

A Network Approach to Public Goods

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Cambridge

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 - Conceptually: market outcomes \leftrightarrow network centrality measures.

Outline

- 1** Setup
- 2 Efficiency
- 3 Lindahl Outcomes and Network Centrality
- 4 Conclusions

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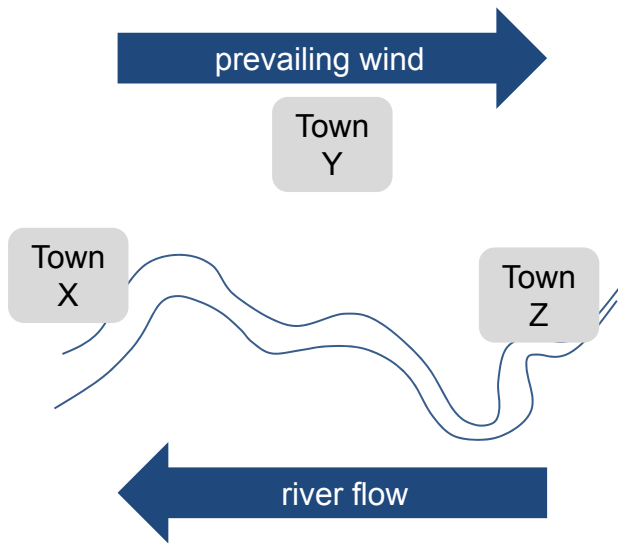
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- positive externalities: $\frac{\partial u_i}{\partial a_j} \geq 0$ if $i \neq j$.

The Environment: An Example



B : The (Marginal) Benefits Matrix

Definition

$$B_{ij} = \begin{cases} \frac{\partial u_i / \partial a_j}{-\partial u_i / \partial a_i} & \text{if } i \neq j \\ 0 & \text{otherwise} \end{cases}$$

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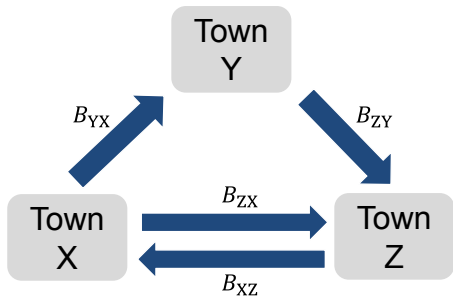
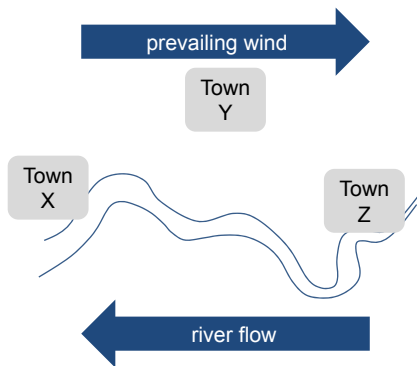
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We assume $B(\mathbf{a})$ is irreducible for all \mathbf{a} .

The Benefits Matrix

We can think of $B(a)$ as a network.

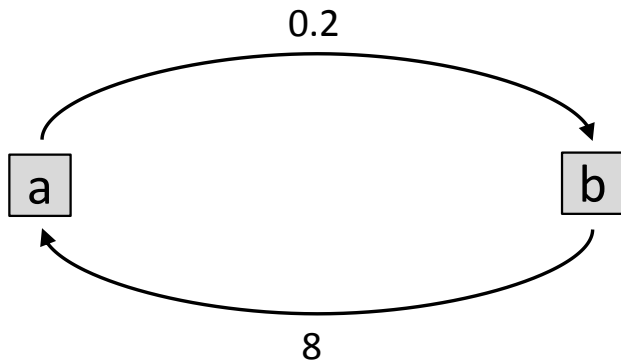


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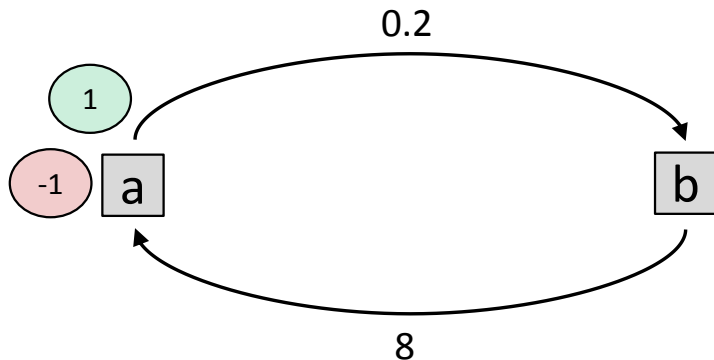
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$$B(\mathbf{0}) = \begin{bmatrix} 0 & 8 \\ 0.2 & 0 \end{bmatrix}$$



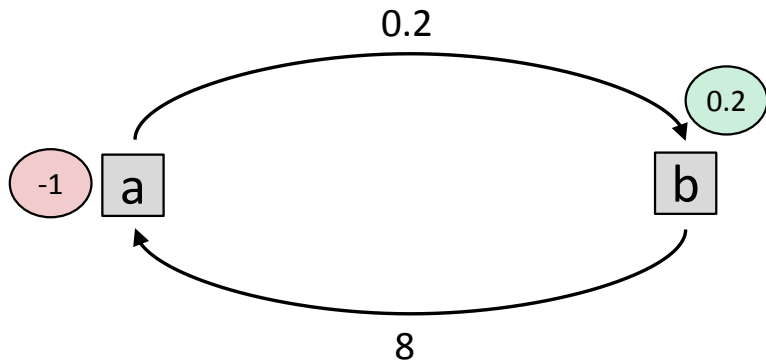
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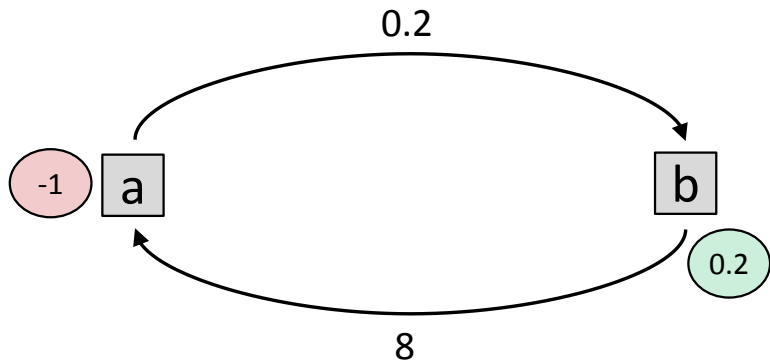
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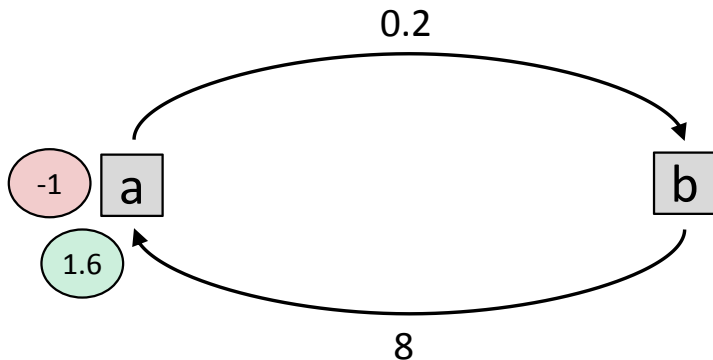
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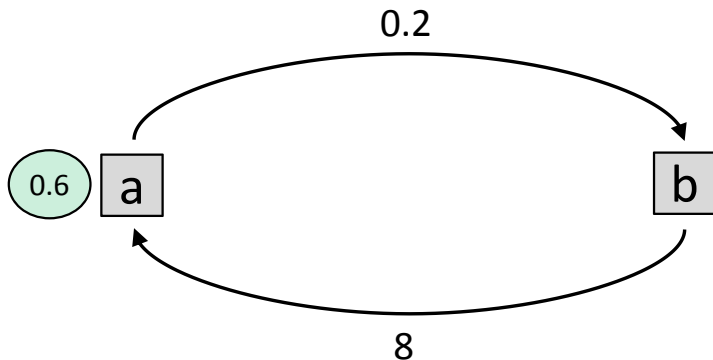
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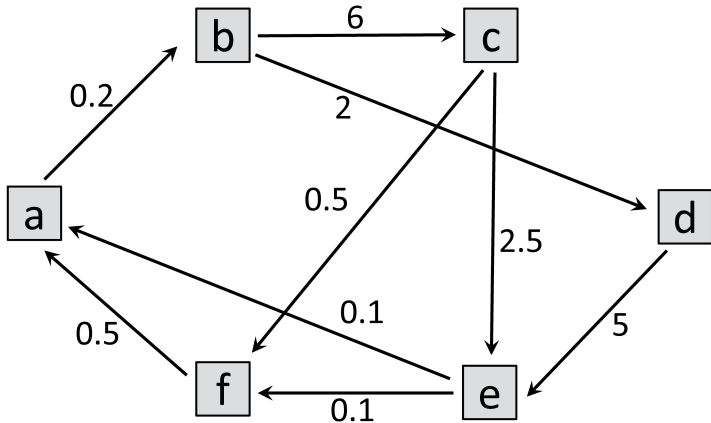
Example: Is a Pareto Improvement Possible?

$$B(\mathbf{0}) = \begin{bmatrix} 0 & B_{12} \\ B_{21} & 0 \end{bmatrix}$$

Result

A Pareto improvement on the status quo $\mathbf{a} = \mathbf{0}$ exists if and only if $B_{12} \cdot B_{21} > 1$.

A More Complicated Example



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The spectral radius $r(\mathbf{M})$ is the maximum magnitude of any eigenvalue of \mathbf{M} .

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An interior action profile \mathbf{a} is Pareto efficient if and only if $r(\mathbf{B}(\mathbf{a})) = 1$.

Proof Sketch: \mathbf{a}^* Pareto-efficient $\Rightarrow r(\mathbf{B}(\mathbf{a}^*)) = 1$

Take PE \mathbf{a}^* , assume $\frac{\partial u_i}{\partial a_i}(\mathbf{a}^*) = -1$.

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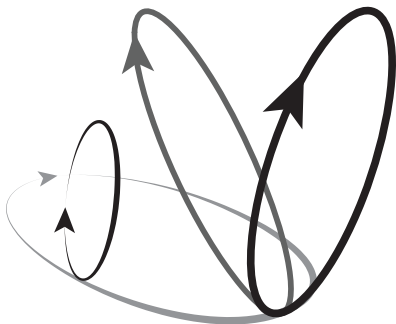


Perron-Frobenius: an eigenvalue λ of \mathbf{B} has a nonnegative left (right) eigenvector if and only if $\lambda = r(\mathbf{B})$. Moreover, \mathbf{B} has an eigenvalue $\lambda \in \mathbb{R}$ equal to $r(\mathbf{B})$.

Interpretation of Spectral Radius

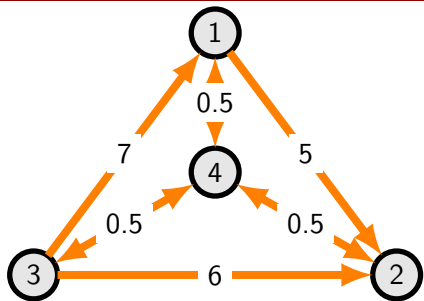
Vague Statement

The spectral radius measures the number/intensity of **cycles** in the benefits matrix.



Spectral Radius in Terms of Cycles

$$B(\mathbf{0}) = \begin{bmatrix} 0 & 0 & 7 & 0.5 \\ 5 & 0 & 6 & 0.5 \\ 0 & 0 & 0 & 0.5 \\ 0.5 & 0.5 & 0.5 & 0 \end{bmatrix}$$

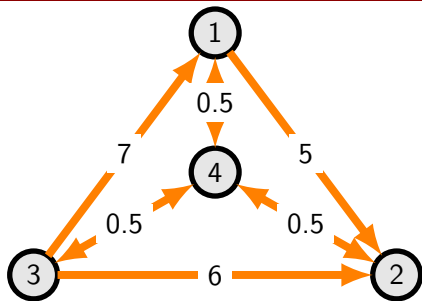


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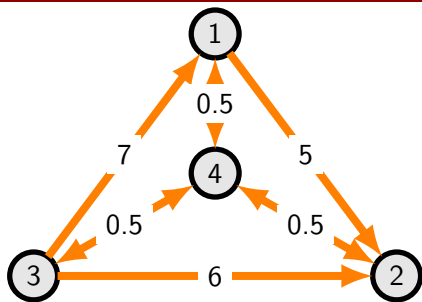
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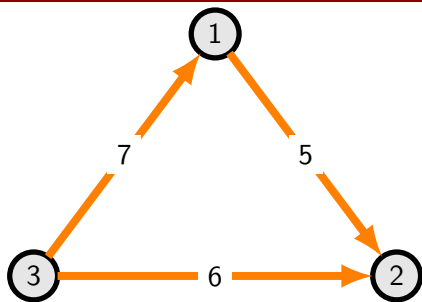
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Player 4 is essential.



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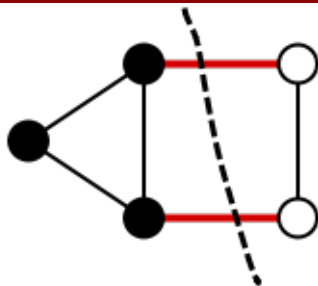
- $(m_i)_{i \in N}$ *deters deviations from \mathbf{a}^** if the restriction of \mathbf{a}^* to M is Pareto efficient given new payoffs (resp. M^c).
- *cost of separation* $c_M(\mathbf{a}^*)$ defined as the infimum of $\sum_{i \in N} m_i(\mathbf{a}^*)$, taken over deviation-detering transfers.

Efficient Separation

Proposition

$$c_M(\mathbf{a}^*) \leq \sum \frac{\theta_i}{\theta_j} B_{ij}(\mathbf{a}^*) a_j^*,$$

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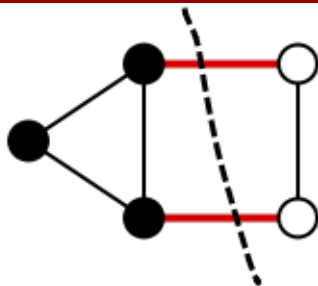
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A minimum cut in a graph with suitable weights \mathbf{W} .



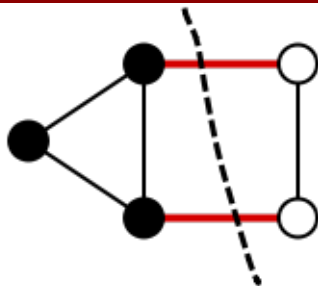
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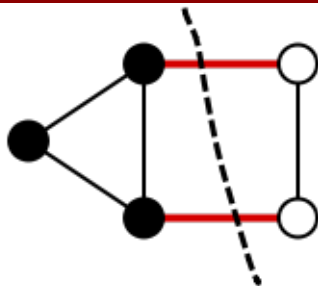
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- Small when *spectral gap* of \mathbf{W} is small.

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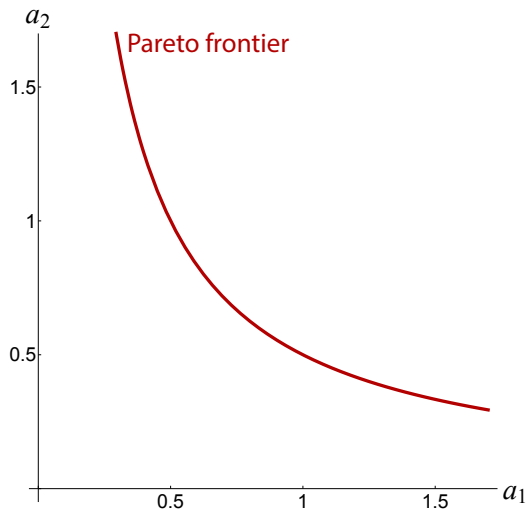
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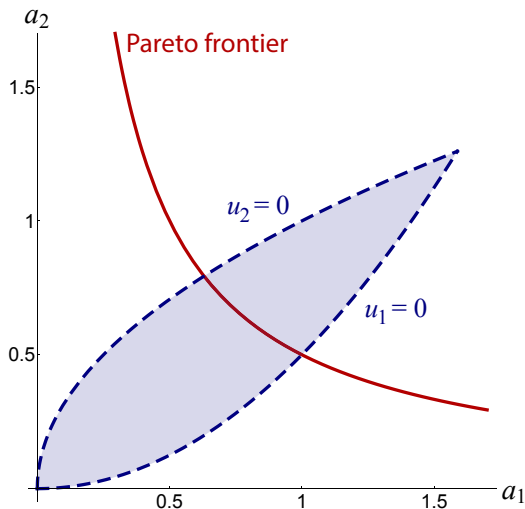
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Multiple Pareto Efficient, Individually Rational Outcomes

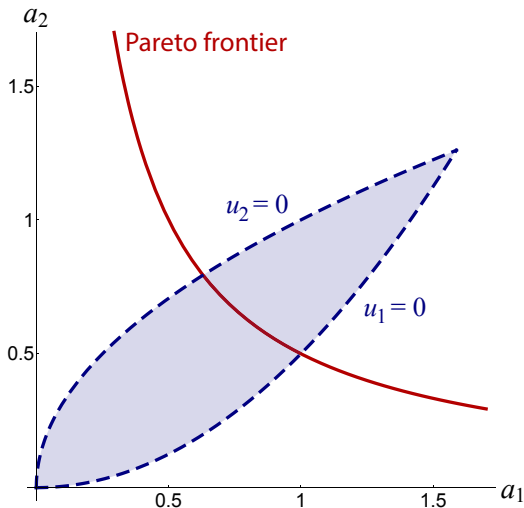


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From now on,
assume set of
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Lindahl Outcome

Conceptually: complete the missing markets for externalities to achieve efficient provision.

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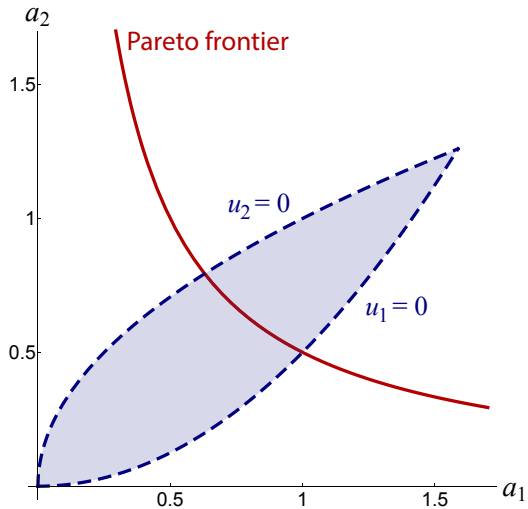
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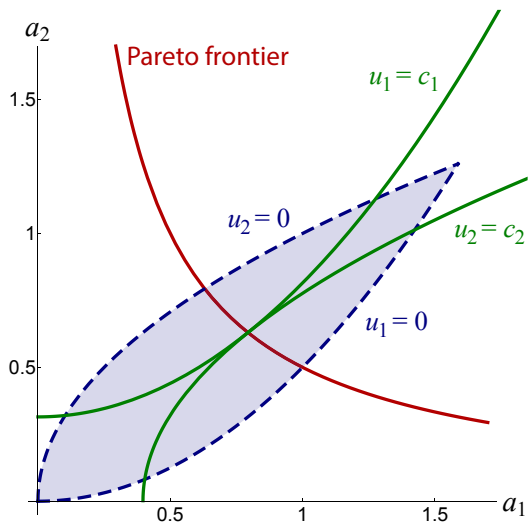
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Main theorem: characterization in terms of network centrality.

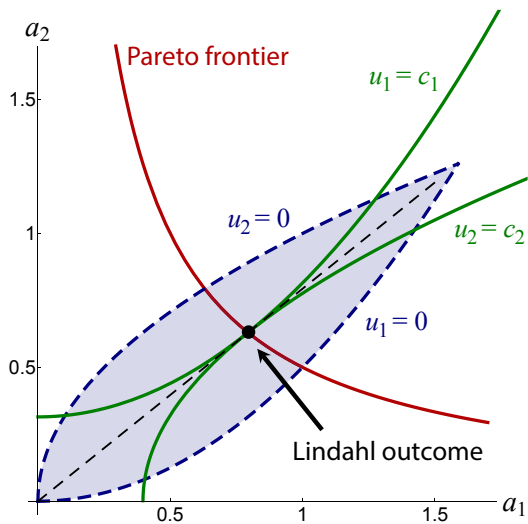
Lindahl Outcome Graphically



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Centrality Property

Definition

$\mathbf{a} \in \mathbb{R}_+^n$ has the centrality property (or is a centrality action profile) if $\mathbf{a} \neq \mathbf{0}$ and

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- Fixed-point definition of actions.

Agents taking high actions are those who benefit a lot (at the margin) from others who are taking high actions.

The Main Theorem

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Theorem

A nonzero \mathbf{a} is a Lindahl outcome if and only if it has the centrality profile.

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will show \Rightarrow . Take $\mathbf{a} \in \mathbb{R}_+^n \setminus \{\mathbf{0}\}$ s.t. $\mathbf{a} = \mathbf{B}(\mathbf{a})\mathbf{a}$.

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Selecting an Outcome: A Bargaining Game

Dávila, Eeckhout, and Martinelli (JPET 09), Penta (JME 11); see also Yildiz (Games 03).

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Theorem

If $\mathbf{0}$ is inefficient and utilities are strictly concave, then: in any *efficient perfect equilibrium*, a Lindahl outcome is played.

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Hurwicz selection of Lindahl outcome.

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- Then Lindahl outcomes are always equilibrium outcomes.
To avoid equilibrium selection fight, Lindahl mechanism is the best bet.

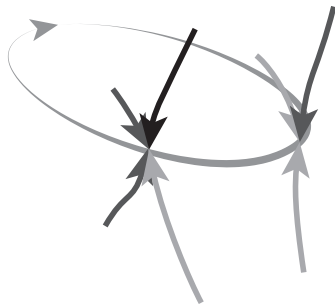
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Walk Interpretation of Eigenvector Centrality

Vague Statement

A node's centrality measures the number/intensity of **walks** in the benefits matrix that end at that node.

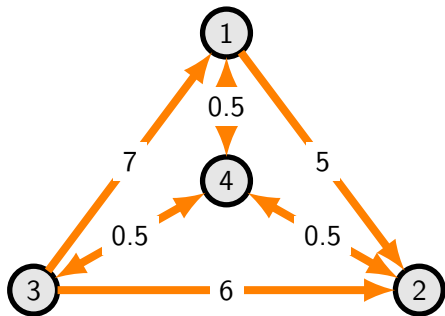


Walks and their Values

$$B(\mathbf{0}) = \begin{bmatrix} 0 & 0 & 7 & 0.5 \\ 5 & 0 & 6 & 0.5 \\ 0 & 0 & 0 & 0.5 \\ 0.5 & 0.5 & 0.5 & 0 \end{bmatrix}$$

Value of walk $w = (3, 1, 2)$:

$$\begin{aligned} v(w; \mathbf{B}) &= B_{13}B_{21} \\ &= 7 \cdot 5 \end{aligned}$$

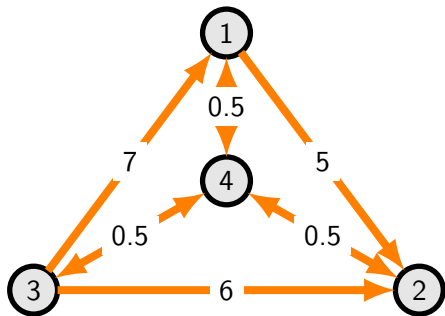


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Walks can repeat nodes: e.g.,
(3, 1, 2, 4, 3, 2).

Centrality in Terms of Walks

Define

$$V_i^\downarrow(\ell; \mathbf{B}) = \sum_{\substack{w \text{ a walk ending at } i \\ \text{of length } \ell}} v(w; \mathbf{B}).$$

Centrality in Terms of Walks

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Fact

Assume $\mathbf{B}(\mathbf{a})$ is aperiodic. \mathbf{a} has the centrality property if and only if

$$\frac{a_i}{a_j} = \lim_{\ell \rightarrow \infty} \frac{V_i^\downarrow(\ell; \mathbf{B})}{V_j^\downarrow(\ell; \mathbf{B})}.$$

Each agent's effort proportional to the total value of long walks he terminates ("total incoming benefits").

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 - Encouraging metaphor, but need to address “markets you can take literally”.

Outline

- 1 Setup
- 2 Efficiency
- 3 Lindahl Outcomes and Network Centrality
- 4 Conclusions**

Further Results

- Analogous characterization with transferable numeraire.
[▶ Details](#)
- Explicit formulas for centrality action profiles in parameterized economies. (New microfoundations for network centrality measures). [▶ Details](#)
- **Next step:** analogous exercise for Walrasian outcomes in other settings to examine key nodes, robustness of market to removing nodes, etc.

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 - No. Has many inefficient equilibria.

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Mechanism H satisfies **payoff-uniqueness** under \mathbf{u} if all elements of $\Sigma_H^*(\mathbf{u})$ are payoff-equivalent (no selection conflict).

Payoff-uniqueness is achievable exactly for those \mathbf{u} such that all Lindahl outcomes under \mathbf{u} are payoff-equivalent. [▶ Proof of theorem](#)

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[▶ Details](#)

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 - Recent applications: Brin and Page (1998); Ballester, Calvó-Armengol, and Zenou (2006); Acemoglu et al. (2012).

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Proof of Cycles Formula for Spectral Radius

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The Spectral Radius as a Measure of Inefficiency: Frictions

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Write $\tau = 1 + t$ (where t is a tax). A tax of $t = r(\mathbf{B}(\mathbf{a})) - 1$ on contributions would be necessary to dissuade a social planner from increasing contributions.

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Proposition

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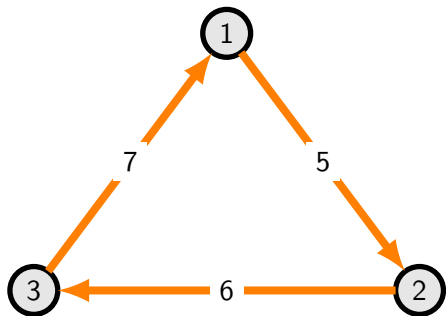
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- By uniqueness of the Perron vector, there is no other egalitarian direction.

Cycles Interpretation

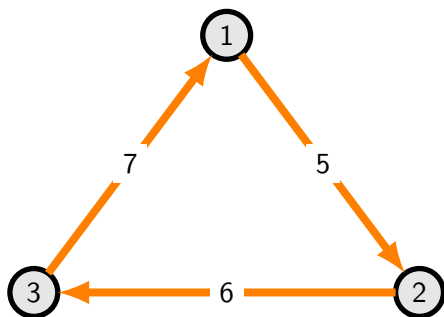
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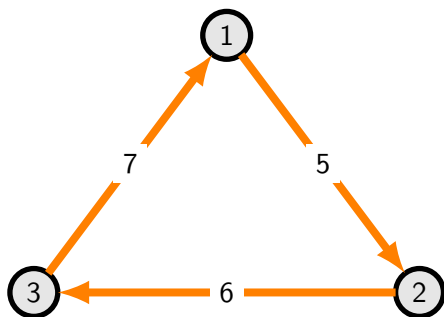


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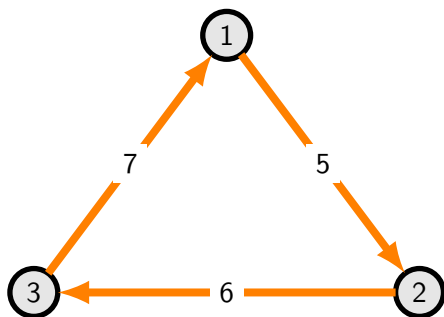


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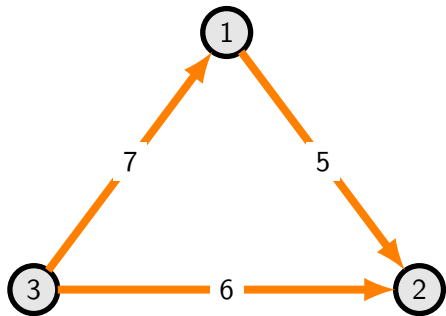
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- Geometric mean of weights along a cycle is always a lower bound on $r(\mathbf{B}(\mathbf{0}))$.
- Cycles also provide an upper bound. If no cycles, then $r(\mathbf{B}(\mathbf{0})) = 0$.

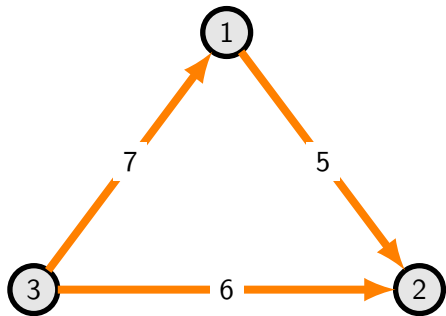


Who is Essential?



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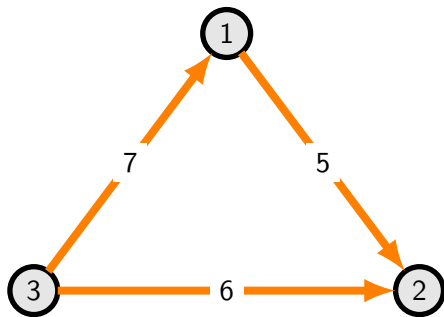


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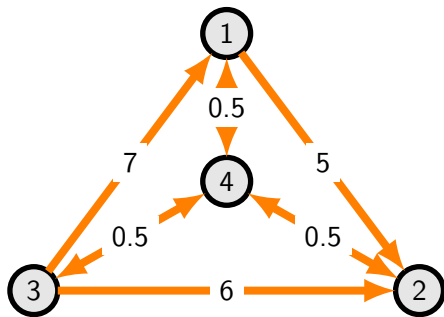
$$r(\mathbf{B}(\mathbf{0})) = 0$$

(no cycles)



Who is Essential?

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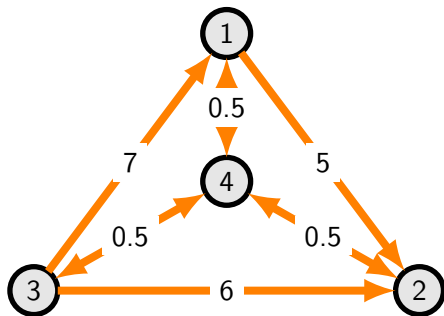


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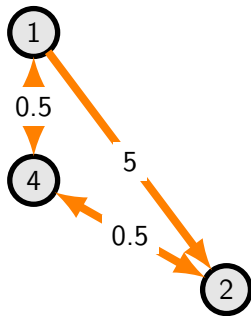
$$r(\mathbf{B}(\mathbf{0})) > 1$$

(lots of cycles)



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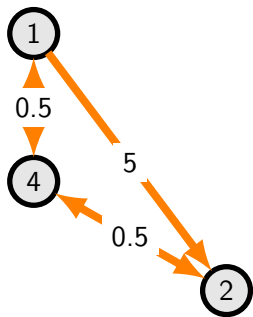
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$$r(\mathbf{B}(\mathbf{0})) \geq (5 \cdot \frac{1}{2} \cdot \frac{1}{2})^{1/3} > 1$$



Gross Substitutes

Assumption (Gross Substitutes)

Let $p_j > 0$ be the price of j 's effort and 1 be i 's wage. Let

$$\mathbf{a}^*(\mathbf{p}) = \operatorname{argmax}_{\mathbf{a}} u_i(\mathbf{a}) \text{ subject to } \sum_{j \neq i} p_j a_j \leq a_i.$$

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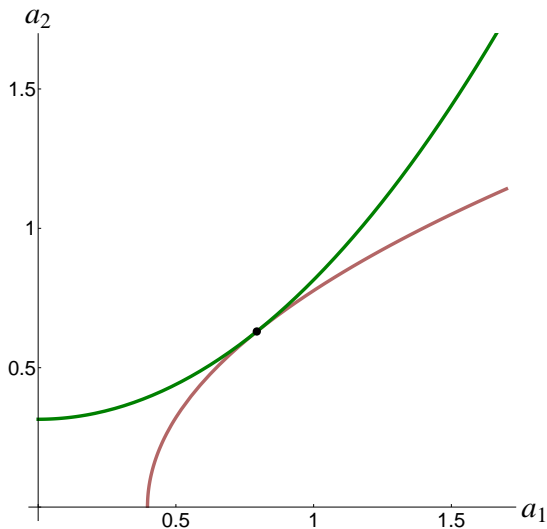
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If only p_j increases, then for $k \neq i, j$, the demand a_k^* does not strictly decrease (in the strong set order); a_i^* does not strictly increase.

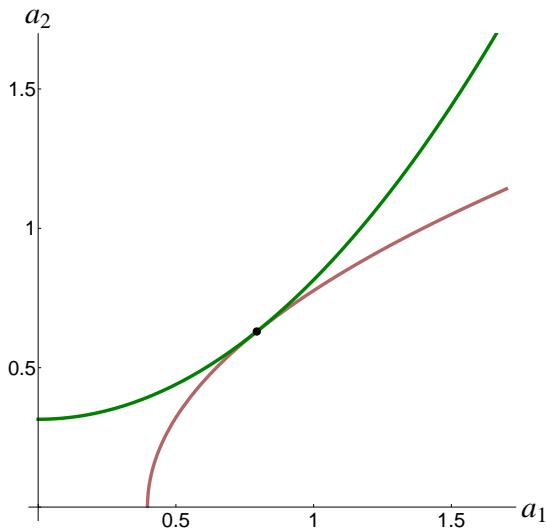
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Consider a Lindahl outcome a under preferences u .



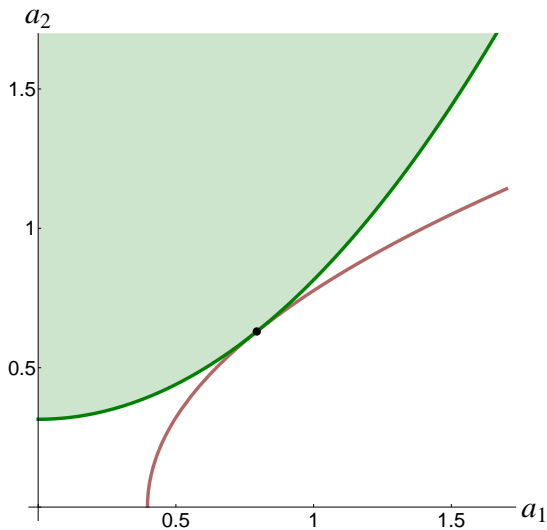
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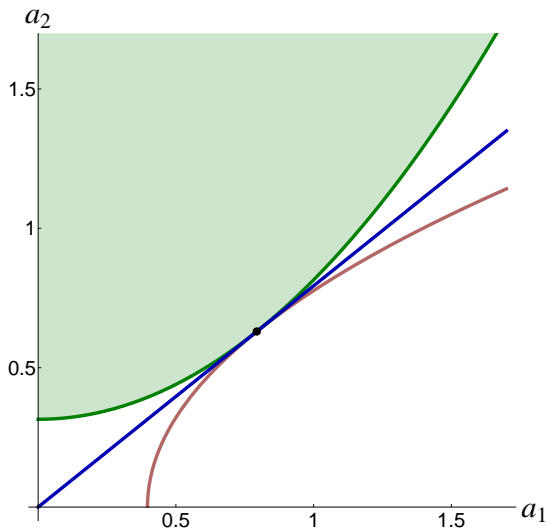
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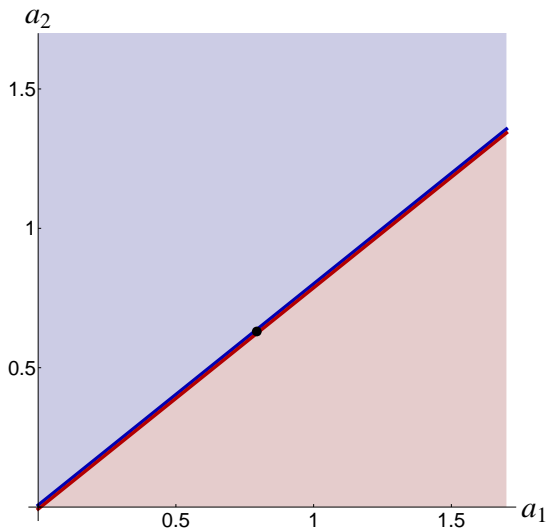
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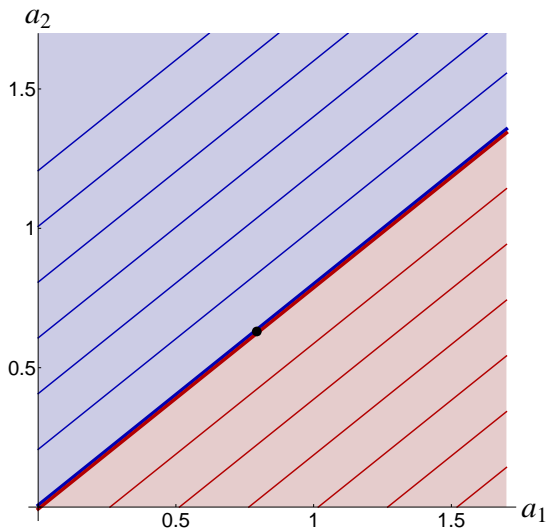
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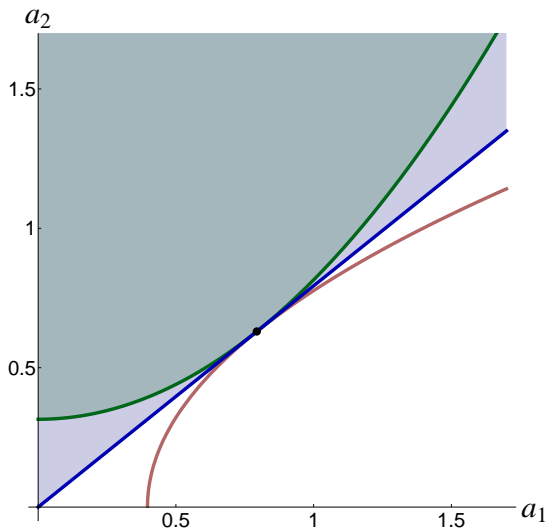
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The Proof that $L \subseteq \Sigma_H^*$ (Hurwicz, Maskin, Postlewaite)

Note that each agent's "better-than- a " set is strictly larger under \hat{u} than under u .

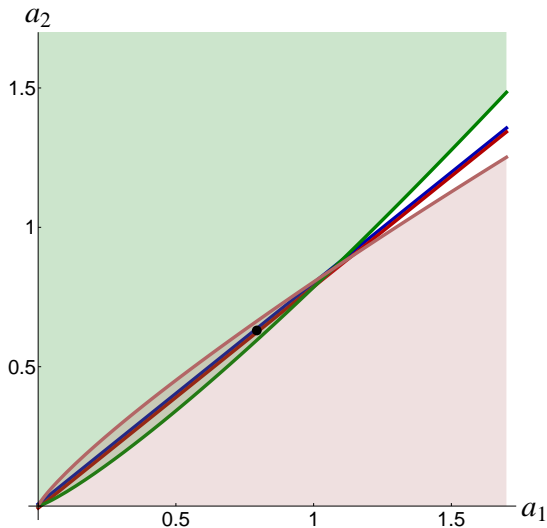
By Maskin's theorem, whatever Σ_H^* implements under \hat{u} must also be implemented under u .



The Proof that $L \subseteq \Sigma_H^*$ (Hurwicz, Maskin, Postlewaite)

Construct preferences increasingly “near” \hat{u} so that IR and PE alone force outcome of Σ_H^* to be near \mathbf{a} .

By continuity, \mathbf{a} must be one of the outcomes implemented under \hat{u} .



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Proposition

The action profile \mathbf{a} is a Lindahl outcome if and only if $\boldsymbol{\theta} = \boldsymbol{\theta} \mathbf{B}$ where $m_i = \theta_i \left(-a_i + \sum_j B_{ij} a_j \right)$.

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Say $\mathbf{h} = \mathbf{1}$. Then $a_i = \binom{\text{total value of walks in } \mathbf{G} \text{ ending at } i}{}$

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- Citations:
 - Yildiz (*Games* '03), Dávila and Eeckhout (*JET* '08), Dávila, Eeckhout, and Martinelli (*J Pub Econ Th* '09), **Penta** (*J Math Econ* '11).