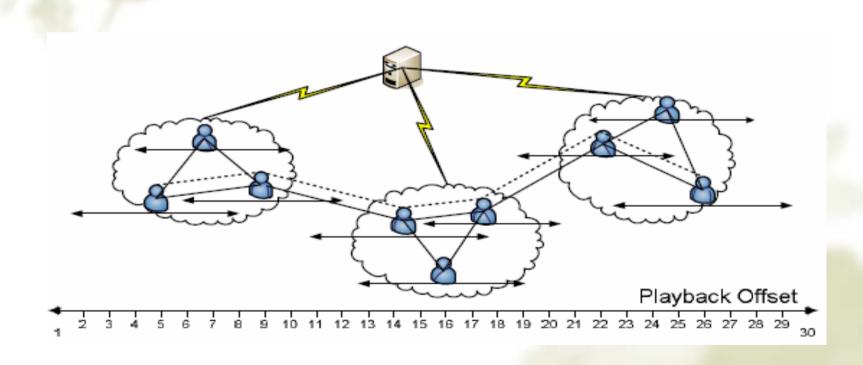


Auction-based incentives and optimal scheduling mechanism design in P2P VoD streaming

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## P2P VoD Streaming





## Challenges

#### Incentive

Peers are selfish. They want to download most while upload least.

#### Scheduling

- How to collaborate: who upload what parts of media data to whom at when?
- Centralized scheduling is impossible.
- Exacerbated by on-demand properties that lower the levels of content overlap among peers



### Possible solutions

#### Incentive

- ◆ Tic for Tac
  - successfully applied in file-sharing application, but poorly in VoD application
- Periodically rebuild the multicast trees
  - increasing the likelihood that a freeloading node's downstream peers will later be upstream of the freeloader and can retaliate by refusing to serve the offender

#### Scheduling

- Rarest first
  - Successfully applied in file-sharing application, but it does not consider the property of streaming
- Hybrid scheme( place certain weights on rareness and deadline )
  - Can not adapt in a dynamic environment



## Our objectives

- Design auction-based mechanism to simultaneously realize two separate goals:
  - Upload incentives
  - Effective block scheduling
- Philosophy:
  - Market can allocate resource optimally
  - Work more, harvest more



## Basic concepts

- An auction is a process of buying and selling goods or services by offering them up for bid, taking bids, and then selling the item to the highest bidder (s).
- \* Bidding is an offer (often competitive) of setting a price one is willing to pay for something.

### P2P VoD Auction Model

- Typical pull-based P2P VoD streaming system
  - Divide each video into many blocks
  - Peers connect to a set of neighbors with similar playback progress
  - Neighbors exchange buffer availability bitmaps (i.e. buffer maps) periodically

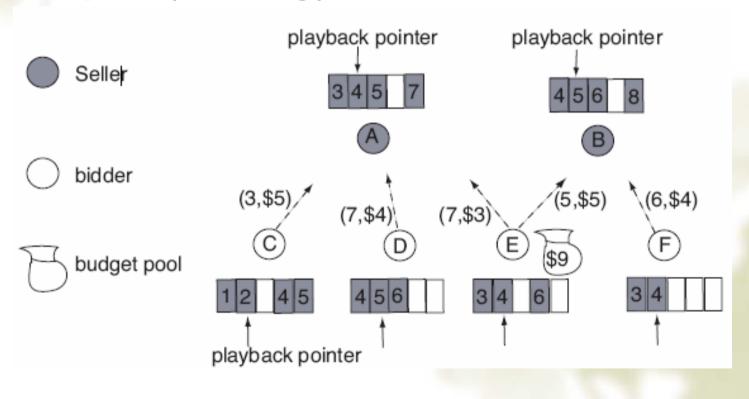
- We model media block exchanges among neighboring peers into a collection of decentralized, locally administrated market hosted by each peer.
  - The goods being auctioned and exchanged in the markets are media blocks.
  - In each market, iterated asynchronous auctions take place.
  - In each auction, the host peer plays the role of seller, selling its buffered media blocks to neighbors who bid for the desired ones out of them.
- Idea: modeling media block instead of modeling media flow makes the model more realistic

- Each peer is furnished with a budget (some kind of virtual currencies).
  - Winner peers in the auctions gain the rights to, download the wined blocks while pays prices out of its budget to the sellers.
- Idea: Incentive:
  - upload blocks
  - -> more budget
  - -> more competitive in bidding for blocks
  - -> enjoy better viewing experience

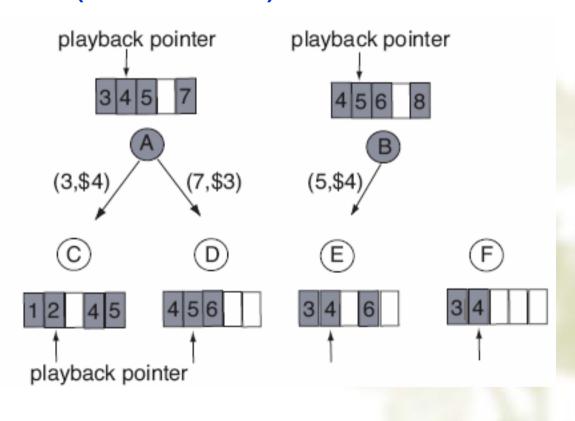
#### Properties:

- Decentralized
- Dynamic: with continuously changing blocks and possibly bidders
- Iterated: along with the streaming process, auctions execute round by round
- Asynchronous

Example: (bidding)



Example: (allocation)





## Mechanism design

- For seller
  - A discriminative second price auction with seller reservation
- For bidder
  - A truthful start with iterative price discovery strategy.



### Mechanism at seller

#### Allocation rule:

Sort received bids by bidding prices, and maximally sell blocks in that order, within the available upload bandwidth. (explain with the example picture)



### Mechanism at seller

- Charging scheme
  - Select the highest bid-independent charge for each winning bid (indicated by the allocation rule discussed above) as the bidding price from the immediately lower bid.
- Properties:
  - → Bid-independent
  - Revenue maximization

### Mechanism at seller

- Seller reservation
  - When the number of bids is larger than its maximum upload capacity; otherwise
  - Purpose: keep the market competitive (market price above 0) without revenue loss

$$o_i = \begin{cases} O_i, & \text{if } m > O_i, \\ m - 1, & \text{if } m \le O_i \end{cases}$$

## Summary of mechanism at seller

#### **Algorithm 1** Protocol at Seller i (in every interval T)

#### (a) Allocation

```
1: receive bids \mathbf{b_i} from neighbors in \mathcal{D}_i
```

2: order  $b_i$  in non-increasing order of bidding prices into list  $l_s$ 

3: set 
$$o_i = O_i$$
 if  $m = |\mathbf{b_i}| > O_i$ ; otherwise set  $o_i = m - 1$ 

4: while  $o_i > 0$  do

5: select next bid 
$$b_{ij}^{(k)} = (I_{ij}^{(k)}, p_{ij}^{(k)})$$
 in list  $l_s$ 

6: let charge  $c_{ij}^{(k)}$  be  $p_{ij'}^{(k')}$ , price in the subsequent bid in  $l_s$ 

7: send charge 
$$c_{ij}^{(k)}$$
 to bidder  $j$ 

8: start transfer of block 
$$I_{ij}^{(k)}$$
 to bidder  $j$ 

9: 
$$o_i \leftarrow o_i - 1$$

10: end while

#### (b) Upon receiving payment from bidder j for block $I_{ij}^{(k)}$

1: update budget,  $e_i \leftarrow e_i + c_{ij}^{(k)}$ 

### Mechanism at bidder

- An optimization problem
- Definition
  - Valuation of a block: a function (over [0,1], differentiable, non-decreasing, and quasi-linear) reflects the urgency level of downloading the block (playback deadline) and rareness level of the block (potential competition and higher resale chance and price)
  - Utility of a block: valuation minus charging price (Notice: the factual charging price is only revealed after the auction is completed)
  - Marginal utility: valuation divided by charging price

### Mechanism at bidder

- Pricing mechanism: truthful start with iterative price discovery
  - First round of bidding at neighbor I, bidding price = true block valuation
  - Subsequent rounds of bidding, estimate the market price according to the results of last round of bidding, bidding price = min (market price estimate, true block valuation).
  - How to estimate market price:
    - ❖ If there are successful bids in the last round
    - ❖ If all bids fail



### Mechanism at bidder

- Bidding strategy:
  - 1. Decide bidding price according to the "Truthful start with iterative price discovery" strategy
  - 2. Resolve an integer program which maximization the overall utility gained under the constraints:
    - ♦ (1) budget constraint
    - (2) for an identical block, only request from a neighbor at a round

### Mechanism at bidder

Maximize 
$$\sum_{i \in \mathcal{D}_j} \sum_{k \in \mathcal{K}_{ij}} (v_{ij}^{(k)}(x_{ij}^{(k)}) - p_{ij}^{(k)}x_{ij}^{(k)})$$
 (5)

Subject to:

$$\begin{cases}
\sum_{i \in \mathcal{D}_j} \sum_{k \in \mathcal{K}_{ij}} p_{ij}^{(k)} x_{ij}^{(k)} \leq e_j & (6) \\
\sum_{i \in \mathcal{D}_j} x_{ij}^{(k)} = z_j^k & \forall k \in \mathcal{K}_{ij} & (7) \\
x_{ij}^{(k)}, z_j^k \in \{0, 1\} & \forall i \in \mathcal{D}_j, \forall k \in \mathcal{K}_{ij} & (8)
\end{cases}$$



### Mechanism at bidder

#### Simplification:

- Compute the price the bidder is willing to pay for each block
- Select blocks with the highest marginal utilities to bid for

## Summary of mechanism at bidder

#### Algorithm 2

#### Initialization

1: set  $\ddot{q}_{ij}$  to a MAX value,  $\forall i \in \mathcal{D}_j$ 

#### Every Interval T

#### (a) Bidding

```
1: for each neighbor i \in \mathcal{D}_j do
```

2: exchange buffer map with i and derive  $K_{ij}$ 

3: set 
$$p_{ij}^{(k)} = \min(v_{ij}^{(k)}, \ddot{q}_{ij}), \forall k \in \mathcal{K}_{ij}$$

4: end for

5: order blocks in  $\cup_{i \in \mathcal{D}_j} \mathcal{K}_{ij}$  in non-increasing order of marginal utility  $v_{ij}^{(k)}/p_{ij}^{(k)}$  into list  $\mathbf{l_b}$  (excluding duplicates)

6:  $p_{ij}^{(k)} \leftarrow$  price of the first block in list  $l_b$ 

7: while  $e_j \geq p_{ij}^{(k)}$  do

8: send bid  $(I_{ij}^{(k)}, p_{ij}^{(k)})$  to the corresponding seller i

9:  $p_{ij}^{(k)} \leftarrow \text{price of the next block in list } \mathbf{l_b}$ 

10: end while

## Summary of mechanism at bidder

```
(b) After Bidding
1: p<sub>i</sub><sup>max</sup> ← highest bidding price sent to neighbor i, ∀i ∈ D<sub>j</sub>
2: set the lowest charge at i, c<sub>i</sub><sup>min</sup> = p<sub>i</sub><sup>max</sup>, ∀i ∈ D<sub>j</sub>
3: for each charge c<sub>ij</sub><sup>(k)</sup> received from i, ∀i ∈ D<sub>j</sub>, do
4: deduct e<sub>j</sub> by c<sub>ij</sub><sup>(k)</sup> received
5: pay c<sub>ij</sub><sup>(k)</sup> to i
6: c<sub>i</sub><sup>min</sup> ← min(c<sub>i</sub><sup>min</sup>, c<sub>ij</sub><sup>(k)</sup>)
7: end for
8: for each neighbor i ∈ D<sub>j</sub> do
9: if no bid was successful (no charge received from i) then
10: q<sub>ij</sub> = p<sub>i</sub><sup>max</sup> + δ
11: else q<sub>ij</sub> = c<sub>i</sub><sup>min</sup> − δ end if
12: end for
```



## Analysis

- Incentive compatibility
  - In mechanism design, a process is said to be incentive compatible if all of the participants fare best when they truthfully reveal any private information asked for by the mechanism
  - Seller incentive compatibility & bidder incentive compatibility

## Seller incentive compatibility

**Theorem 1.** The discriminative second price auction with seller reservation in Algorithm 1 is a revenue-maximizing equilibrium mechanism for a VoD seller

- ❖ The discriminative second price auction that we design is bid-independent →
- ❖ Truthful auction →
- ♦ (according to revelation principle) →

## Buyer incentive compatibility

**Theorem 2.** In the auction at seller i described in Algorithm l, for each block  $k \in \mathcal{K}_{ij}$ , bidding a price equal to the minimum between the block valuation and the market price at i, i.e.,  $\min(v_{ij}^{(k)}, \tilde{p}_i)$ , is a dominant strategy for bidder j.

## Proof of scheduling optimization

- First show that a Nash Equilibrium exists at the stable state of VoD streaming.
- Then show that the optimal solution to the distributed local optimization problem carried out through block auctions in Algorithm 1 and 2 can be combined to construct an optimal solution to the global optimization problem.

## Proof of scheduling optimization

**Theorem 3.** Without peer joins/departures and VCR operations, the following is a Nash equilibrium in the auction defined by Algorithm 1 at seller i: each participating bidder j with  $v_{ij}^{(k)} \geq \tilde{p_i}$  bids and pays  $\tilde{p_i}$ , all other participating bidders bid their true valuations and lose the auction.

#### Proof:

√1. Assume

$$v_{ij}^{(k)} \ge \tilde{p_i}$$

 $\sim$ 2. Assume  $v_{ij}^{(k)} < \tilde{p}_i$ 

$$v_{ij}^{(k)} < \tilde{p_i}$$



## Proof of optimality

**Theorem 4.** Algorithms 1 and 2 solve (10), i.e., achieves social welfare maximization, in a stable P2P VoD overlay.

#### Define the global optimization problem:

Maximize  $\sum_{j \in \mathcal{N}} \sum_{i \in \mathcal{D}_j} \sum_{k \in \mathcal{K}_{ij}} v_{ij}^{(k)}(a_{ij}^{(k)})$  (10)

Subject to:

$$\mathcal{P}_{global} \begin{cases} \sum_{i \in \mathcal{D}_{j}} \sum_{k \in \mathcal{K}_{ij}} p_{ij}^{(k)} a_{ij}^{(k)} \leq \hat{e}_{j} & \forall j \in \mathcal{N} \\ \sum_{j \in \mathcal{D}_{i}} \sum_{k \in \mathcal{K}_{ij}} a_{ij}^{(k)} \leq o_{i} & \forall i \in \mathcal{N} \\ \sum_{i \in \mathcal{D}_{j}} a_{ij}^{(k)} = z_{j}^{k} & \forall j \in \mathcal{N}, \forall k \in \mathcal{K}_{ij} \end{cases}$$

$$a_{ij}^{(k)}, z_{j}^{k} \in \{0, 1\}, \forall j \in \mathcal{N}, \forall i \in \mathcal{D}_{j}, \forall k \in \mathcal{K}_{ij}$$



#### Proof procedure

- (1) Prove the relaxation form of (5) always hat an integral optimal solution, i.e., the integrality gap of (5) is non-existent
- (2) the KKT conditions of the relaxation of (5), aggregated across all peers, are equivalent to the KKT conditions of the relaxation of (10)



### **Evaluation**

- Multi-thread P2P network simulator in Java
- Supports peer dynamics (VCR operations, peer joins and departures)

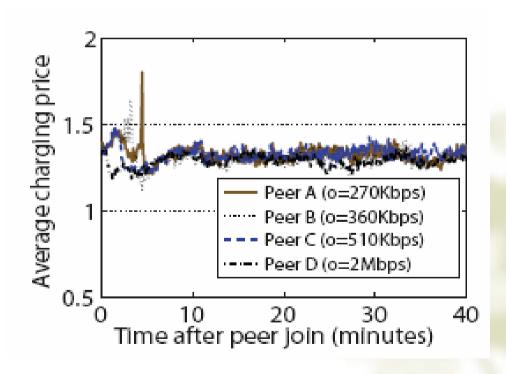
#### **Evaluation**

#### Configuration:

- 80 minute video is streamed
- Playback bitrate 450 Kbps
- Server upload capacity = 10 Mbps
- Peer upload capacity distribution = Pareto distribution with range=[250 Kbps, 10 Mbps] and k=2 or 3 (default)
- See Peer lifetime = 30 minutes
- Average duration of VCR operations = 5 minutes
- → Buffer on each peer = 20-minute playback
- → Buffer maps exchange period = 5 sec.

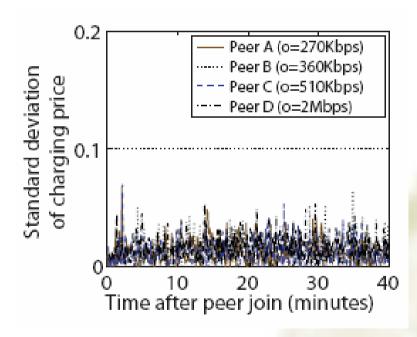
## **Evaluation**

### Stable charging price

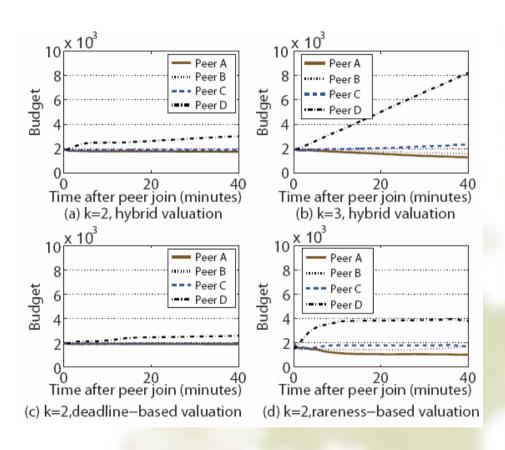


### **Evaluation**

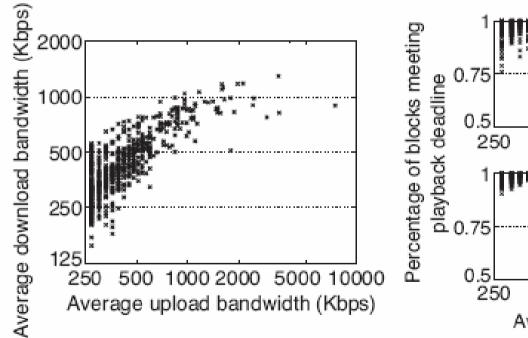
Stable charging price (cont.)

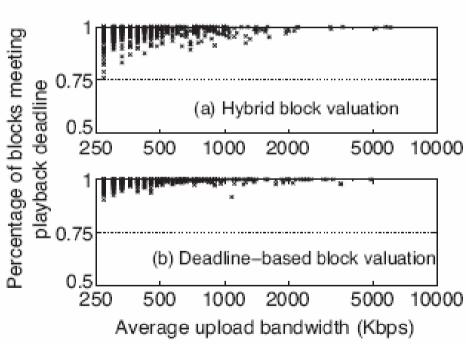


 Evolution of budget at peers with different upload capacities



#### Incentive

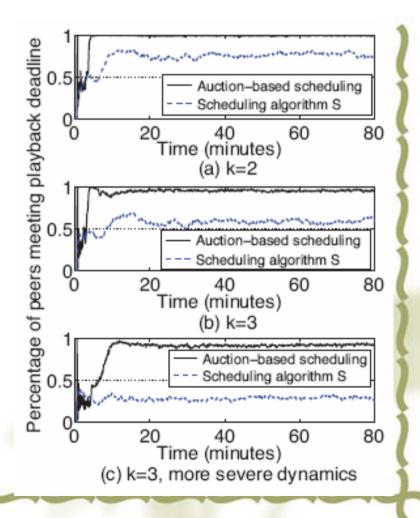






### Performance compared with other mechanisms

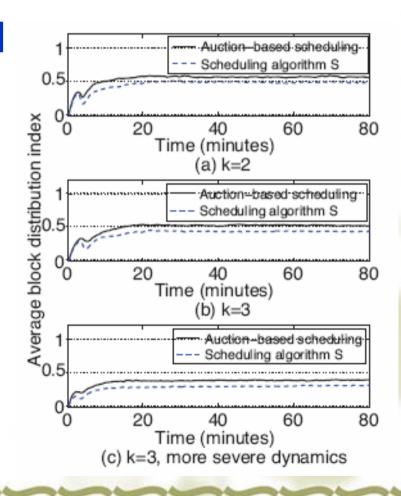
Evolution of the playback deadline satisfaction





 Performance compared with other mechanisms

Evolution of the average block distribution index





## Summary

#### Design auction-based mechanism

to incentive peers to contribute its maximal upload capacity and achieve optimal scheduling of block exchanges among peers

#### Analysis proves

- the incentive compatibility of the scheme
- the existence of Nash equilibrium under certain conditions

#### Evaluation shows

that the scheme exhibits good performance in realistic senarios



## Summary

- Chuan Wu, Zongpeng Li, Xuanjia Qiu, Francis C.M. Lau, "Auction-based P2P VoD Streaming: Incentives and Optimal Scheduling", submitted to INFOCOM 2010
- Future work:
  - Cross-overlay help
  - Research on evolution of budget distribution
  - Wealth condensation? Taxation?
  - **S** Emulation

