

Local Network Effects, Rationality and The Structure of Technology Networks

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How I got started on this research

How does one optimally price IT-based products?

- Unusual cost structure
- Threat of digital piracy
- Presence of (mostly) positive network effects

How I got started on this research

How does one optimally price IT-based products?

- Unusual cost structure
- Threat of digital piracy
- Presence of (mostly) positive network effects
 - Network effects depend on individual usage, may be heterogeneous in value across customers
 - *Nonlinear pricing and type-dependent network effects (2004)*
 - *Network effects, nonlinear pricing and entry deterrence (2005)*

Other aspects of network effects (that seem to matter)

- Agents often are not capable of (or interested in) forming rational expectations which are fulfilled
 - They don't have enough information about the preferences of other agents
 - They don't have the ability to compute a rational expectations equilibrium even if they did have the information
 - They don't pay attention to every product all the time
 - They base their beliefs on "local" information
- Adoption is often gradual and "viscous", rather than being instantaneous
 - The realized dynamic process of adoption often determines eventual outcomes

Other aspects of network effects (that seem to matter)

- Network effects are often "local"
 - Interpersonal communication technologies, business to business technologies, online marketplaces...
- The structure of underlying social or business networks affects the adoption of network goods
 - An agent's "local" network affects their value from adoption...
 - ...but so does the structure of the rest of the social network
 - Local networks are connected
 - One's neighbors' local networks affect one's adoption
 - Structure of the "adoption network" (or technology network) depends on the structure of the underlying social network

My research questions

- How is the adoption of a technology which displays network effects affected by:
 - The extent to which the network effects are local
 - The extent to which their value differs across potential adopters
 - The structure of an underlying social or business network
 - The "boundedness" of consumer rationality
- What can one infer about each of these from the observed structure of an adoption network?
- What are the implications of a model of this kind for:
 - Optimal price paths for new network goods
 - Choosing how connected targeted early adopters should be
 - The benefits of mandated IT standards in an organization, or in an inter-organizational supply network
 - The social optimality of universal access to a technology

Models of networks: a framework(?)

Network effects in economics

- Network effects are homogeneous
- Network effects are "global"
- Social network is complete
- Agents are unboundedly rational
- Adoption is instantaneous
- Adoption cost is strategic (price)

"Science of networks" models

- Network effects are homogeneous
- Network effects are local
- Social network is complete
- Agents are myopic
- Adoption is gradual (discrete)
- Adoption cost is constant (zero)

Local network effects...

- Network effects are **heterogeneous**
- Network effects are **local**
- Social network is **any graph**
- Agents are unboundedly rational
- Adoption is instantaneous
- Adoption cost is constant (but can be a one-shot strategic variable)

Dynamic pricing of network...

- Network effects are heterogeneous
- Network effects are "global"
- Social network is complete
- Agents are **boundedly rational** (myopic, stubborn, combination)
- Adoption is **gradual** (continuous)
- Adoption cost is strategic and varies over time

Underlying social network: example

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"Adoption network": an example

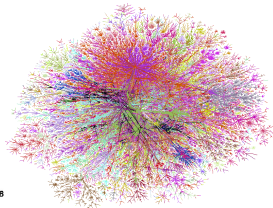
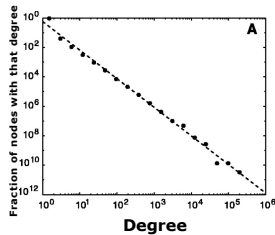
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Adoption networks: another example

- **Degree** of a node: number of other nodes a node is connected to (or number of edges originating from the node)
- **Degree distribution** of a network: Fraction of nodes in a network that have a particular degree, as a function of degree

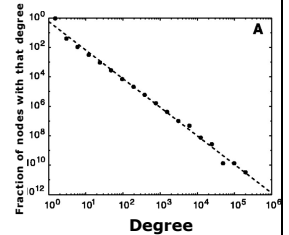


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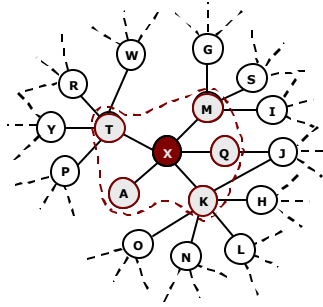


Degree distribution of the Web

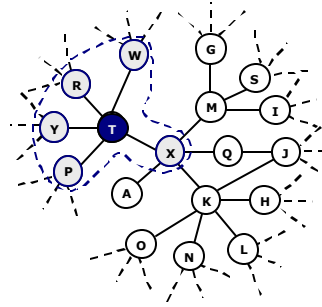
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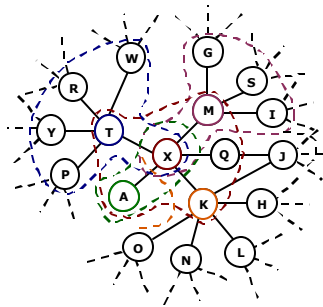
Local networks



Local networks



Local networks



Local network effects

- Agents in this kind of network generally have:
 - different local networks
 - perfect information about the structure of their local network
 - some information about the structure of the other local networks they belong to (their neighbors' local networks)
 - very little or no information about the exact structure of the rest of the social network
- These agents make their adoption decisions based on their local networks, and this information.

A model of local network effects

- Set of potential customers $N = \{1, 2, 3, \dots, n\}$
- Single homogeneous network good that costs c
- Customers connected by an underlying social network modeled as an instance of a random graph (more on this soon).
- Each customer has:
 - A neighbor set G_i
 - A degree d_i (number of neighbors)
 - A valuation type θ_i (measure of adoption complementarity)
- Each customer makes an adoption choice $a_i \in \{0, 1\}$
- Payoff from adoption for customer i :

$$a_i [u(\sum_{j \in G_i} a_j, \theta_i) - c]$$

- More generally formulated in the paper

Where the social network comes from

$$N = \{1, 2, 3, \dots, n\} \quad \Gamma_i = 2^{N \setminus \{i\}}$$

Set of graphs: $\Gamma \subset \Gamma_1 \times \Gamma_2 \times \dots \times \Gamma_n$

Distribution over this set: $\rho : \Gamma \rightarrow [0, 1]$

Drawing from this distribution yields G

Restrictions on the social network (r)

For each x in D , denote

$\Gamma_j(x)$ = subset of Γ_j such that for each $X \in \Gamma_j(x)$, $|X| = x$

Restrict the distribution over r as follows:

For each i , for each $j \in G_i$, $\Pr[G_j \in \Gamma_j(x) | G_i, \theta_i] = q(x)$

For each i , for each $j \notin G_i$, $\Pr[G_j \in \Gamma_j(x) | G_i, \theta_i] = \hat{q}(x)$

Generalizes to posteriors conditional on degree
Admits generalized random graphs, standard models
of "small world" networks

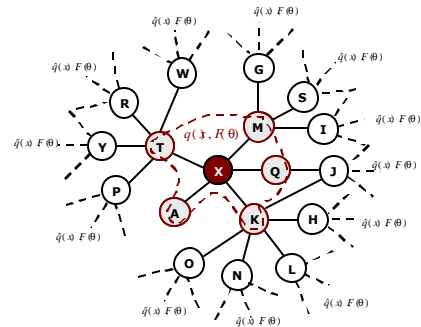
Sequence of the game

- Nature draws θ_i for each i , draws $G \in \Gamma$
- Each agent i observes their type
- Each agent i chooses either to adopt ($a_i=1$) or not ($a_i=0$)
- Payoffs are realized

Information

- After each agent realizes their neighbor set and type:
 - They know the exact structure of their local network
 - They have very little information about the structure of the rest of the network
 - Posterior $\hat{q}(x)$ on degree of non-neighbors
 - They have inexact (but better) information about the structure of the local networks they belong to
 - Posterior $q(x)$ on degree of neighbors
- They know their θ_i , do not know anyone else's

Information



Equilibria

- Each symmetric Bayes-Nash equilibrium involves a threshold strategy:

$$s(d_i, \theta_i) = \begin{cases} 0, & \theta_i < \theta^*(d_i) \\ 1, & \theta_i \geq \theta^*(d_i) \end{cases}$$

with threshold $\theta^* = [\theta(1), \theta(2), \dots, \theta(m)]$

- "No adoption" is always an equilibrium for pure network goods
- The equilibria can be Pareto ordered: $\Theta^* = \{\theta^A, \theta^B, \dots\}$
 $\theta^A < \theta^B < \dots$

Main theory results

- The ordering of equilibria is based on the equilibrium probability of neighbor adoption

$$\lambda(\theta) = \sum_{x=1}^m q(x) [1 - F(\theta(x))]$$

- "Higher" equilibria strictly Pareto-dominate lower ones, and therefore, there is a best equilibrium, which has the highest value of $\lambda(\theta^*)$
- Each fulfilled expectations outcome with a local expectation $\hat{\lambda}$ of neighbor adoption has a corresponding Bayes-Nash equilibrium with $\lambda(\theta^*) = \hat{\lambda}$
 - Coordinating adoption may be simpler if it is (a) local and (b) based on a simple parameter
- Greatest equilibrium is "weakly" coalition proof: establishes a basis for stability in the standard model

The structure of adoption networks

Consider a generalized random graph with degree distribution $p(x)$, and moment generating function (MGF)

$$\Phi_p(w) = \sum_{x=0}^{\infty} p(x)w^x$$

For identical θ , and for a threshold degree δ^* , the MGF of the degree distribution of the adoption network is

$$\Phi_\alpha(w) = \Phi_p[1 - \bar{Q}(\delta^*) + w\bar{Q}(\delta^*)]$$

where

$$\bar{Q}(x) = \Pr[d_j \geq x \mid j \in G_i] = \sum_{j=x}^m q(x)$$

Summary: Models of networks

Network effects in economics

- Network effects are homogeneous
- Network effects are "global"
- Social network is complete
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Local network effects...

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- Network effects are **local**
- Social network is **arbitrary graph**
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Dynamic pricing of network...

- Network effects are **heterogeneous**
- Network effects are **"global"**
- Social network is **complete**
- Agents are **boundedly rational (myopic, stubborn, combination)**
- Adoption is **gradual (continuous)**
- Adoption cost is **strategic and varies over time**

Summary of results

- Simple way of modeling adoption of a technology with local network effects as a game of incomplete information between agents connected in an underlying social network
- This game has at least one (and generally many) symmetric Bayes-Nash equilibria in pure strategies
 - All equilibria involve generalized threshold strategies (a threshold degree associated with each value of the agent's "strength" of network effect)
 - These equilibria can be strictly Pareto ranked, based on a simple parameter: the probability a neighbor might adopt
 - One-to-one mapping between equilibria of the game and "fulfilled expectations equilibria" with local expectations

Summary of results

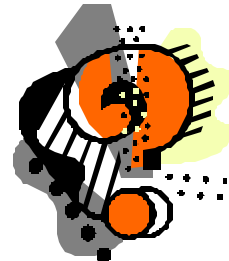
- A simple closed-form expression that describes the structure of an adoption network in terms of the structure of the social network (and vice versa)
- Some answers to other questions
 - Monopoly pricing is generally higher than a standard model that ignores network structure would predict
 - A monopolist always gives free versions to a fraction of their customers (and if possible, would target low-degree customers rather than highly connected customers)
 - The social optimality of universal access (or the optimality of mandated IT standards) relies on social/business networks not being too clustered.

Summary of related results

In a model of adoption with "boundedly rational" expectation formation, and bounded attention to changes in prices:

- For the corresponding model with unboundedly rational consumers: constant optimal price.
- This rational expectations equilibrium price is never a steady state of the optimal dynamic pricing policy
- When customers are myopic, for a range of forms of customer heterogeneity, the optimal price path is a target policy:
 - Price at zero until a critical mass is reached ("bargains")
 - Set a steady state price, higher than the price predicted by the rational expectations model, after critical mass is reached ("ripoffs")
- This result generalizes to
 - Mixtures between myopic and unboundedly rational
 - Mixtures between myopic and "stubborn" (for at least one example)

Questions and discussion?



<http://oz.stern.nyu.edu/research/>