

# Economics of Networks Networked Markets

Evan Sadler

Massachusetts Institute of Technology

April 26, 2018

# Agenda

Perfect matchings

Bargaining

Competitive equilibrium in a two-sided market

Supply networks and aggregate volatility

Suggested Reading:

- EK chapters 10 and 11; Jackson chapter 10
- Manea (2011), “Bargaining in Stationary Networks”
- Acemoglu et al. (2012), “The Network Origins of Aggregate Fluctuations”

# Buyer-Seller Networks

Often assume trade is unrestricted: any buyer can costlessly interact with any seller

# Buyer-Seller Networks

Often assume trade is unrestricted: any buyer can costlessly interact with any seller

Not true in practice:

- Product heterogeneity
- Geographic proximity
- Search costs
- Reputation

# Buyer-Seller Networks

Often assume trade is unrestricted: any buyer can costlessly interact with any seller

Not true in practice:

- Product heterogeneity
- Geographic proximity
- Search costs
- Reputation

Develop theory of buyer-seller networks

- Connections to bargaining, auctions, market-clearing prices

# Buyer-Seller Networks

Often assume trade is unrestricted: any buyer can costlessly interact with any seller

Not true in practice:

- Product heterogeneity
- Geographic proximity
- Search costs
- Reputation

Develop theory of buyer-seller networks

- Connections to bargaining, auctions, market-clearing prices

Questions:

- Can every buyer (seller) find a seller (buyer)?
- Do market clearing prices exist?
- Is the outcome of trade efficient?

# Perfect Matchings

A simple model:

- Disjoint sets of buyers and sellers  $B$  and  $S$ ,  $|B| = |S| = n$
- Bipartite graph  $G$  (all edges connect a buyer to a seller)
- Write  $N(A)$  for set of neighbors of agents in  $A$
- A *matching* is a subset of edges such that no two share an endpoint

# Perfect Matchings

A simple model:

- Disjoint sets of buyers and sellers  $B$  and  $S$ ,  $|B| = |S| = n$
- Bipartite graph  $G$  (all edges connect a buyer to a seller)
- Write  $N(A)$  for set of neighbors of agents in  $A$
- A *matching* is a subset of edges such that no two share an endpoint

Say  $i$  and  $j$  are matched if the matching contains an edge between them



# Perfect Matchings

A simple model:

- Disjoint sets of buyers and sellers  $B$  and  $S$ ,  $|B| = |S| = n$
- Bipartite graph  $G$  (all edges connect a buyer to a seller)
- Write  $N(A)$  for set of neighbors of agents in  $A$
- A *matching* is a subset of edges such that no two share an endpoint

Say  $i$  and  $j$  are matched if the matching contains an edge between them

A matching is *perfect* if every buyer is matched to a seller and vice versa

- Contains  $\frac{n}{2}$  edges

# Perfect Matchings

## Theorem

*The bipartite graph  $G$  has a matching of size  $|S|$  if and only if for every  $A \subseteq S$  we have  $|N(A)| \geq |A|$*

# Perfect Matchings

## Theorem

*The bipartite graph  $G$  has a matching of size  $|S|$  if and only if for every  $A \subseteq S$  we have  $|N(A)| \geq |A|$*

Clearly necessary (why?), sufficiency is harder

# Perfect Matchings

## Theorem

*The bipartite graph  $G$  has a matching of size  $|S|$  if and only if for every  $A \subseteq S$  we have  $|N(A)| \geq |A|$*

Clearly necessary (why?), sufficiency is harder

Call a set  $A \subseteq S$  with  $|A| > |N(A)|$  a constricted set

# Perfect Matchings

## Theorem

*The bipartite graph  $G$  has a matching of size  $|S|$  if and only if for every  $A \subseteq S$  we have  $|N(A)| \geq |A|$*

Clearly necessary (why?), sufficiency is harder

Call a set  $A \subseteq S$  with  $|A| > |N(A)|$  a constricted set

Elegant alternating paths algorithm to find maximum matching and constricted sets (EK, 10.6)

# Rubinstein Bargaining

A seemingly unrelated problem...

# Rubinstein Bargaining

A seemingly unrelated problem...

Consider one buyer and one seller

- Seller has an item the buyer wants
- Seller values at 0, buyer at 1
- At time 1, seller proposes a price, buyer accepts or rejects
- If accept, game ends, realize payoffs
- If reject, proceed to time 2, buyer makes offer

# Rubinstein Bargaining

A seemingly unrelated problem...

Consider one buyer and one seller

- Seller has an item the buyer wants
- Seller values at 0, buyer at 1
- At time 1, seller proposes a price, buyer accepts or rejects
- If accept, game ends, realize payoffs
- If reject, proceed to time 2, buyer makes offer

Bargaining with alternating offers



# Rubinstein Bargaining

A seemingly unrelated problem...

Consider one buyer and one seller

- Seller has an item the buyer wants
- Seller values at 0, buyer at 1
- At time 1, seller proposes a price, buyer accepts or rejects
- If accept, game ends, realize payoffs
- If reject, proceed to time 2, buyer makes offer

Bargaining with alternating offers

Players are impatient, discount future at rate  $\delta$

# The One-Shot Deviation Principle

Game has infinite time horizon, cannot use backward induction

- Payoff is a discounted infinite sum

# The One-Shot Deviation Principle

Game has infinite time horizon, cannot use backward induction

- Payoff is a discounted infinite sum

Useful fact: one-shot deviation principle

# The One-Shot Deviation Principle

Game has infinite time horizon, cannot use backward induction

- Payoff is a discounted infinite sum

Useful fact: one-shot deviation principle

## Theorem (Blackwell, 1965)

*In an infinite horizon game with bounded per-period payoffs, a strategy profile  $s$  is a SPE if and only if for each player  $i$  there is no profitable deviation  $s'_i$  that agrees with  $s_i$  everywhere except at a single time  $t$ .*

# The One-Shot Deviation Principle

Game has infinite time horizon, cannot use backward induction

- Payoff is a discounted infinite sum

Useful fact: one-shot deviation principle

## Theorem (Blackwell, 1965)

*In an infinite horizon game with bounded per-period payoffs, a strategy profile  $s$  is a SPE if and only if for each player  $i$  there is no profitable deviation  $s'_i$  that agrees with  $s_i$  everywhere except at a single time  $t$ .*

HUGE simplification: only need to check deviations in a single period

- Proof is beyond our scope

# Rubinstein Bargaining

Consider a profile of the following form:

- There is a pair of prices  $(p_s, p_b)$
- The seller always proposes  $p_s$  and accepts any  $p \geq p_b$
- The buyer always offers  $p_b$  and accepts any  $p \leq p_s$

# Rubinstein Bargaining

Consider a profile of the following form:

- There is a pair of prices  $(p_s, p_b)$
- The seller always proposes  $p_s$  and accepts any  $p \geq p_b$
- The buyer always offers  $p_b$  and accepts any  $p \leq p_s$

Suppose the seller proposes in the current period

# Rubinstein Bargaining

Consider a profile of the following form:

- There is a pair of prices  $(p_s, p_b)$
- The seller always proposes  $p_s$  and accepts any  $p \geq p_b$
- The buyer always offers  $p_b$  and accepts any  $p \leq p_s$

Suppose the seller proposes in the current period

Acceptance earns the buyer  $1 - p_s$ , rejection earns  $\delta(1 - p_b)$

- Incentive compatible if  $p_s \leq 1 - \delta + \delta p_b$



# Rubinstein Bargaining

Consider a profile of the following form:

- There is a pair of prices  $(p_s, p_b)$
- The seller always proposes  $p_s$  and accepts any  $p \geq p_b$
- The buyer always offers  $p_b$  and accepts any  $p \leq p_s$

Suppose the seller proposes in the current period

Acceptance earns the buyer  $1 - p_s$ , rejection earns  $\delta(1 - p_b)$

- Incentive compatible if  $p_s \leq 1 - \delta + \delta p_b$

Similarly, when buyer proposes, acceptance is incentive compatible if  $p_b \geq \delta p_s$

# Rubinstein Bargaining

In equilibrium, seller proposes highest acceptable price

- $p_s = 1 - \delta + \delta p_b$

# Rubinstein Bargaining

In equilibrium, seller proposes highest acceptable price

- $p_s = 1 - \delta + \delta p_b$

Similarly, buyer offers lowest acceptable price

- $p_b = \delta p_s$

# Rubinstein Bargaining

In equilibrium, seller proposes highest acceptable price

- $p_s = 1 - \delta + \delta p_b$

Similarly, buyer offers lowest acceptable price

- $p_b = \delta p_s$

Solving yields

$$p_s^* = \frac{1}{1 + \delta}, \quad p_b^* = \frac{\delta}{1 + \delta}$$

# Rubinstein Bargaining

In equilibrium, seller proposes highest acceptable price

- $p_s = 1 - \delta + \delta p_b$

Similarly, buyer offers lowest acceptable price

- $p_b = \delta p_s$

Solving yields

$$p_s^* = \frac{1}{1 + \delta}, \quad p_b^* = \frac{\delta}{1 + \delta}$$

## Theorem (Rubinstein 1982)

*The alternating offers bargaining game has a unique SPE with offers  $(p_s^*, p_b^*)$  that are immediately accepted.*

# Bargaining in a Bipartite Network

Let's extend this framework to a bipartite graph  $G$  connecting sellers  $S$  to buyers  $B$

# Bargaining in a Bipartite Network

Let's extend this framework to a bipartite graph  $G$  connecting sellers  $S$  to buyers  $B$

At time 1, sellers simultaneously announce prices

- A buyer can accept a single offer from a linked seller
- All buyers who accept offers are cleared from the market along with their sellers
- In case of ties, social planner chooses trades to maximize total number of transactions

# Bargaining in a Bipartite Network

Let's extend this framework to a bipartite graph  $G$  connecting sellers  $S$  to buyers  $B$

At time 1, sellers simultaneously announce prices

- A buyer can accept a single offer from a linked seller
- All buyers who accept offers are cleared from the market along with their sellers
- In case of ties, social planner chooses trades to maximize total number of transactions

Others proceed to time 2, when buyers make offers

- Alternating offers framework as before
- Previous model equivalent to a single buyer linked to a single seller



## Example: Two Sellers, One Buyer

Suppose there are two sellers linked to a single buyer

## Example: Two Sellers, One Buyer

Suppose there are two sellers linked to a single buyer

Buyer will choose seller who offers lowest price

- If sellers offer same  $p > 0$ , buyer randomizes
- Profitable deviation: offer  $p - \epsilon$  to ensure a sale

## Example: Two Sellers, One Buyer

Suppose there are two sellers linked to a single buyer

Buyer will choose seller who offers lowest price

- If sellers offer same  $p > 0$ , buyer randomizes
- Profitable deviation: offer  $p - \epsilon$  to ensure a sale

In unique SPE, both sellers offer  $p = 0$

- Logic is reminiscent of Bertrand competition
- The “short” side of the market has all the bargaining power

# Bargaining in Networks

What if there are two buyers and one seller?

# Bargaining in Networks

What if there are two buyers and one seller?

- Same logic applies, sells at price 1

# Bargaining in Networks

What if there are two buyers and one seller?

- Same logic applies, sells at price 1

What if there are two buyers, each linked to same two sellers?

# Bargaining in Networks

What if there are two buyers and one seller?

- Same logic applies, sells at price 1

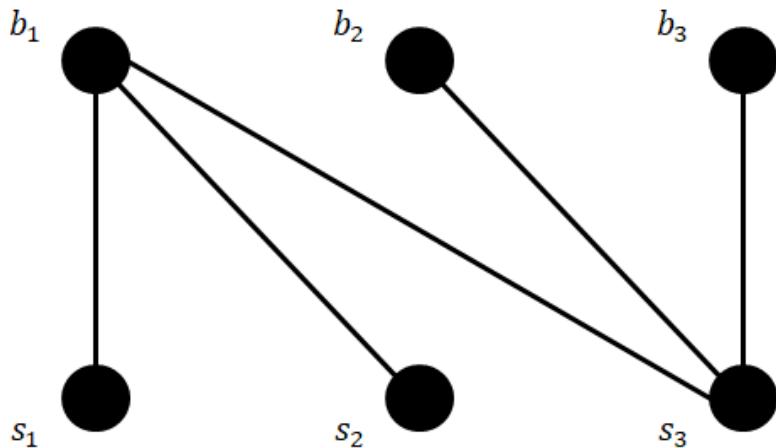
What if there are two buyers, each linked to same two sellers?

Work backwards, what happens if one pair trades and exits the market?

- Bargaining power is the same as in the one buyer one seller example

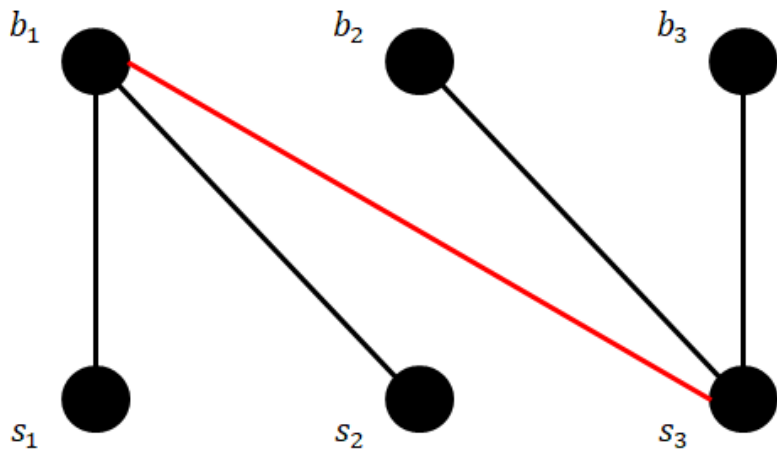
# Bargaining in Networks

Less clear what happens in more complicated graphs





# Redundant Links



# Bargaining in Networks

Existence of perfect matching ensures near-equal bargaining power

# Bargaining in Networks

Existence of perfect matching ensures near-equal bargaining power

Once we eliminate redundant links, reduction to three cases

- Price 0, 1, or close to  $\frac{1}{2}$

# Bargaining in Networks

Existence of perfect matching ensures near-equal bargaining power

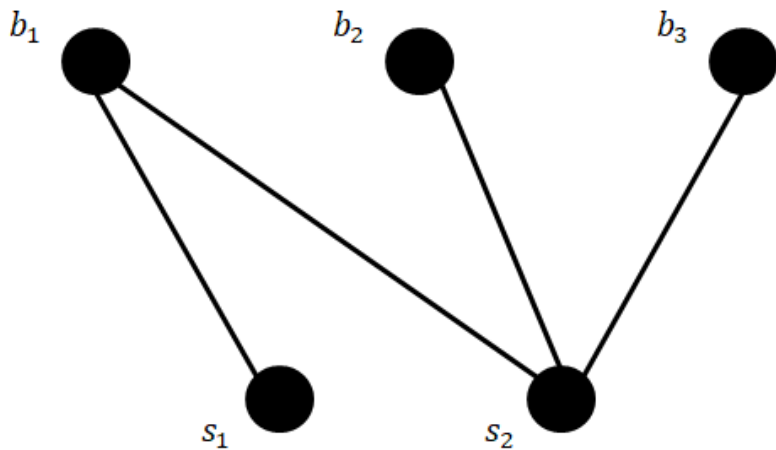
Once we eliminate redundant links, reduction to three cases

- Price 0, 1, or close to  $\frac{1}{2}$

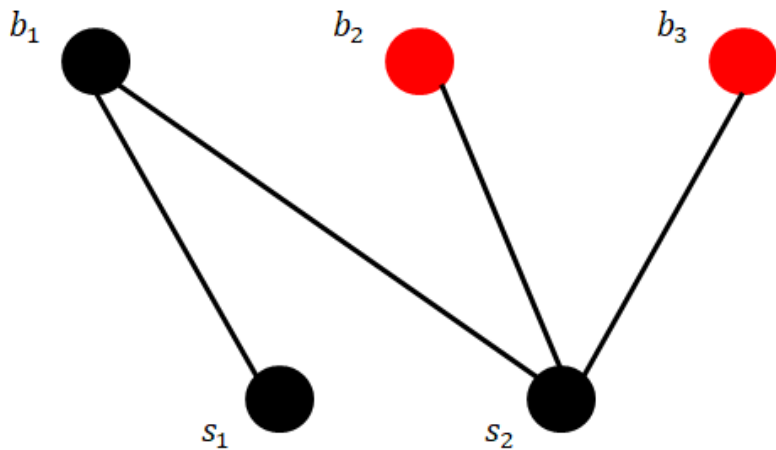
Decomposition algorithm, three sets  $G_S$ ,  $G_B$ ,  $G_E$  initially empty

- First, identify sets of two or more sellers linked to a single buyer, remove and add to  $G_S$
- Next, identify remaining sets of two or more buyers linked to a single seller, remove and add to  $G_B$
- Repeat: for each  $k \geq 2$ , look for sets of  $k + 1$  or more sellers (buyers) linked to  $k$  buyers (sellers); remove and add to the corresponding sets
- Add remaining players to  $G_E$

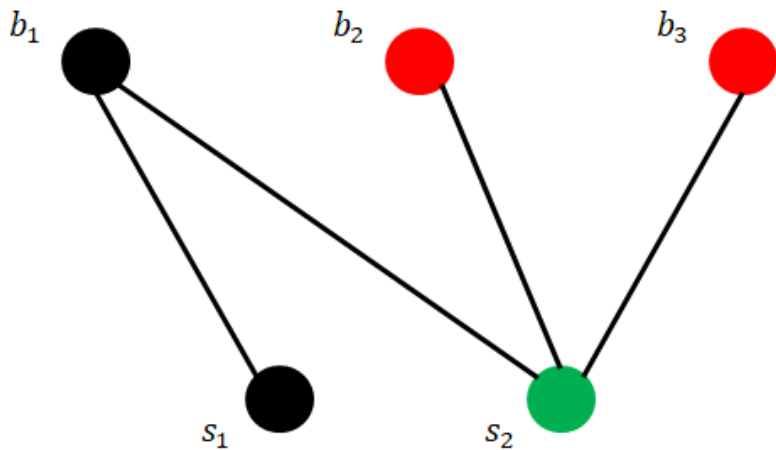
# Decomposition Example



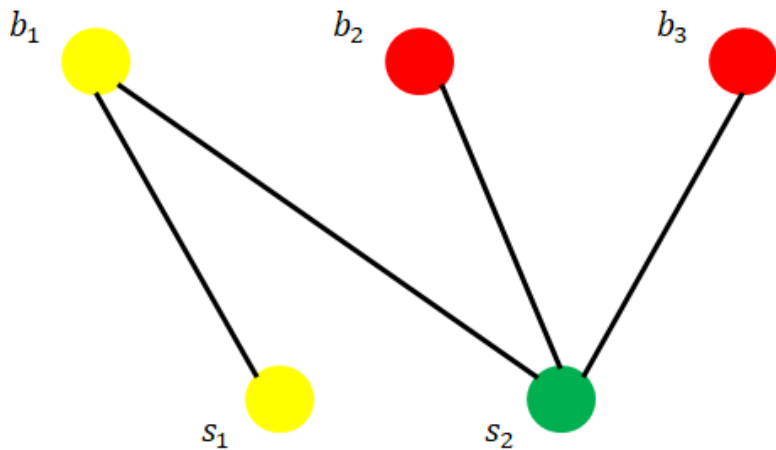
# Decomposition Example



# Decomposition Example



# Decomposition Example





# Bargaining in Networks

This simple algorithm pins down bargaining payoffs

## Theorem

*There exists a SPE in which:*

- *Sellers in  $G_S$  get 0, buyers in  $G_S$  get 1*
- *Sellers in  $G_B$  get 1, buyers in  $G_B$  get 0*
- *Sellers in  $G_E$  get  $\frac{1}{1+\delta}$ , buyers in  $G_E$  get  $\frac{\delta}{1+\delta}$*

# Bargaining in Networks

This simple algorithm pins down bargaining payoffs

## Theorem

*There exists a SPE in which:*

- *Sellers in  $G_S$  get 0, buyers in  $G_S$  get 1*
- *Sellers in  $G_B$  get 1, buyers in  $G_B$  get 0*
- *Sellers in  $G_E$  get  $\frac{1}{1+\delta}$ , buyers in  $G_E$  get  $\frac{\delta}{1+\delta}$*

Prediction matches well with experimental findings

# Valuations and Prices

Suppose now that buyers have heterogeneous values for different sellers' products

- Each seller has an item, values it at zero, wants to maximize profits
- Posted price

# Valuations and Prices

Suppose now that buyers have heterogeneous values for different sellers' products

- Each seller has an item, values it at zero, wants to maximize profits
- Posted price

Buyer  $i$  values seller  $j$  at  $v_{ij}$ , wants at most one object

- Buy from seller  $j$ , pay  $p_j \geq 0$
- Buyer utility  $v_{ij} - p_j$ , seller utility  $p_j$

# Valuations and Prices

Suppose now that buyers have heterogeneous values for different sellers' products

- Each seller has an item, values it at zero, wants to maximize profits
- Posted price

Buyer  $i$  values seller  $j$  at  $v_{ij}$ , wants at most one object

- Buy from seller  $j$ , pay  $p_j \geq 0$
- Buyer utility  $v_{ij} - p_j$ , seller utility  $p_j$

The transaction generates surplus  $v_{ij}$

# Valuations and Prices

For a buyer  $i$ , set of *preferred sellers* given prevailing prices  $\mathbf{p}$

$$D_i(\mathbf{p}) = \{j : v_{ij} - p_j = \max_k [v_{ik} - p_k]\}$$

Preferred seller graph contains edge  $ij$  if and only if  $j \in D_i(\mathbf{p})$

# Valuations and Prices

For a buyer  $i$ , set of *preferred sellers* given prevailing prices  $\mathbf{p}$

$$D_i(\mathbf{p}) = \{j : v_{ij} - p_j = \max_k [v_{ik} - p_k]\}$$

Preferred seller graph contains edge  $ij$  if and only if  $j \in D_i(\mathbf{p})$

A perfect matching in the preferred seller graph means we can match every buyer to a preferred seller, and no item is allocated to more than one buyer

- Note, whether such a matching exists will depend on the prices

# Valuations and Prices

Who sells to whom?



# Valuations and Prices

Who sells to whom?

## Definition

A price vector  $\mathbf{p}$  is *competitive* if there is an assignment  $\mu : B \rightarrow S \cup \{\emptyset\}$  such that  $\mu(i) \in D_i \mathbf{p}$ , and if  $\mu(i) = \mu(i')$  for some  $i \neq i'$ , then  $\mu(i) = \emptyset$  (i.e. buyer  $i$  is unmatched). The pair  $(\mathbf{p}, \mu)$  is a *competitive equilibrium* if  $\mathbf{p}$  is competitive, and additionally if seller  $j$  is unmatched in  $\mu$ , then  $p_j = 0$ .

# Valuations and Prices

Who sells to whom?

## Definition

A price vector  $\mathbf{p}$  is *competitive* if there is an assignment  $\mu : B \rightarrow S \cup \{\emptyset\}$  such that  $\mu(i) \in D_i \mathbf{p}$ , and if  $\mu(i) = \mu(i')$  for some  $i \neq i'$ , then  $\mu(i) = \emptyset$  (i.e. buyer  $i$  is unmatched). The pair  $(\mathbf{p}, \mu)$  is a *competitive equilibrium* if  $\mathbf{p}$  is competitive, and additionally if seller  $j$  is unmatched in  $\mu$ , then  $p_j = 0$ .

Competitive equilibrium prices are market-clearing prices

- Equate supply and demand
- Corresponds to perfect matching in preferred seller graph

# Existence and Efficiency

## Theorem (Shapley and Shubik, 1972)

*A competitive equilibrium always exists. Moreover, a competitive equilibrium maximizes the total valuation for buyers across all matchings (i.e. it maximizes total surplus).*

# Existence and Efficiency

## Theorem (Shapley and Shubik, 1972)

*A competitive equilibrium always exists. Moreover, a competitive equilibrium maximizes the total valuation for buyers across all matchings (i.e. it maximizes total surplus).*

Proof beyond our scope

More general versions of this result are known as the First Fundamental Theorem of Welfare

# Bargaining in Stationary Networks

What if there are multiple opportunities to trade over time?

# Bargaining in Stationary Networks

What if there are multiple opportunities to trade over time?

Simplest stationary model:

- Set of players  $N = \{1, 2, \dots, n\}$
- Undirected graph  $G$
- Common discount rate  $\delta$
- No buyer-seller distinction, any pair can generate a unit surplus

# Bargaining in Stationary Networks

What if there are multiple opportunities to trade over time?

Simplest stationary model:

- Set of players  $N = \{1, 2, \dots, n\}$
- Undirected graph  $G$
- Common discount rate  $\delta$
- No buyer-seller distinction, any pair can generate a unit surplus

In each period, a directed link  $ij$  is chosen uniformly at random

- Player  $i$  proposes a division to player  $j$
- Player  $j$  accepts or rejects

# Bargaining in Stationary Networks

What if there are multiple opportunities to trade over time?

Simplest stationary model:

- Set of players  $N = \{1, 2, \dots, n\}$
- Undirected graph  $G$
- Common discount rate  $\delta$
- No buyer-seller distinction, any pair can generate a unit surplus

In each period, a directed link  $ij$  is chosen uniformly at random

- Player  $i$  proposes a division to player  $j$
- Player  $j$  accepts or rejects

If accept, players exit the game and are replaced by new, identical players



# Bargaining in Stationary Networks

## Theorem

*There exists a unique payoff vector  $v$  such that in every subgame perfect equilibrium, the expected payoff to player  $i$  in any subgame is  $v_i$ . Whenever  $i$  is selected to make an offer to  $j$ , we have*

- If  $\delta(v_i + v_j) < 1$ , then  $i$  offers  $\delta v_j$  to  $j$ , and  $j$  accepts*
- If  $\delta(v_i + v_j) > 1$ , then  $i$  makes an offer that  $j$  rejects*

Proof is beyond our scope; for generic  $\delta$ , always have  $\delta(v_i + v_j) \neq 1$

# Bargaining in Stationary Networks

## Theorem

*There exists a unique payoff vector  $v$  such that in every subgame perfect equilibrium, the expected payoff to player  $i$  in any subgame is  $v_i$ . Whenever  $i$  is selected to make an offer to  $j$ , we have*

- If  $\delta(v_i + v_j) < 1$ , then  $i$  offers  $\delta v_j$  to  $j$ , and  $j$  accepts*
- If  $\delta(v_i + v_j) > 1$ , then  $i$  makes an offer that  $j$  rejects*

Proof is beyond our scope; for generic  $\delta$ , always have  $\delta(v_i + v_j) \neq 1$

Intuition for strategies:  $\delta(v_i + v_j)$  is the joint outside option

- Players make a deal if doing so is better than the outside option for both

# Bargaining in Stationary Networks

Can place bounds on payoffs in *limit equilibria*

- As  $\delta \rightarrow 1$ , equilibrium payoff vectors converge to a vector  $\mathbf{v}^*$

# Bargaining in Stationary Networks

Can place bounds on payoffs in *limit equilibria*

- As  $\delta \rightarrow 1$ , equilibrium payoff vectors converge to a vector  $\mathbf{v}^*$

Let  $M$  denote an independent set of players (no two linked)

- Let  $L(M)$  denote set of players linked to those in  $M$

# Bargaining in Stationary Networks

Can place bounds on payoffs in *limit equilibria*

- As  $\delta \rightarrow 1$ , equilibrium payoff vectors converge to a vector  $\mathbf{v}^*$

Let  $M$  denote an independent set of players (no two linked)

- Let  $L(M)$  denote set of players linked to those in  $M$

## Theorem

*For any independent set  $M$ , we have*

$$\min_{i \in M} v_i^* \leq \frac{|L(M)|}{|M| + |L(M)|}, \quad \max_{j \in L(M)} v_j^* \geq \frac{|M|}{|M| + |L(M)|}$$

Manea (2011) provides an algorithm to compute the payoffs

# Supply Networks

During the financial crises, policy makers feared that firm failures could propagate through the economy

- The president of Ford lobbied for GM and Chrysler to be bailed out
- Feared that common suppliers would go bankrupt, disrupting Ford's operations

# Supply Networks

During the financial crises, policy makers feared that firm failures could propagate through the economy

- The president of Ford lobbied for GM and Chrysler to be bailed out
- Feared that common suppliers would go bankrupt, disrupting Ford's operations

Such cascade effects are not a feature of standard theory

- In a perfectly competitive market with many firms, the effects of a shock to one are spread evenly across the others
- A failure has a small effect on aggregate output

# Supply Networks

During the financial crises, policy makers feared that firm failures could propagate through the economy

- The president of Ford lobbied for GM and Chrysler to be bailed out
- Feared that common suppliers would go bankrupt, disrupting Ford's operations

Such cascade effects are not a feature of standard theory

- In a perfectly competitive market with many firms, the effects of a shock to one are spread evenly across the others
- A failure has a small effect on aggregate output

Structure of supply networks can help tell us when cascade effects are possible and how severe they might be



# Supply Networks: A Model

Variant of a multisector input-output model

- Representative household endowed with one unit of labor
- Household has Cobb-Douglas preferences over  $n$  goods:

$$u(c_1, c_2, \dots, c_n) = A \prod_{i=1}^n (c_i)^{1/n}$$

- Each good  $i$  produced by a competitive sector, can be consumed or used as input to other sectors
- Output of sector  $i$  is

$$x_i = z_i^\alpha l_i^\alpha \prod_{j=1}^n x_{ij}^{(1-\alpha)w_{ij}}$$

- $l_i$  is the labor input,  $x_{ij}$  is the amount of commodity  $j$  used to produce commodity  $i$ ,  $w_{ij}$  is the input share of commodity  $j$ ,  $z_i$  is a sector productivity shock (independent across sectors)

# Supply Networks: A Model

Output:

$$x_i = z_i^\alpha l_i^\alpha \prod_{j=1}^n x_{ij}^{(1-\alpha)w_{ij}}$$

# Supply Networks: A Model

Output:

$$x_i = z_i^\alpha l_i^\alpha \prod_{j=1}^n x_{ij}^{(1-\alpha)w_{ij}}$$

Assumption:  $\sum_{j=1}^n w_{ij} = 1$

- Constant returns to scale

# Supply Networks: A Model

Output:

$$x_i = z_i^\alpha l_i^\alpha \prod_{j=1}^n x_{ij}^{(1-\alpha)w_{ij}}$$

Assumption:  $\sum_{j=1}^n w_{ij} = 1$

- Constant returns to scale

Input-output matrix  $W$  with entries  $w_{ij}$  captures inter-sector relationships

- Can think of  $W$  as a weighted network linking sectors

# Supply Networks: A Model

Output:

$$x_i = z_i^\alpha l_i^\alpha \prod_{j=1}^n x_{ij}^{(1-\alpha)w_{ij}}$$

Assumption:  $\sum_{j=1}^n w_{ij} = 1$

- Constant returns to scale

Input-output matrix  $W$  with entries  $w_{ij}$  captures inter-sector relationships

- Can think of  $W$  as a weighted network linking sectors

Define weighted out-degree  $d_i = \sum_{j=1}^n w_{ji}$ , and let  $F_i$  be the distribution of  $\epsilon_i = \log z_i$

# Supply Networks: A Model

Output:

$$x_i = z_i^\alpha l_i^\alpha \prod_{j=1}^n x_{ij}^{(1-\alpha)w_{ij}}$$

Assumption:  $\sum_{j=1}^n w_{ij} = 1$

- Constant returns to scale

Input-output matrix  $W$  with entries  $w_{ij}$  captures inter-sector relationships

- Can think of  $W$  as a weighted network linking sectors

Define weighted out-degree  $d_i = \sum_{j=1}^n w_{ji}$ , and let  $F_i$  be the distribution of  $\epsilon_i = \log z_i$

Economy characterized by a set of sectors  $N$ , distribution of sector shocks  $\{F_i\}_{i \in N}$ , network  $W$

# Equilibrium Output

Acemoglu et al (2012) show that the output in equilibrium (i.e. when the representative consumer maximizes utility and firms maximize profits) is given by

$$y \equiv \log(GDP) = \sum_{i=1}^n v_i \epsilon_i$$

where  $\mathbf{v}$  is the *influence vector*

$$\mathbf{v} = \frac{\alpha}{n} [I - (1 - \alpha)W']^{-1} \mathbf{1}$$

# Equilibrium Output

Acemoglu et al (2012) show that the output in equilibrium (i.e. when the representative consumer maximizes utility and firms maximize profits) is given by

$$y \equiv \log(GDP) = \sum_{i=1}^n v_i \epsilon_i$$

where  $\mathbf{v}$  is the *influence vector*

$$\mathbf{v} = \frac{\alpha}{n} [I - (1 - \alpha)W']^{-1} \mathbf{1}$$

Influence vector is closely related to Bonacich centrality



# Equilibrium Output

Acemoglu et al (2012) show that the output in equilibrium (i.e. when the representative consumer maximizes utility and firms maximize profits) is given by

$$y \equiv \log(GDP) = \sum_{i=1}^n v_i \epsilon_i$$

where  $\mathbf{v}$  is the *influence vector*

$$\mathbf{v} = \frac{\alpha}{n} [I - (1 - \alpha)W']^{-1} \mathbf{1}$$

Influence vector is closely related to Bonacich centrality

Shocks to more central sectors have a larger impact on aggregate output

# Aggregate Volatility

Let  $\sigma_i^2$  denote the variance of  $\epsilon_i$

# Aggregate Volatility

Let  $\sigma_i^2$  denote the variance of  $\epsilon_i$

We can compute the standard deviation of aggregate output as

$$\sqrt{\text{var}(y)} = \sqrt{\sum_{i=1}^n \sigma_i^2 v_i^2}$$

# Aggregate Volatility

Let  $\sigma_i^2$  denote the variance of  $\epsilon_i$

We can compute the standard deviation of aggregate output as

$$\sqrt{\text{var}(y)} = \sqrt{\sum_{i=1}^n \sigma_i^2 v_i^2}$$

If we have a lower bound on sector output variances  $\underline{\sigma}$ , then this implies

$$\sqrt{\text{var}(y)} = \Theta(\|v\|_2)$$

# Aggregate Volatility

Let  $\sigma_i^2$  denote the variance of  $\epsilon_i$

We can compute the standard deviation of aggregate output as

$$\sqrt{\text{var}(y)} = \sqrt{\sum_{i=1}^n \sigma_i^2 v_i^2}$$

If we have a lower bound on sector output variances  $\underline{\sigma}$ , then this implies

$$\sqrt{\text{var}(y)} = \Theta(\|v\|_2)$$

Volatility scales with the Euclidean norm of the influence vector

# Example

Suppose all sectors supply each other equally

- $w_{ij} = \frac{1}{n}$  for all  $i, j$

## Example

Suppose all sectors supply each other equally

- $w_{ij} = \frac{1}{n}$  for all  $i, j$

The influence vector then has  $v_i = \frac{c}{n}$  for some  $c$  and all  $i$

## Example

Suppose all sectors supply each other equally

- $w_{ij} = \frac{1}{n}$  for all  $i, j$

The influence vector then has  $v_i = \frac{c}{n}$  for some  $c$  and all  $i$

Also assume  $\sigma_i = \sigma$  for all  $i$



## Example

Suppose all sectors supply each other equally

- $w_{ij} = \frac{1}{n}$  for all  $i, j$

The influence vector then has  $v_i = \frac{c}{n}$  for some  $c$  and all  $i$

Also assume  $\sigma_i = \sigma$  for all  $i$

Aggregate volatility is then

$$\sqrt{\text{var}(y)} = \sigma \sqrt{\sum_{i=1}^n v_i^2} = \frac{\sigma c}{\sqrt{n}}$$

Goes to zero as number of sectors becomes large

# Example

Suppose we have a dominant sector 1 that is the only supplier to all others

- $w_{1j} = 1$  for all  $j$

## Example

Suppose we have a dominant sector 1 that is the only supplier to all others

- $w_{1j} = 1$  for all  $j$

This implies  $v_1 = c$  for some  $c$ , independent of  $n$

## Example

Suppose we have a dominant sector 1 that is the only supplier to all others

- $w_{1j} = 1$  for all  $j$

This implies  $v_1 = c$  for some  $c$ , independent of  $n$

This implies a lower bound on aggregate volatility

$$\sqrt{\text{var}(y)} \geq \sigma_1 c$$

Volatility does not shrink with  $n$

# Asymptotics

Can interpret economy with large  $n$  as more disaggregated

- Increased specialization
- Might expect less volatility

# Asymptotics

Can interpret economy with large  $n$  as more disaggregated

- Increased specialization
- Might expect less volatility

For economy with  $n$  sectors, define the coefficient of variation

$$CV(d^{(n)}) = \frac{STD(d^{(n)})}{\bar{d}}$$

# Asymptotics

Can interpret economy with large  $n$  as more disaggregated

- Increased specialization
- Might expect less volatility

For economy with  $n$  sectors, define the coefficient of variation

$$CV(d^{(n)}) = \frac{STD(d^{(n)})}{\bar{d}}$$

## Theorem

*Consider a sequence of economies with increasing  $n$ . Aggregate volatility satisfies*

$$\sqrt{\text{var}(y)} \geq c \frac{1 + CV(d^{(n)})}{\sqrt{n}}$$

*for some  $c$ .*