

Fast Convergence of Natural Bargaining Dynamics in Exchange Networks

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joint work with

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Outline

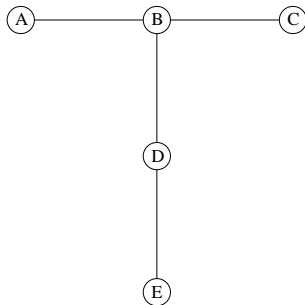
- 1 Exchange networks and Nash bargaining (NB) solns
- 2 Natural dynamics
- 3 Main results

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- 2 Natural dynamics
- 3 Main results

Setting

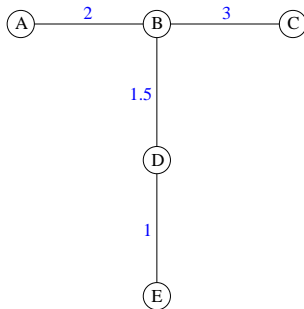
- $G = (V, E)$
- Edge weights $w_{ij} \in (0, W], (ij) \in E$
- $w_{ij} \equiv$ profit from $i - j$ trade
- 1-exchange rule
- Profit from trade split b/w partners



WHAT REALLY HAPPENS?

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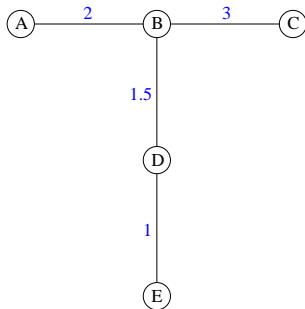
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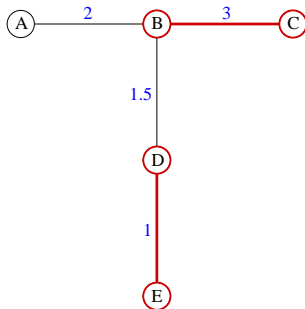
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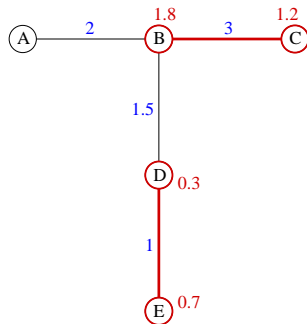
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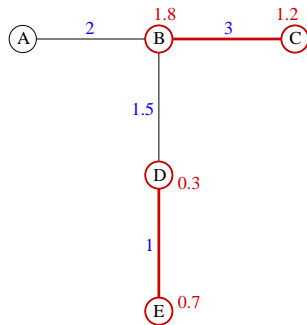
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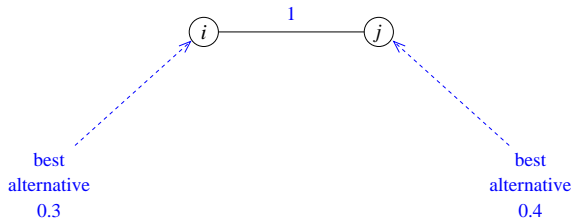
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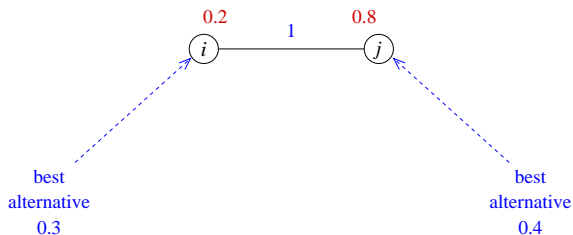


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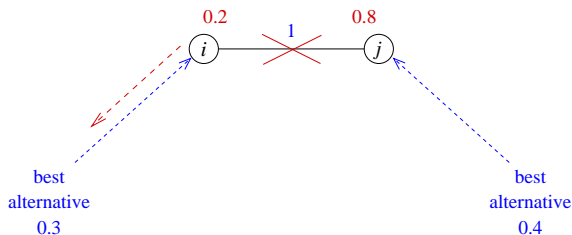
Stability and Balance



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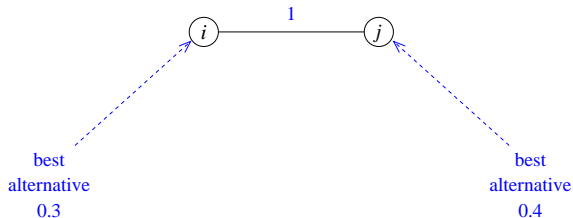


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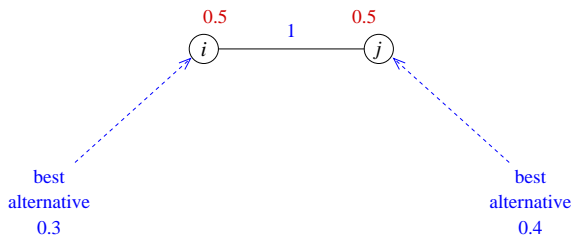


i prefers outside alternative
unstable!

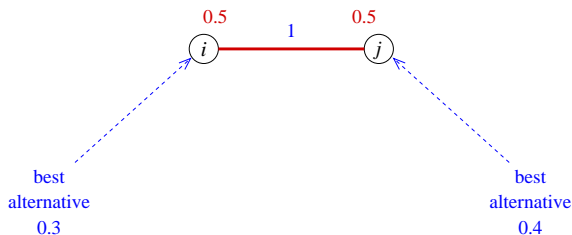
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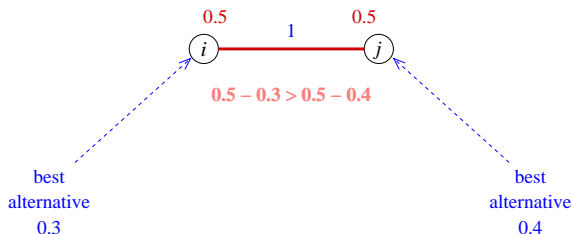


Stability and Balance



Both i and j are happy
i.e. **stable**

Stability and Balance

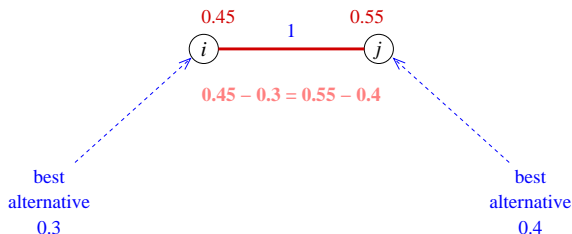


But the allocation lacks **balance**:

$$\text{surplus}(j) = 0.2$$

$$\text{surplus}(i) = 0.1$$

Stability and Balance



both stable and balanced

Nash bargaining (NB) solution for (i, j) pair

Nash bargaining solution on a graph

- $M \equiv$ set of trades, a matching
- $\gamma_i \equiv$ earning of node i

Definition

$(M, \underline{\gamma})$ is a *trade outcome* if:

- For all $(i, j) \in M$, $\gamma_i + \gamma_j = w_{ij}$
- For all unmatched i , $\gamma_i = 0$

NB soln on graph:

Pairwise NB soln for each pair in M , with outside γ 's fixed

Definition

$(M, \underline{\gamma})$ is a **NB soln** if it is a **stable** and **balanced** trade outcome.

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- For $(i,j) \notin M$, implicit $i \rightarrow j$ 'offer': $(w_{ij} - \gamma_i)_+$
- **Stability:** For all $(i,j) \in E$, $\gamma_i + \gamma_j \geq w_{ij}$.
 - Equality on matched edges
 - No pair has incentive to deviate
- **Balance:** For all $(i,j) \in M$,

$$\gamma_i - \max_{k \in \partial i \setminus j} (w_{ik} - \gamma_k)_+ = \gamma_j - \max_{l \in \partial j \setminus i} (w_{jl} - \gamma_l)_+$$

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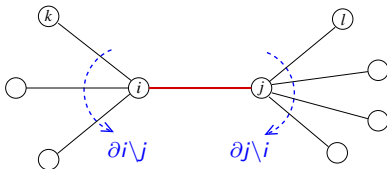
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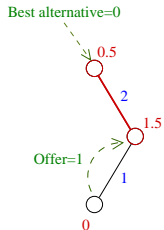
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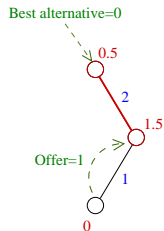
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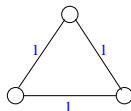
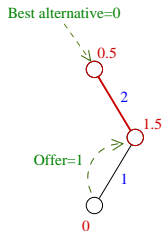


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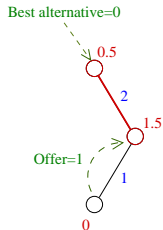
Unique NB soln

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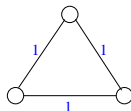


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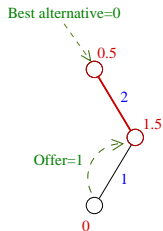


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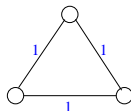


No stable soln

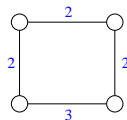
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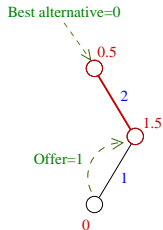
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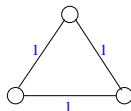
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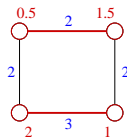
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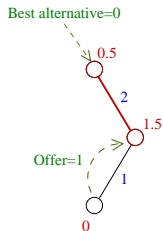
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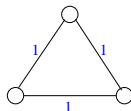
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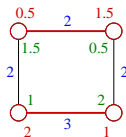
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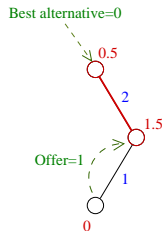
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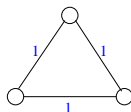
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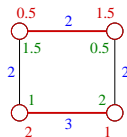
Examples



Unique NB soln



No stable soln



Multiple NB solns

Existence of NB solns

Lemma (Sotomayor '05, Kleinberg & Tardos '08)

Stable solutions correspond to maximum weight matching (MWM).

Theorem (Kleinberg & Tardos STOC '08)

NB soln exists \Leftrightarrow Stable soln exists

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$NB \text{ soln exists} \iff Stable \text{ soln exists} \iff$
 $\begin{matrix} LP \text{ relaxation to} \\ MWM \text{ problem} \\ \text{is tight.} \end{matrix}$

Linear programming (LP) relaxation to MWM problem:

$$\begin{aligned}
 &\text{maximize} && \sum_{(ij) \in E} w_{ij} x_{ij}, \\
 &\text{subject to} && \sum_{j \in \partial i} x_{ij} \leq 1 \quad \forall i \in V, \quad x_{ij} \geq 0 \quad \forall (ij) \in E
 \end{aligned}$$

Construction of NB solutions

Theorem (Kleinberg & Tardos '08)

NB solns can be constructed in polynomial time.

- Global algorithm

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Theorem (Azar, Birnbaum, Celis, Devanur, Peres FOCS '09)

There is a local algorithm that finds an ϵ -Nash bargaining solution in $O(|V|/g) + 2^{O(|V|+|E|)}/\epsilon$ iterations.

- Improved bounds in special cases [Celis, Devanur, Peres WINE '10]

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Theorem (this paper)

There is a local bargaining process that converges to an ϵ -Nash bargaining solution in $O(|V|^4/g^2 + 1/\epsilon^2)$ iterations.

- Process is a **natural model** for agent behavior.

Outline

- 1 Exchange networks and Nash bargaining (NB) solns
- 2 Natural dynamics
- 3 Main results

A natural dynamics

- Nodes exchange information/messages *on graph*
- $\alpha_{i \setminus j}^t$: Current estimated “**best alternative**” of i to j
- $\underline{\alpha}^t$ constitutes network ‘state’.
- $m_{i \rightarrow j}^t$: “**Offer**” from i to j determined as

$$m_{i \rightarrow j}^t = (w_{ij} - \alpha_{i \setminus j}^t)_+ ?$$

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$$m_{i \rightarrow j}^t = (w_{ij} - \alpha_{i \setminus j}^t)_+ - \frac{1}{2}(w_{ij} - \alpha_{i \setminus j}^t - \alpha_{j \setminus i}^t)_+$$

A natural dynamics

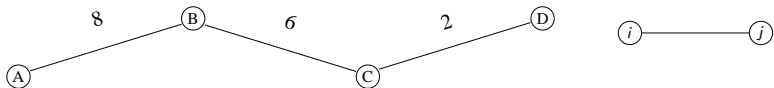
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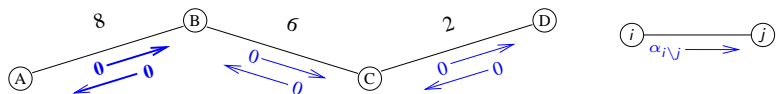
Update rule (synchronous):

$$\alpha_{i \setminus j}^{t+1} = \max_{k \in \partial i \setminus j} m_{k \rightarrow i}^t$$

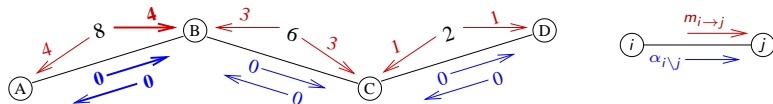
Example of update under natural dynamics



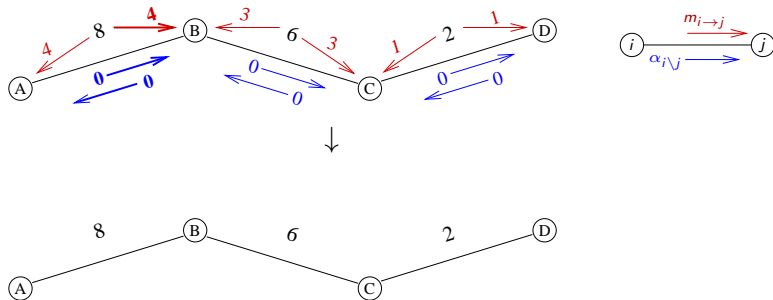
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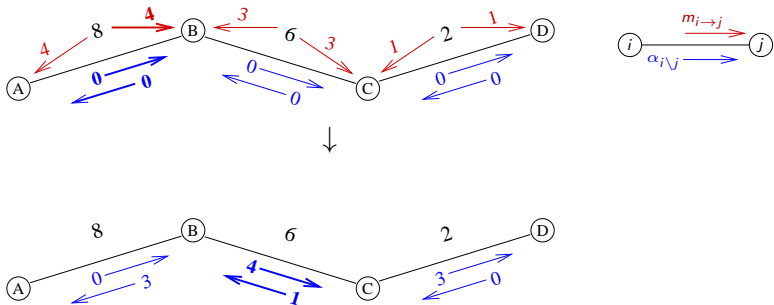
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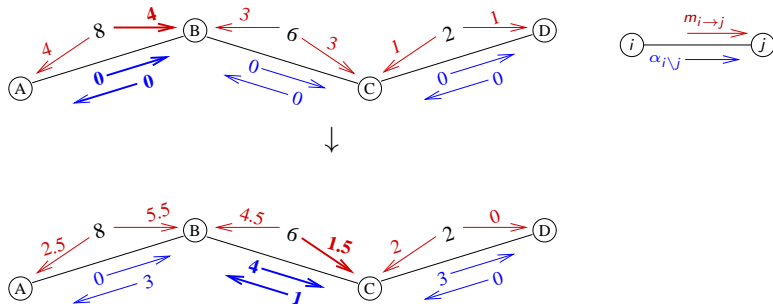
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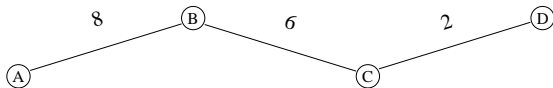
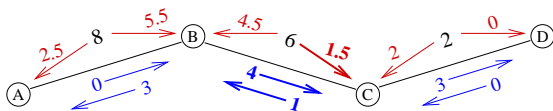
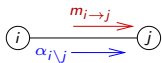
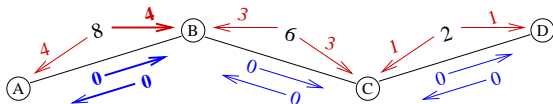
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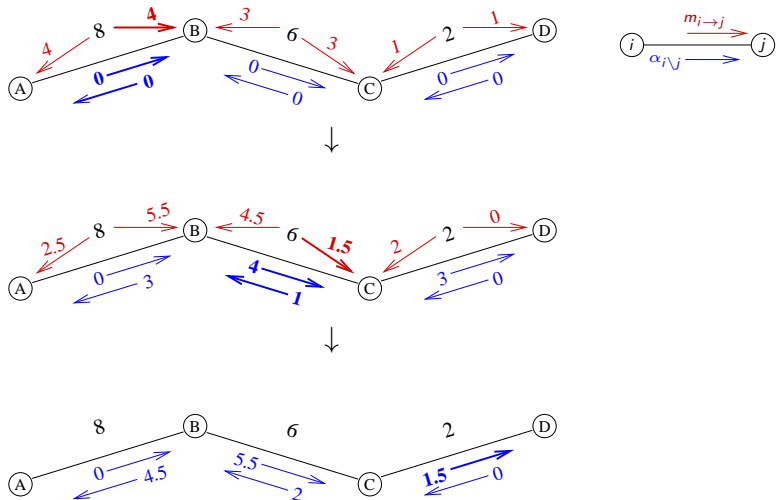
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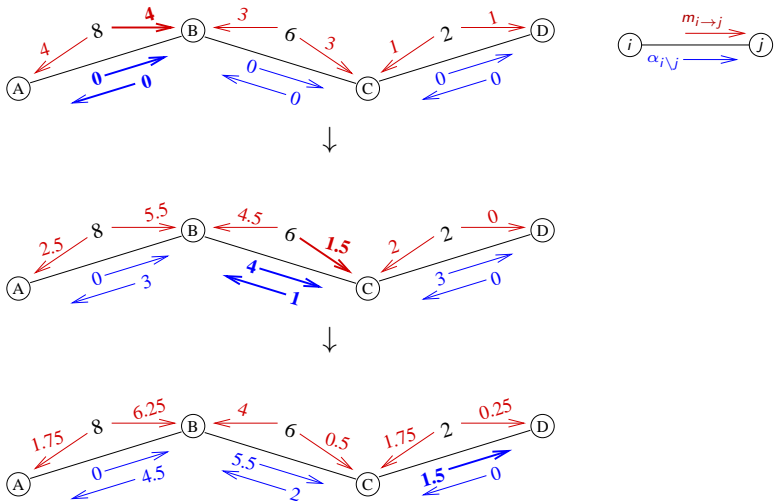
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Natural dynamics (contd.)

- Oscillations on even cycles
- Hence, damping introduced in updates

$$\underline{\alpha}^{t+1} = \kappa \mathsf{T}\underline{\alpha}^t + (1 - \kappa) \underline{\alpha}^t$$

$$(\mathsf{T}\underline{\alpha}^t)_{i \setminus j} \equiv \max_{k \in \partial i \setminus j} m_{k \rightarrow i}^t,$$

$$\kappa \in (0, 1)$$

- Interpretation as reluctance/inertia towards new info
- γ_i^t : Current earning estimate of i

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What can we hope for?

We want:

- Fast convergence
- Fixed points give NB solutions

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Short summary

- $FP \leftrightarrow NB$ solns
- Dynamics converges rapidly to ϵ -FP
- ϵ -FP corresponds to (6ϵ) -NB soln for small ϵ

Implies **fast convergence to ϵ -NB soln.**

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Definitions

Definition (Fixed point)

$$FP \text{ if } \underline{\alpha} = T\underline{\alpha}$$

where $(T\underline{\alpha})_{i \setminus j} \equiv \max_{k \in \partial i \setminus j} m_{k \rightarrow i}$.

Definition (ϵ -Fixed point)

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- Pairs in M receive unique (non-zero) best offers from each other.
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Main results: Fixed points

FPs \leftrightarrow NB solns:

Theorem

Suppose LP-MWM is tight with unique optimum M^ . Then*

- *$\underline{\alpha}$ is FP $\Rightarrow \underline{\alpha}$ induces M^* and $(M^*, \underline{\gamma})$ is NB soln.*
- *$(M, \underline{\gamma}_{\text{NB}})$ is NB soln \Rightarrow (i) $M = M^*$, (ii) unique FP with earnings $\underline{\gamma}_{\text{NB}}$*

Note:

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Assume $w_{ij} \leq 1$ for all $(ij) \in E$. Take any initialization, any $\epsilon > 0$. Define

$$\tau^*(\epsilon) = \frac{1}{\pi\kappa(1-\kappa)\epsilon^2}$$

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Definition of Approximate NB soln

Definition (ϵ -NB soln)

$(M, \underline{\gamma})$ is ϵ -NB soln if it is:

- A valid trade outcome
- Stable
- ϵ -balanced, i.e., for every $(i, j) \in M$

$$\left| \left[\gamma_i - \max_{k \in \partial i \setminus j} (w_{ik} - \gamma_k)_+ \right] - \left[\gamma_j - \max_{l \in \partial j \setminus i} (w_{jl} - \gamma_l)_+ \right] \right| \leq \epsilon$$

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Approximate fixed points

ϵ -FP \rightarrow (6ϵ) -NB soln:

Theorem

Suppose LP-MWM is tight with unique optimum M^ , and gap $g > 0$.*

Let $(\underline{\alpha}, \underline{m}, \underline{\gamma})$ be an ϵ -FP for $\epsilon < g/(6|V|^2)$. Then

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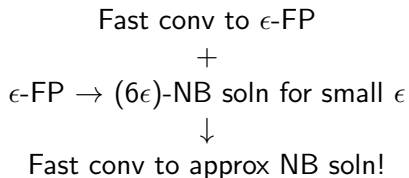
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Theorem (Fast convergence to ϵ -NB)

Assume $w_{ij} \leq 1$ for all $(ij) \in E$. Suppose LP-MWM is tight with unique optimum M^* , and gap $g > 0$. For appropriately chosen $C = C(\kappa)$, define

$$\tau^*(\epsilon) = C \max(|V|^4/g^2, 1/\epsilon^2)$$

For any $t \geq \tau^*(\epsilon)$ we have

- $\underline{\alpha}^t$ induces matching M^*
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Can any local algorithm be faster?

ϵ -NB solution = M^* + balanced allocation

- Extensive work on local algorithms for M^*
- Standard algorithms — max-product BP, auction alg.[Bertsekas '88]
 $O(|V|/g)$ iterations for convergence, tight in worst case
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Example: Bipartite graphs

- NB soln always exists
- We show gap g is large whp under small **random perturbations**
- 'True' polynomial convergence

Proof sketch: Convergence

Natural dynamics is $\underline{\alpha}^{t+1} = \kappa T\underline{\alpha}^t + (1 - \kappa) \underline{\alpha}^t$

Lemma

T is **non-expansive** in sup-norm, i.e.,

$$\|T\underline{\alpha} - T\underline{\beta}\|_{\infty} \leq \|\underline{\alpha} - \underline{\beta}\|_{\infty}$$

- [Ishikawa '76] implies convergence to FP.
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Lots more work to characterize FPs!

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- Nodes may have different bargaining power
- No obvious generalization of KT algorithm.
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