattice —x—

Small World —*—

Regular graph —□—

F1/F2 space for a complete graph.

Convergence in F1/F2 space for a Complete Graph at iteration:



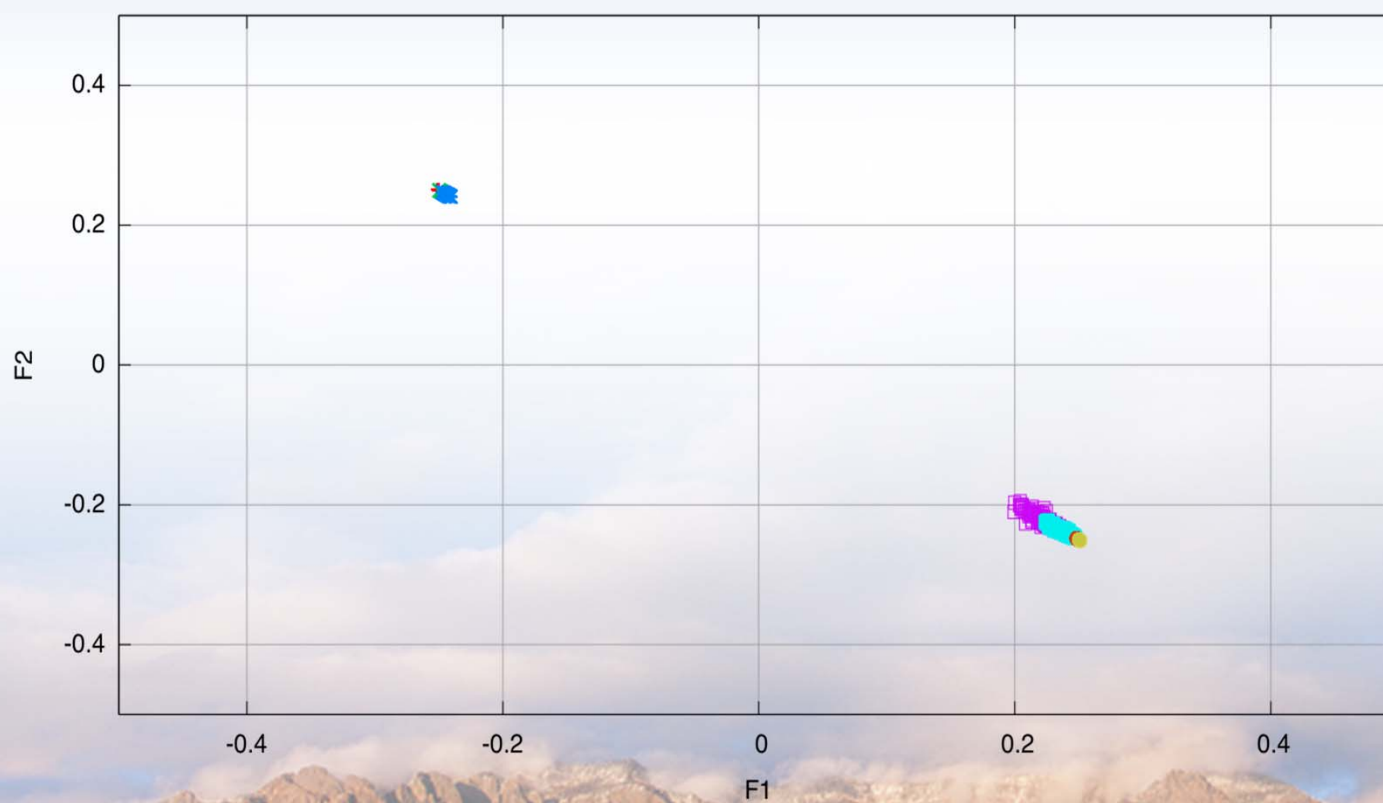
Initial Distribution Impacts Final Attitude



- Each group are the states of all agents at the end of a single run.
- Color indicates the initial distribution (f_p) of attitudes.
 - e.g, 0.1: 10% of initial population with $F1=1, X=1, F2=-1$
- Linear relationship between f_p and final state for a complete graph
- Nonlinear – polarizing effect for lattice graphs.

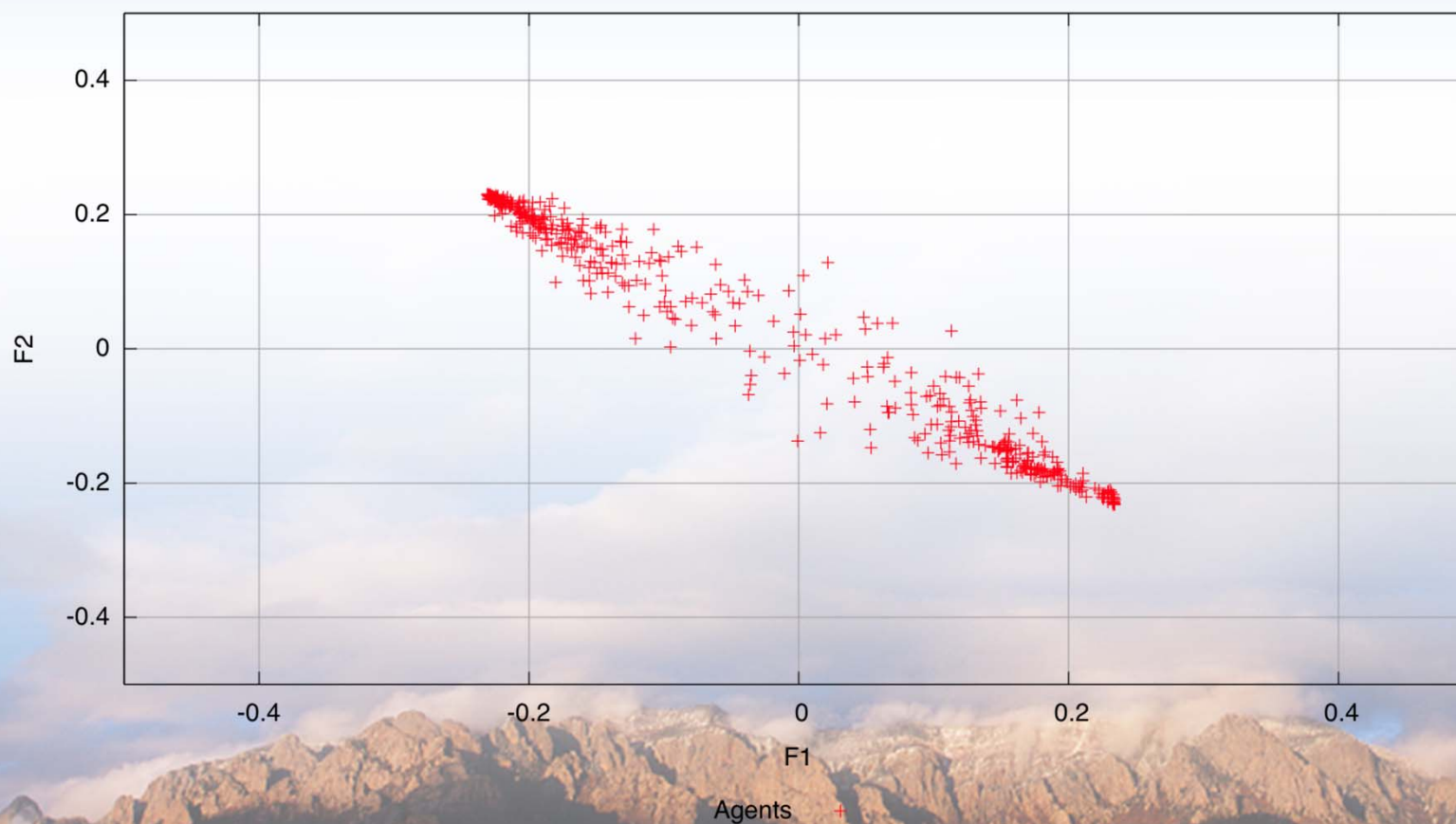
F1/F2 space for a lattice graph

Convergence in F1/F2 space for a Lattice Graph at iteration:



Non converging run for SW networks

Agent in F1/F2 space for a Small World Graph at iteration: 90,000



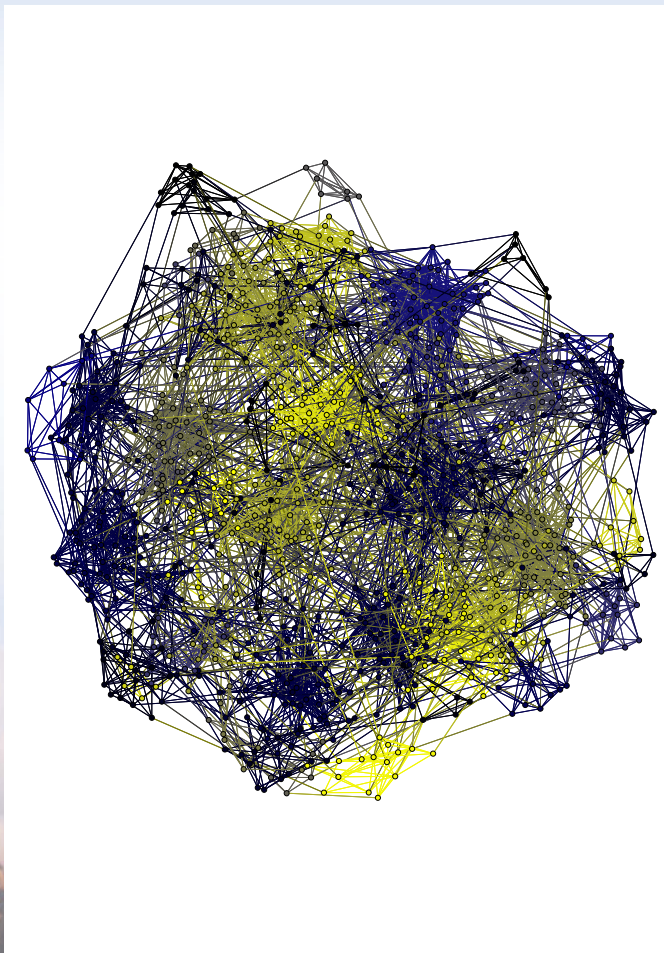


Current Experiments

- Focus on *community structured* social networks.
- Groups of nodes with many links, vs. others with fewer links.



Example Community Graph





Influence Maximization



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- Suppose we have all nodes start with negative opinion.
- We can set m nodes to have positive opinion.
- Which m nodes should we pick to have a positive attitude diffuse through a population?



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Related Work

- Well studied problem for *Independent Cascade* model
 - Progressive: Agent is positive, always stays positive.
 - For each positive neighbor, node has probability p of switching to negative
 - Occurs only once per neighbor.
- Approximation algorithm found.
- Heuristics like degree-based etc. work reasonably well.



Key Difference

- Non-Progressive: Attitudes can change valence and strength.
 - Positive agents can be overwhelmed quickly and turn negative.
- Not a problem in the independent cascade models (since positive agents can never turn back to negative).

Hypothesis: Positive structures



- Hypothesis: Assign positive agents to sub-graph structures rather than individual agents.
- Have “supporting” positive interaction between positive agents.
- Example structure: k-clique.

Doesn't yet work...



The End

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- (Kunda and Thagard, 1996): Z. Kunda and P. Thagard. Forming impressions from stereotypes, traits, and behaviors: A parallel- constraint-satisfaction theory. Psychological Review, 103(2):284–308, 1996.