

Bilateral and Multilateral Exchanges for Peer-Assisted Content Distribution

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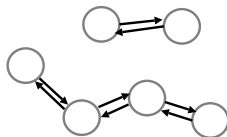
*Joint work with Ramesh Johari (Stanford)
and Michael J. Freedman (Princeton)*

Peer-assisted content distribution

- Users upload files to each other
- Work well only if users share files and upload capacity
- P2P systems try to incentivize users to share

Bilateral and multilateral exchange

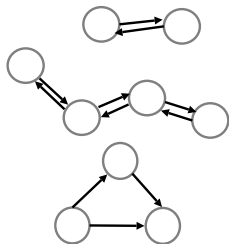
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*Drawback of Bilateral Exchange:
only works between users that have
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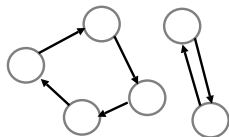
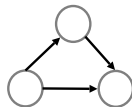
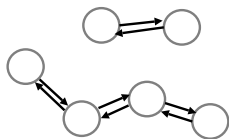


Bilateral and multilateral exchange

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Multilateral exchange allows users to trade in more general ways but is more complex to implement (e.g., virtual currency)



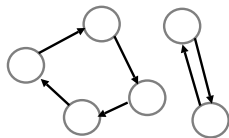
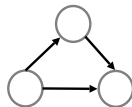
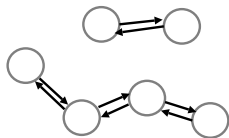
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Tradeoff: simplicity vs. participation



Bilateral vs. multilateral

- 1** Comparison of equilibria
What are the efficiency properties of the allocations that arise at equilibria?
- 2** Quantitative comparison
What proportion of users cannot participate?

Preliminaries

- 1** View content exchange as an economy:
Demand = download requests for content
Supply = scarce system resources
- 2** What files do peers have?
We focus on exchange on a timescale over which the set of files peers have remains constant.
- 3** Rates vs. bytes
We focus on download/upload rates, rather than total number of bytes transferred.
- 4** The network
In the model we study, the constraint is on upload capacity. More generally, a network structure may constrain uploads and downloads.

Notation



r_{ijf} = upload rate of file f from i to j

$d_{if} = \sum_j r_{jif}$ = download rate of f for peer i

$u_i = \sum_{j,f} r_{ijf}$ = upload rate of peer i

$v_i(\mathbf{d}_i, u_i)$ = utility to peer i from (\mathbf{d}_i, u_i)

B_i = bandwidth constraint of user i

\mathcal{X} = set of feasible rate vectors

$\mathcal{X} = \{\mathbf{r} : \mathbf{r} \geq 0; u_i \leq B_i; r_{ijf} = 0 \text{ if } i \text{ does not have file } f\}$

Bilateral content exchange

- Peers exchange content on a pairwise basis
- Let $R_{ij} = \sum_f r_{ijf}$ = rate of upload from i to j
- Exchange ratio: $\gamma_{ij} = R_{ji}/R_{ij}$
- As if there exist prices p_{ij} , p_{ji} ,
and all exchange is settlement-free: $p_{ij}R_{ij} = p_{ji}R_{ji}$
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Example: BitTorrent

- Peer j splits upload rate B_j equally among k_j peers with highest rates to j (the “active set”)
- For a peer i in the active set: $\gamma_{ij} = \frac{B_j}{k_j R_{ij}}$

Multilateral content exchange

- Users can trade a virtual currency, where downloading from peer j costs p_j per unit rate
- Similar to an *exchange economy*

In multilateral exchange,
users optimize given *prices*

Multilateral optimization

$$\begin{aligned} \max \quad & v_i(\mathbf{d}_i, u_i) \\ \text{s.t.} \quad & \sum_j p_j R_{ji} \leq p_i \sum_j R_{ij} \\ & \mathbf{r} \in \mathcal{X} \end{aligned}$$

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In bilateral exchange, users
optimize given *exchange ratios*

Bilateral optimization

$$\begin{aligned} \max \quad & v_i(\mathbf{d}_i, u_i) \\ \text{s.t.} \quad & R_{ji} \leq \gamma_{ij} R_{ij} \forall j \\ & \mathbf{r} \in \mathcal{X} \end{aligned}$$

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At an equilibrium all users have optimized, and the market clears

Multilateral equilibrium (ME)

\mathbf{r}^* and prices \mathbf{p}^*

Bilateral equilibrium (BE)

\mathbf{r}^* and exchange ratios γ^*

Under mild conditions, both ME and BE exist

Pareto efficiency

An allocation \mathbf{r} is Pareto efficient if:
no user's utility can be strictly improved
without strictly reducing another user's utility

- ME are always Pareto efficient
(First fundamental theorem of welfare economics)
- BE may not be Pareto efficient

Pareto efficiency

When are BE efficient?

Theorem

Assume utility approaches $-\infty$ as upload rate approaches capacity
A BE (γ^*, \mathbf{r}^*) is Pareto efficient if and only if there exists a supporting vector of prices \mathbf{p}^* such that $(\mathbf{p}^*, \mathbf{r}^*)$ is a ME

[Hard part to prove is the “only if”]

Pareto efficiency: proof sketch

Given Pareto efficient BE (γ^*, \mathbf{r}^*)

find price vector \mathbf{p}^* such that $(\mathbf{p}^*, \mathbf{r}^*)$ is a ME

- The proof exploits a connection between equilibria and reversible Markov chains
- Let R_{ij}^* = total rate from i to j at BE and $R_{ii}^* = -\sum_j R_{ij}^*$
- For simplicity, suppose \mathbf{R}^* is an irreducible rate matrix of a continuous time MC (generalizes to nonirreducible case)

Pareto efficiency: proof sketch

Let \mathbf{p} be the unique invariant distribution of \mathbf{R}^*
If \mathbf{R}^* is reversible, then:

$$p_i R_{ij}^* = p_j R_{ji}^* \Rightarrow \gamma_{ij}^* = p_i / p_j \Rightarrow BE \equiv ME$$

What is the intuition for this result?

- The invariant distribution gives a vector of prices at which agents could potentially trade
- When \mathbf{R}^* is reversible, agents' trades balance on a pairwise basis with one vector of prices

Pareto efficiency: proof sketch

What if \mathbf{R}^* is not reversible? $\frac{p_i}{p_j} > \gamma_{ij}^*$ for some $R_{ij}^* > 0$

$\Rightarrow i$ “overpaid” to transact with j at BE

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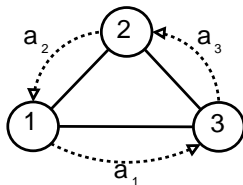
Pareto improvement:

Increase u_i^* and $R_{i,i-1}^*$ by a_i

User i better off if $\frac{a_{i+1}}{a_i} > \gamma_{i,i+1}^*$

Possible to find such a_i 's,

because $\prod_i \gamma_{i,i+1}^* < 1$



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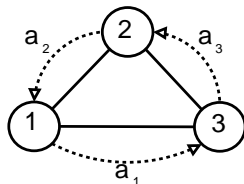
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\mathbf{r}^* Pareto efficient BE $\Rightarrow \mathbf{R}^*$ reversible

Pareto efficiency: proof sketch

- Of course, in general \mathbf{R}^* may not be irreducible
Instead, the graph of trades in the BE may have multiple connected components
- To complete the proof, we consider supporting price vectors \mathbf{p} that arise as linear combinations of the unique invariant distributions on each component
- We show that if no supporting prices for the BE exist, then a cycle of agents (possibly spanning multiple connected components) can be found who have a Pareto improving trade

Bilateral vs. multilateral

1 Pareto efficiency of equilibria

2 Participation

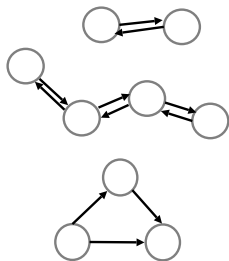
How many peers are able to trade bilaterally and multilaterally?

We use a random model to quantify the density of trade produced by the two models

Participation

Two peers are complementary if each has what the other wants
A peer can trade bilaterally if she has a complementary peer

A peer can trade multilaterally if it belongs on a cycle of peers along which peers want to trade



Participation: asymptotic analysis

- N users, K files
- Each user has one file to upload, and wants to download one file
- The probability a user wants or has the f -th most popular file is proportional to f^{-s} (Zipf's law)
 - $s = 0$: uniform popularity
 - $s > 1$: popularity concentrated in relatively few files
- Metric: expected proportion of users that cannot participate

Participation: asymptotic analysis for $s < 1$

Let ρ_{ME} (resp., ρ_{BE}) be the expected number of unmatched peers in multilateral (resp., bilateral) exchange

Theorem

When $s \in [0, 1)$:

- If $N > K^2$, then $\rho_{BE} \rightarrow 0$
- If $N < K^2$, then $\rho_{BE} \geq (1 - s)^2$
- If $K \log K < N$, then $\rho_{ME} \rightarrow 0$

If N scales faster than $K \log K$ but slower than K^2 , multilateral is significantly better than bilateral

Theorem

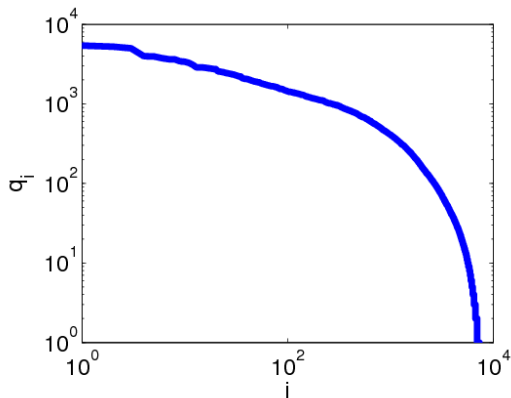
If $s > 1$, then $\rho_{BE} \rightarrow 0$ for any scaling of K and N

- So in this case, bilateral performs very well
- Intuition:
high concentration of popularity in a small number of files
- This result also holds:
 - when peers upload and download multiple files
 - for more general random graph models

BitTorrent popularity data

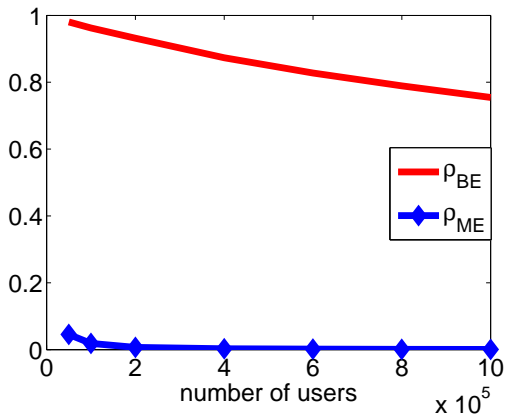
Dataset from [Piatek et al., 2008]

1.4M downloads, 680K peers, 7.3K files



Data-driven comparison

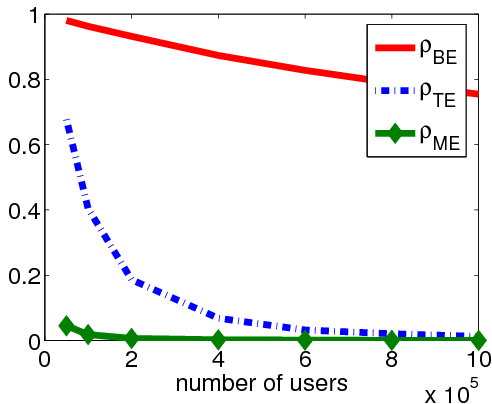
What if we sample a random graph from the BT distribution?



Multilateral exchange matches many more peers than bilateral

Data-driven comparison

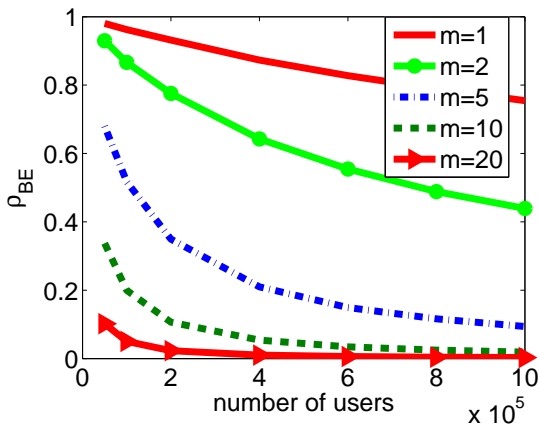
What if users can trade in triangles?



Trilateral exchange converges much faster than bilateral

Data-driven comparison

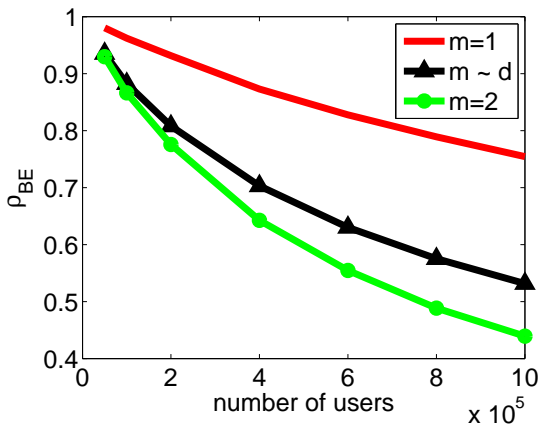
However, as the number of files a peer has increases, bilateral rapidly approaches multilateral



$m = \#$ of files a peer has available for uploading

Data-driven comparison

What if the number of files that users possess varies across different users?



d = distribution
from dataset

- mean = 2.0084

- high variance

Conclusions

- A BE is Pareto efficient if and only if it corresponds to a ME
- Bilateral exchange performs very well in expectation if the file popularity is very concentrated and/or users share a sufficiently large number of files

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Open issue: extend comparison to a dynamic setting, where

- downloads complete and preferences change over time
- users join and leave the system