



*Modelling the Dynamics of
Decision-Making on Networks*

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SIAM Conference on Dynamical Systems

May 22–26, 2011

Energy and Complexity

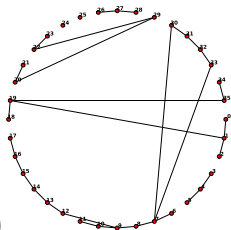
Aim of research:

- Aid city government decision-making on energy policy,
 - e.g. strategy for reducing consumption and fuel poverty.
- Model households as dynamical systems, connected via social network.
- Dissemination of technology or ideas can be studied using models of diffusion on networks.
- Simulate interventions related to adoption of new technology or energy use strategies.
 - e.g. incentives for home insulation.



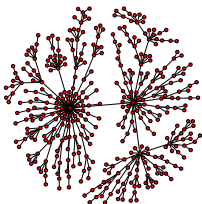
Types of Model Network

- Model networks constructed to give different qualitative features of real world interactions:



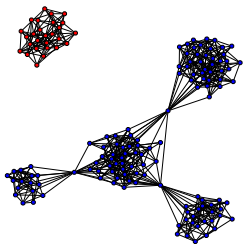
(a)

(a): *Small world* network with 20% rewiring of a regular lattice.



(b)

(b): Preferential attachment graph with a *scale-free* degree distribution.



(c)

(c): Simple model of weakly-connected communities.



Real-World Social Networks

- Different types of social connection exist; these include:
 - geographical neighbours, distant friendships, family trees, *communities*.

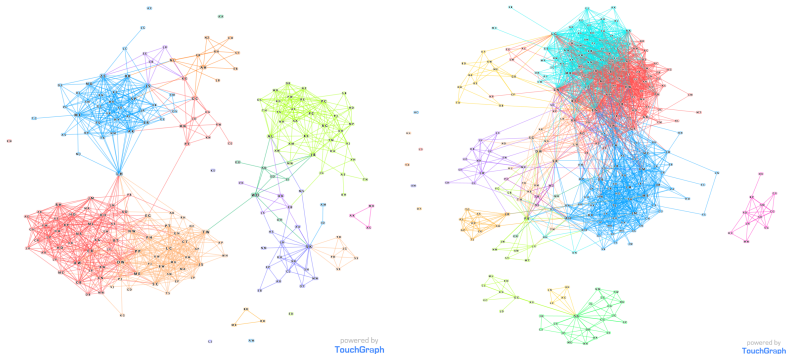
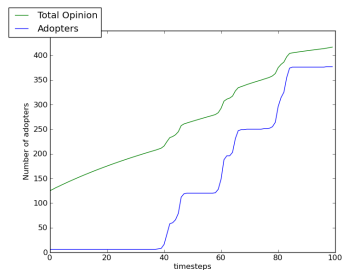
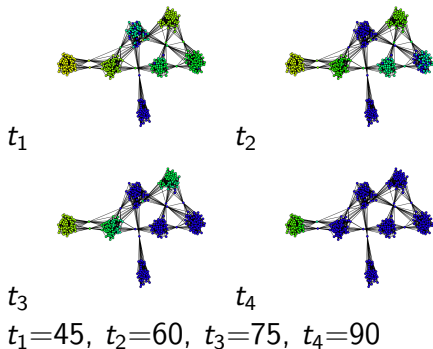


Figure: Inter-friend contacts on the Facebook website.



Example Uptake Results



Model of Social Influences

Individual's preference weighted relative to social influence¹:

- *Utility* (benefit) of product to individual:

$$U = \alpha p + (1 - \alpha)s$$

p: personal utility: intrinsic value of product to individual,
Intrinsic to product and individual, could depend on:

- potential savings,
- relative or absolute,
- pay-back time;
- environmental credentials (may change),
- negative effects of barriers to adoption.

¹S.A. Delre, W. Jager, T.H.A. Bijmolt, and M.A. Janssen. *Will it spread or not? The effects of social influences and network topology on innovation diffusion* (2010)



Model of Social Influences

$$U = \alpha p + (1 - \alpha)s$$

s : social utility: fraction of other individuals with technology,

- Data suggests individuals assign different relative value to personal contacts and society².
 - someone buys when adoption within society and contact network are above respective thresholds,
 - individuals classed as early, majority or late adopters.

α : relative weighting of personal to social value.

²T.W. Valente. *Social network thresholds in the diffusion of innovations* (1996).



Model Specifications

- Purchase state variable $x_i = 0, 1$:
0: not purchased, 1: purchased,
- At each time-step evaluate *utility* of product:

$$U_i = \alpha p_i + \beta s_i + \gamma m$$

$$s_i = \sum_j^{K_i} \rho_{ij} x_j / \sum_j^{K_i} \rho_{ij}, \quad m = \sum_{k=1}^N \sigma_k x_k / \sum_k^N \sigma_k.$$

- If U_i crosses a threshold θ_i then $x_i \rightarrow 1$.



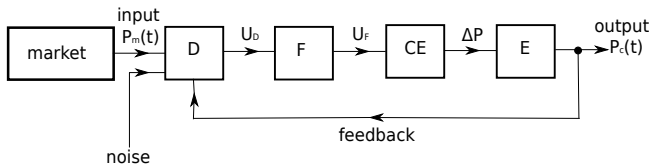
Dynamics of Opinion Formation

- Time-scales for updating opinion (τ_1) and making purchases (τ_2) may be different:
 - τ_1 opinion updated after interacting with friends and taking in media (e.g. daily, weekly),
 - τ_2 purchase decisions made less frequently (motivated by monthly pay-day, weather, prices, breakages etc.).
- Need to model intermediate dynamics of opinion in between decisions.



Modelling Consumer Opinion Formation

- Processes modelled as elements in a control system:



With constant market price and simplest filter, can obtain following³:

$$\dot{X} = \gamma - X - \Phi(X).$$

X current dimensionless
price-deviation,

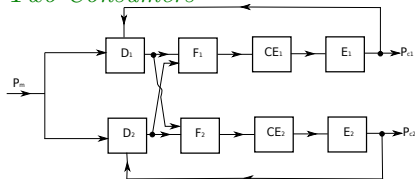
γ initial dimensionless
price-deviation.

$\Phi(X)$ nonlinear function of
deviation.



Interacting Consumers

Two Consumers

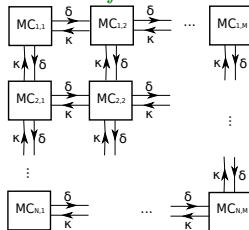


The equations can then be written:

$$\dot{X}_1 + X_1 + \Phi(X_1) = \gamma_1 + \kappa\Phi(X_2),$$

$$\dot{X}_2 + X_2 + \Phi(X_2) = \gamma_2 + \delta\Phi(X_1).$$

Lattice of Consumers



With equations:

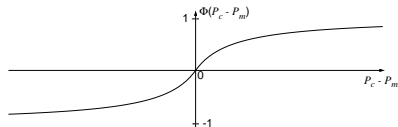
$$\begin{aligned} \dot{X}_{n,m} + X_{n,m} + \Phi(X_{n,m}) &= \gamma_{n,m} \\ &+ \delta\Phi(X_{n-1,m}) + \kappa\Phi(X_{n+1,m}) \\ &+ \delta\Phi(X_{n,m-1}) + \kappa\Phi(X_{n,m+1}). \end{aligned}$$



Analytical Results

Form of the Discriminator

Choose $\Phi(X)$ to limit large deviations:



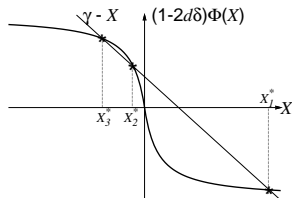
$$\text{e.g.: } \Phi(X) = \frac{\beta X}{1 + |\beta X|}$$

Single Consumer

Equilibrium states found from:

$$\gamma - X = \Phi(X).$$

Two Interacting Consumers



Clustering of Opinion Over Time

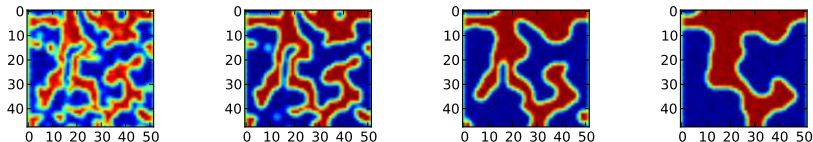
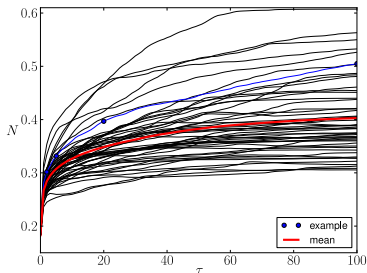


Figure: $\delta = 2$; $\tau = 2, 5, 20, 100$



Measure using *Mix-Norm*^a:

$$N^2 = \sum_{k,l} \frac{|a_{k,l}|^2}{\sqrt{1+k^2+l^2}},$$

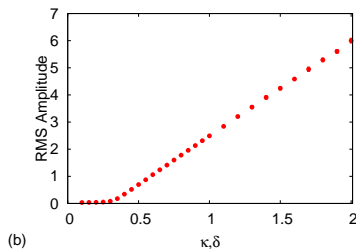
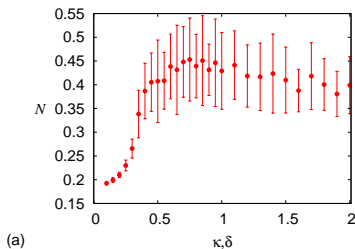
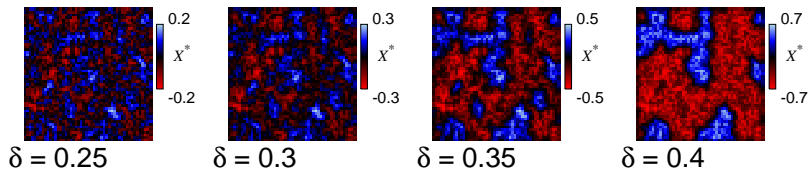
$a_{k,l}$: Fourier transform coefficients;

N larger for coarser structure.

^aMathew *et al.*, Physica D (2005)



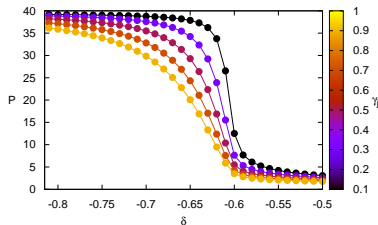
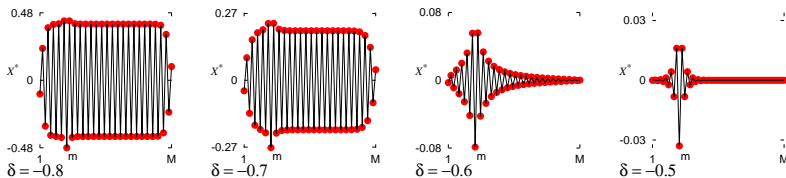
Clustering and Coupling Strength



(All for $\tau = 100$)



Localized Patterns (1D lattice)



Use participation number:

$$P = \frac{1}{\sum z_n^2},$$

$$\text{with } z_n = \frac{|X_n|}{\sum_n |X_n|}.$$



Summary and Future Work

- Spread of energy technologies can be modelled as diffusion on networks.
 - Need to choose correct network model(s).
 - Also dynamics of individual decisions.
- Model exchange of opinions as a coupled dynamical system:
 - find clustering of opinions over time,
 - depends on the strength of opinion exchanges.

Next step:

- Combine opinion dynamics model with discrete purchase decision.

