

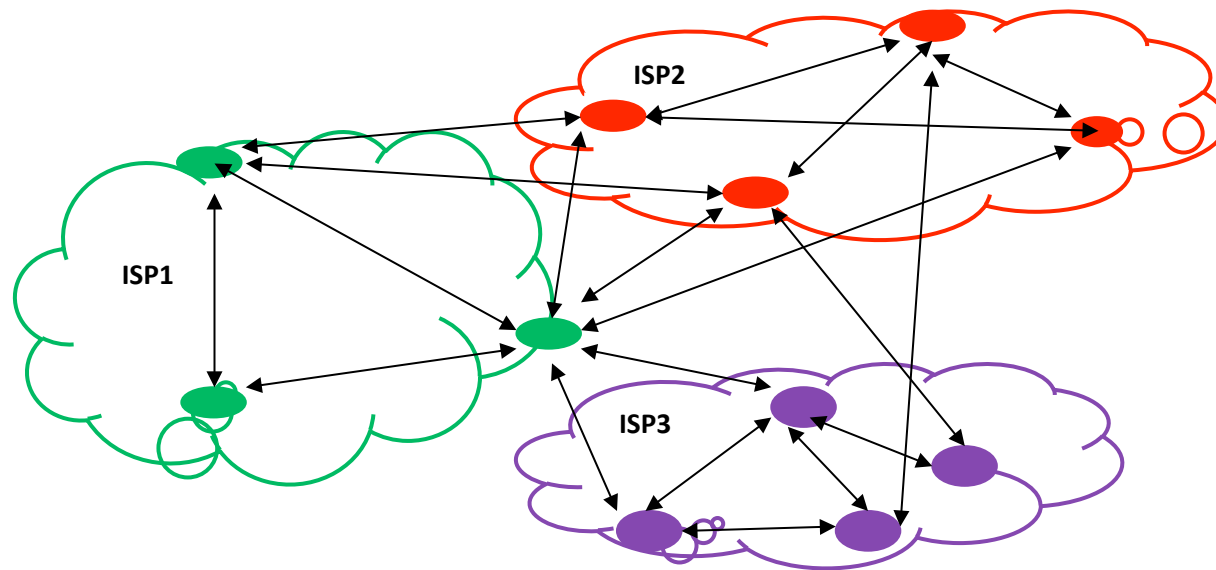
# Designing ISP-Friendly P2P Using Game-based Control



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# Problems with existing P2P

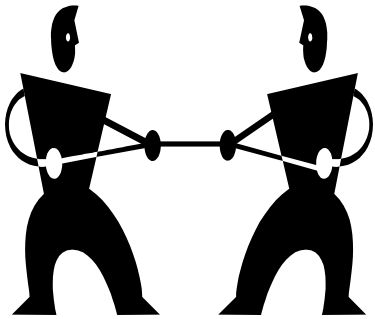
- Oblivious of ISP domains
- Can result in huge data flow across ISP boundaries
- Hence increased cost for an ISP



# Problem Overview

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- We need a P2P system that trades off **transit price** and **delay**



- **Price** is reduced by localizing traffic within an ISP domain
- **Delay** can be reduced by choosing the best peer, irrespective of the ISPs

*Key Question: How to achieve the optimal point?*

# Related Work

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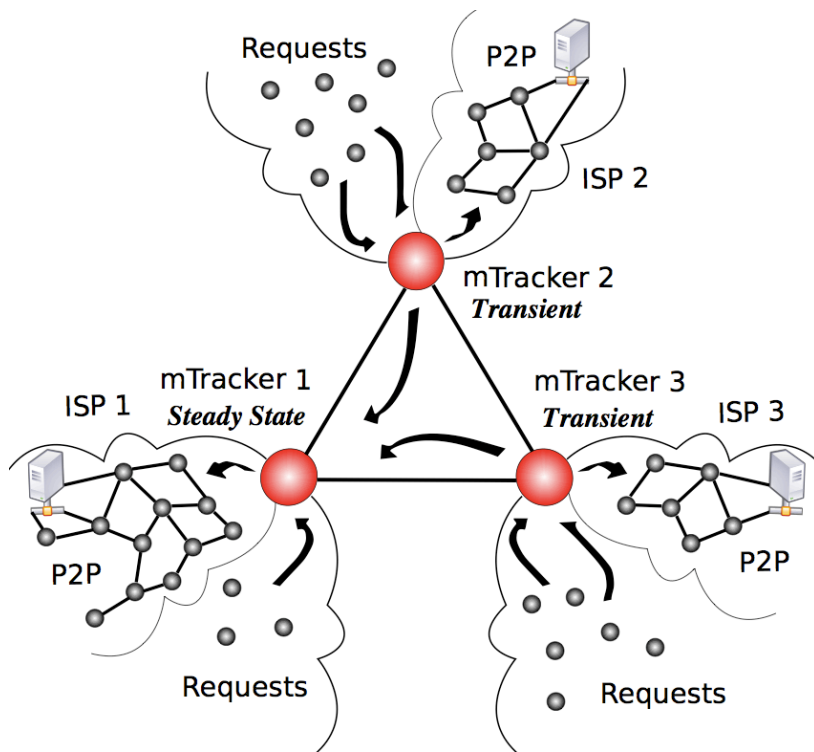
- V. Aggarwal, A. Feldmann, and C. Scheideler, Can ISPs and P2P users cooperate for improved performance?

*ACM Computer Communication Review, 37(3), July 2007.*

- H. Xie, Y. R. Yang, A. Krishnamurthy, Y. Liu, and A. Silberschatz, P4P: Portal for P2P applications.

*In Proc. ACM SIGCOMM, Aug. 2008.*

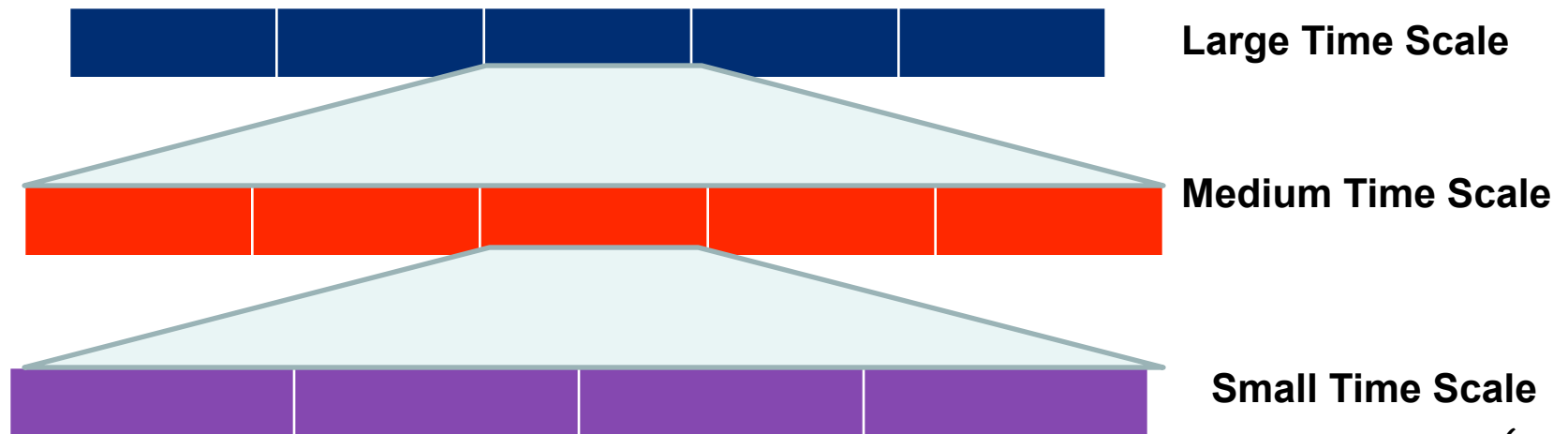
# MultiTrack for BitTorrent-like P2P



- **Steady State** : Load is less than the available capacity
- **Transient State** : Load is more than the available capacity
- **Must split traffic taking into account both delay and cost.**

# Assumptions

- Capacity at mTracker  $i$  (or the peer swarm) is assumed to be  $C^i$  users/time
- New requests arrive at mTracker  $j$  in a Poisson process with parameter  $x_j$  users/time
- Delay is convex increasing in load.



# Population Game

- A population game  $G$ , has  $Q$  non-atomic populations and for each population  $j$ :
  - A mass  $x_j$ ,
  - A strategy set  $S_j = \{1, \dots, S_j\}$
  - A marginal payoff for each strategy  $F_j^i(\mathbf{X}), i \in S_j$  where  $\mathbf{X}$  is the state of the system
- A state  $\mathbf{X}$  (or a strategy distribution) is the way the population is partitioned into the different strategies available,  $\mathbf{X} = \{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_Q\}$

$$\vec{x}_j = \{x_j^1, x_j^2, \dots, x_j^{S_j}\} \rightarrow \text{Sum is exactly } x_j$$

# Dynamics

- Every player follows selfish dynamics, maximizing their own payoff.
- User strategies evolve with time as they adapt to the state.

**Replicator Dynamics:** *Rich become richer and poor become poorer*

$$\dot{x}_j^s = x_j^s \left( \underbrace{F_j^s(\mathbf{X})}_{\text{Payoff per unit}} - \underbrace{\frac{1}{x_j} \sum_{i=1}^{S_j} x_j^i F_j^i(\mathbf{X})}_{\text{Average payoff per unit}} \right)$$



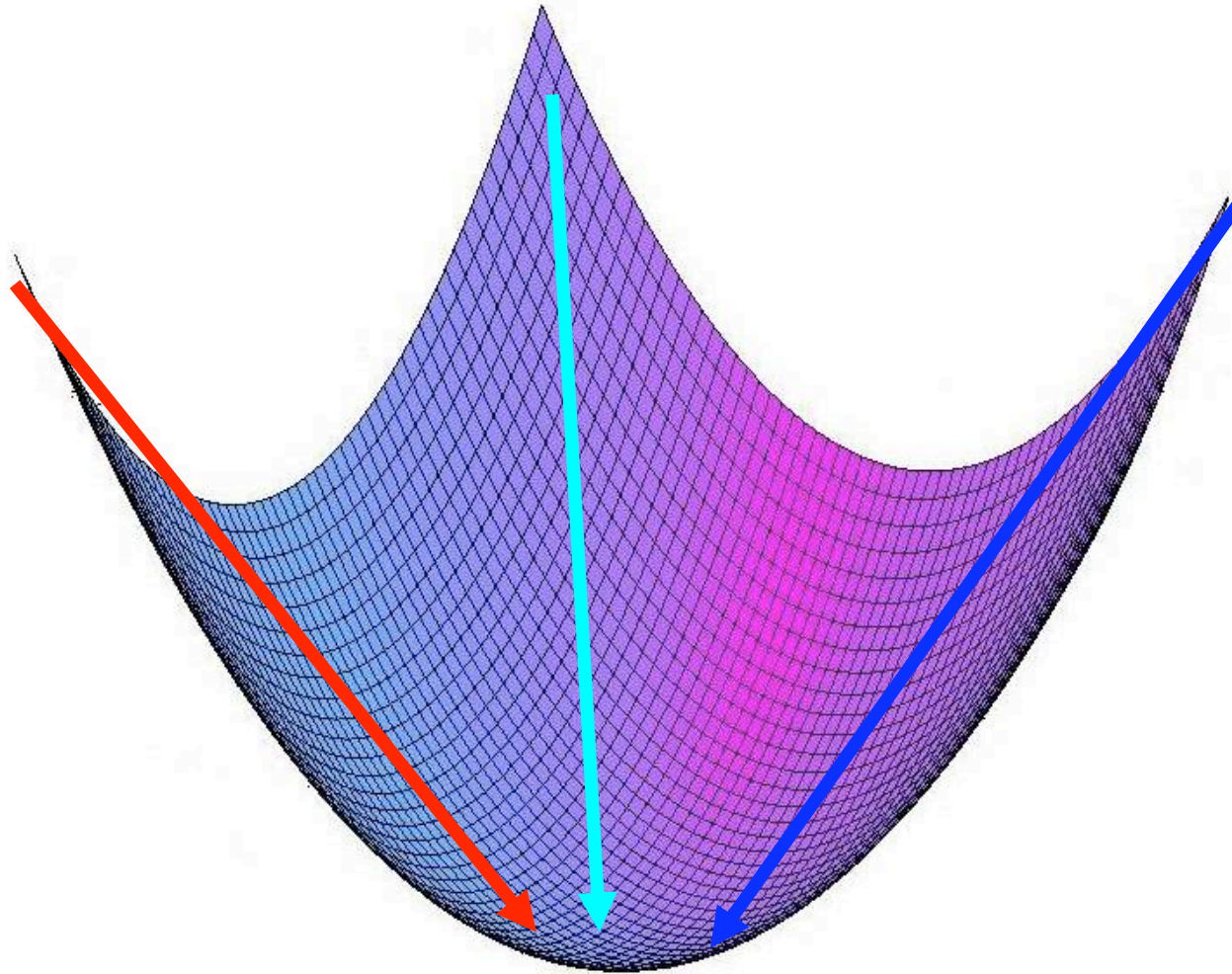
# Marginal Payoff/Cost

- $F_j^i(\mathbf{X})$  represents per unit payoff for mTracker  $j$  in forwarding request to strategy  $i$  in state  $\mathbf{X}$ :
  - Delay at mTracker  $i$
  - Transit cost from mTracker  $j$  to mTracker  $i$
  - Congestion cost at mTracker  $i$

$$F_j^i(\mathbf{X}) = \underbrace{\frac{1}{C^i - \sum_{l=1}^Q x_l^i}}_{\text{Marginal delay}} + \underbrace{p_j^i}_{\text{transit cost}} + \underbrace{\frac{\sum_{l=1}^Q x_l^i}{(C^i - \sum_{l=1}^Q x_l^i)^2}}_{\text{Congestion}}$$

# Lyapunov Function

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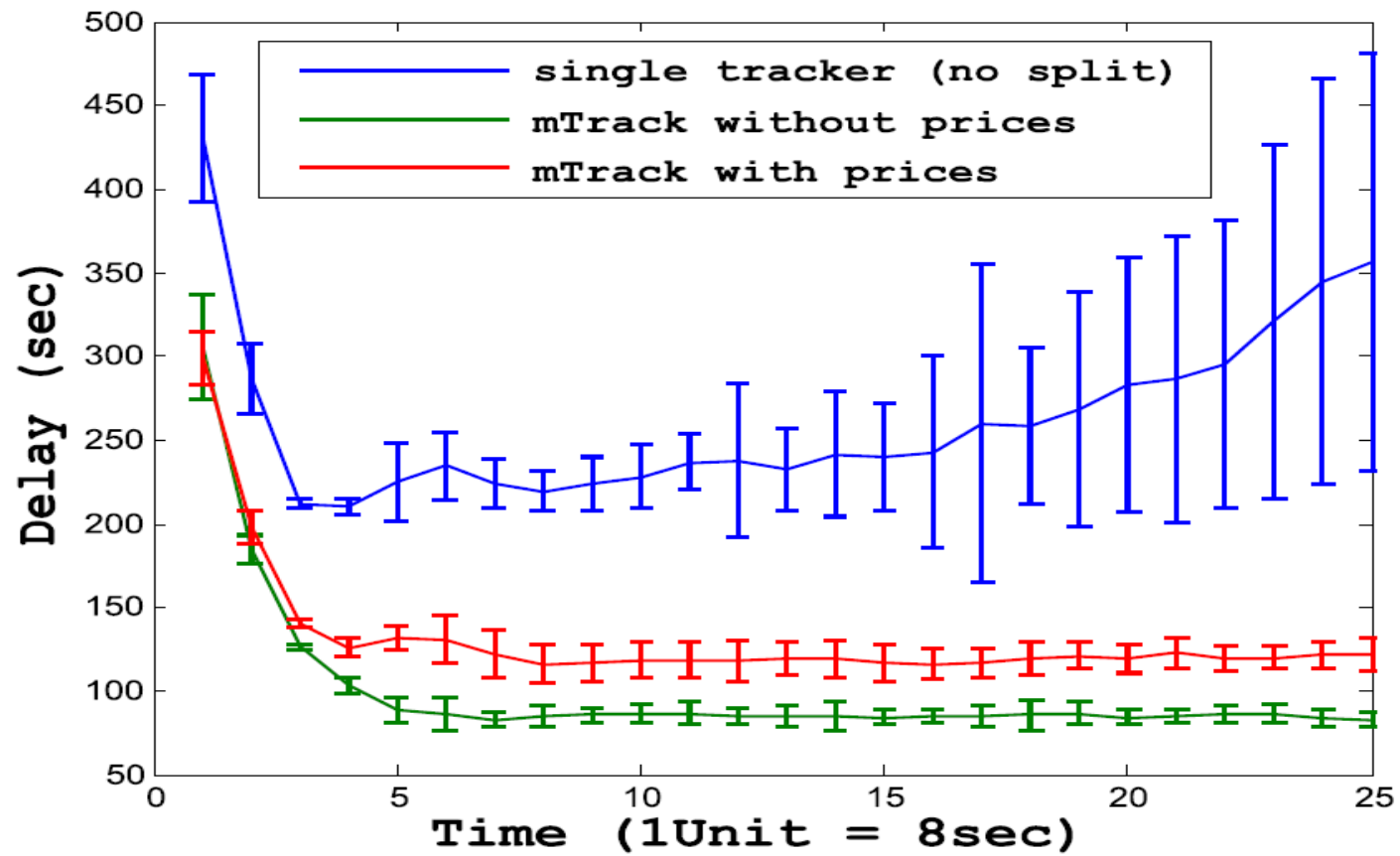
# Total System cost

- The total cost of the system when in state  $\mathbf{X}$  is:

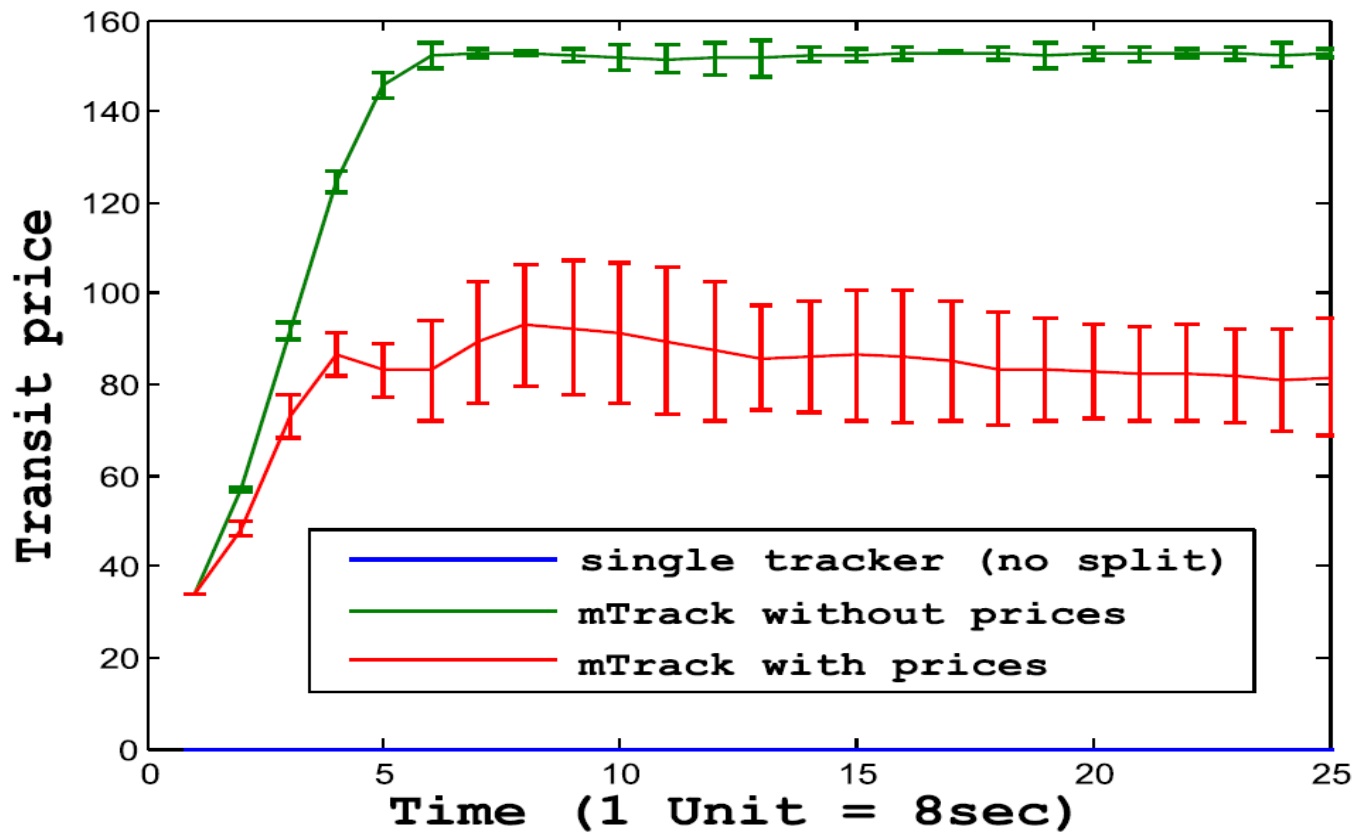
$$\mathcal{C}(\mathbf{X}) = \sum_{i=1}^Q \left\{ \underbrace{\frac{\sum_{r=1}^Q x_r^i}{C^i - \sum_{l=1}^Q x_l^i}}_{\text{Delay}} + \underbrace{\sum_{r=1}^Q p_r^i x_r^i}_{\text{Transit cost}} \right\}$$

- We use  $\mathcal{C}(\mathbf{X})$  as our Lyapunov function
- We prove that the system of *mTrackers* that uses negative replicator dynamics is *globally asymptotically stable*.

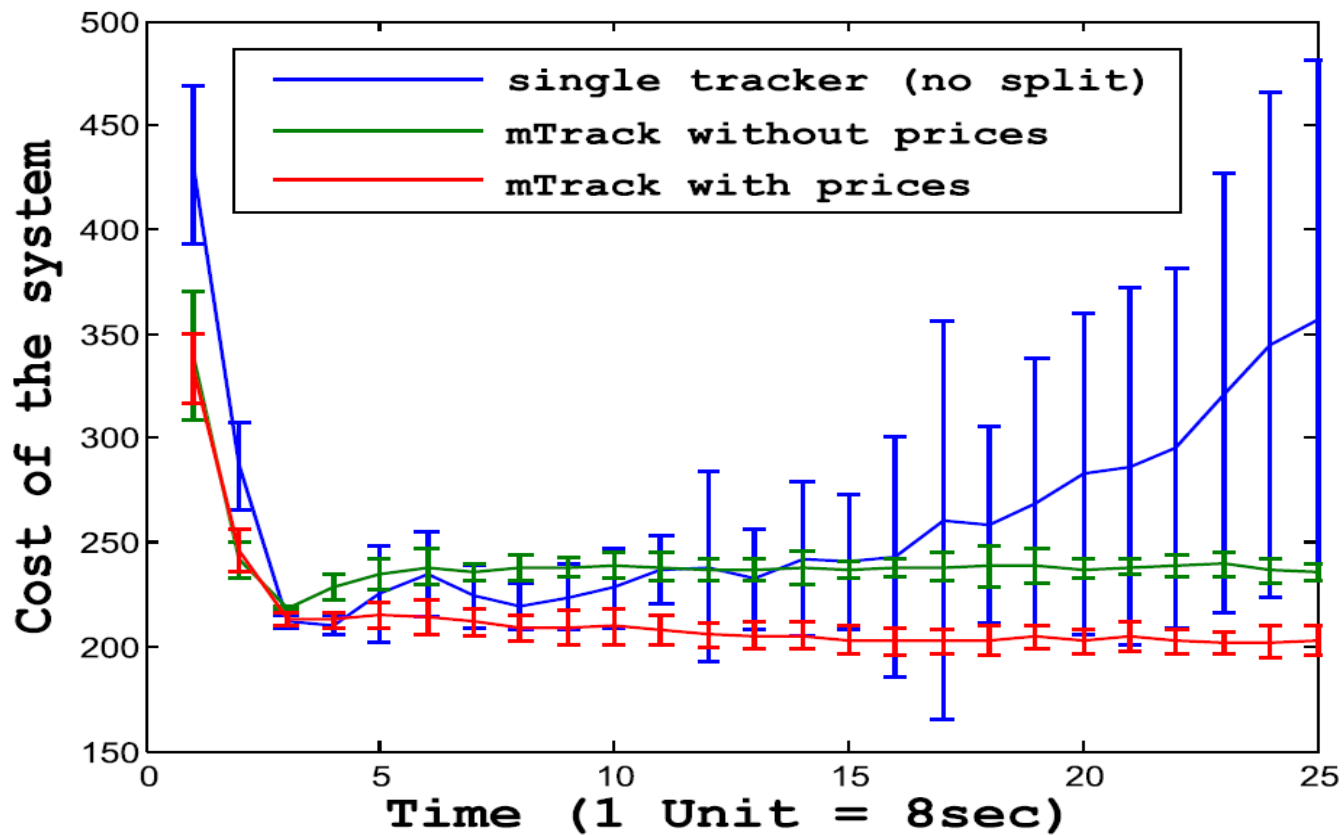
# Delay



# Transit Cost



# Total Cost (Delay + Transit)



# Insights and ongoing work

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- *Key insight:*  
It is possible to align incentives in terms of delay of a P2P user and the transit costs of an ISP.
  
- *Ongoing work:*  
Admission Control.  
Potential testbed.