

# Algorithmic Problems in Network Economics

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## Networked World



• A classical view of the internet



- Open, evolutionary architecture
- Lacks central control and coordination
- Dynamically varying infrastructure and users
- Resource sharing
- Interesting mix of computational and strategic complexities





#### servers

QuickTime?and a decompressor are needed to see this picture.

#### clients

- Matching n clients (users) to m servers (access points)
- A compatibility graph:

- edge (i,j) if client i can be served by j

- Identical servers with unit resource
- Latency as cost of matching:

– a server matched to k clients has latency = k

• Quality of matching in this uncoordinated world?

## Price of Anarchy



• Selfish Routing [Roughgarden et al., Papadimitriou]



- (Social) optimum = 0.5 flow on each link
  - latency = 3/4
- Self-interested (Nash) optimum flow = 1 on top link
  - latency = 1
- Price of Anarchy = Ratio of Social to Nash Optimum
  - this example 4/3

## Anarchy in Load Balancing



• What is the worst-case ratio between costs of optimum and Nash matching?

QuickTime?and a decompressor are needed to see this picture.

Input

Opt Cost = 3 Nash Cost = 5 Arbitrary Cost = 5

## Anarchy in Load Balancing



- With identical servers, OPT is always NASH, but not vice versa.
- I.e. best case Nash = Opt
- Ratio between worst-case Nash and Opt?

QuickTime?and a decompressor are needed to see this picture.

Input

Opt Cost = 3 Nash Cost = 5 Arbitrary Cost = 5



• Theorem 1: For identical servers, price of anarchy is atmost

$$(1+2/\sqrt{3}) = 2.155$$

• Theorem 2: Price of anarchy is at least 2.001

## More Bounds



• For non-identical servers, social optimum no longer Nash Equilibrium.



- Theorem 3: PoA < 5/2.
- For Lp norm latency, PoA = O(p/logp)
- Selfish Load Balancing, S.-Toth-Zhou, Algorithmica '07.
- Price of routing unsplittable flow, Awerbuch, Azar, Epstein, STOC '05

## **Algorithms**



- Nash matching by local swaps:
  - in each round, a user switches to better server.
  - Provably  $O(n^2)$  rounds.
- Instead suppose clients arrive one by one and each chooses the best available server at that time.
  - -Greedy Matching
  - -Not necessarily a Nash matching
  - -But can be shown to be O(1) factor optimal.



## Mobility and Load Balancing



- Wireless access points (APs) at airport, malls, etc.
- User can select and use any AP
  - Selected AP need not be in range
  - User moves towards selected AP if necessary
- Strategic tradeoffs between cost of mobility and wireless service quality
  - Users are rational, selfish entities
    - Maximize personal benefit
    - No regard for system cost

#### Modeling the Game



- User arrive sequentially
- AP bandwidth shared equally among attached users
  - AP with fewer attached users preferable
- Distance of AP from user's location
  - Closer AP preferable (less mobility, better signal)
- Cost function (user I and AP j),

 $\mathbf{C}_{ij} = \mathbf{x}^* \mathbf{x}_j + \mathbf{\beta}^* \mathbf{d}_{i,j}$ 

where  $x_j =$  number of users at AP *j* 

 $d_{i,i}$  = distance between user *i* and AP *j* 

 $\gamma$  ,  $\beta$  are constants (same for all users)

#### Simple Distributed Algorithm



- Greedy algorithm
  - Upon arrival, each user picks the AP with currently minimum cost
  - -No future swaps done.
- Theorem: The greedy always produces a Nash equilibrium
- Social optimal always Nash.

#### Price of Anarchy



- $\beta = 0$  (Mobility cost zero)
  - Only Nash equilibriums are those that distribute users evenly
  - Pessimistic price of anarchy = 1
- Y = 0 (Users bandwidth-agnostic)
  - Unbounded price of anarchy
- General case (neither  $\beta$  nor  $\gamma$  zero):

- Open



#### **Spectrum Auctions**







- Auctions: efficient allocation of scarce resources
- Auctioneer: dynamic price discovery based on demand
- Users: request and acquire spectrum when they need it



#### **Computational Complexity**



- Externality: interference
  - Spatial reuse possible
  - Nearby users cannot use same channel



- Combinatorial auctions NP-complete
- Hard even without expressive bidding due to graph coloring
- Focus on computational efficiency, without strategic considerations.

• Bids: the desired quantity of spectrum f at a per-unit price p



### **Bidding by Price-Quantity Curves**









#### **The Auction Clearing Problem**

Allocate price(s) and spectrum to maximize the total revenue R(.) subject to Interference Constraints

## **Analytical Bounds**



Theoretical bounds	Clearing with Uniform Pricing	Clearing with Discriminatory Pricing
Revenue efficiency	$\boldsymbol{R} \geq \frac{1}{3}\boldsymbol{R}_{OPT}$	$R \geq \frac{n}{3(n+1)}R_{OPT}$
complexity	$O(n\log n + n\log U)$	polynomial
When the conflict graph is a tree	$R = R_{OPT}$	$R = R_{OPT}$



### Strategy-proof Spectrum Auctions



#### Strategy-proof Spectrum Auctions

- Input:
  - Spectrum as k channels: 1, 2, ..., k
  - A set of n bidders
- Output:
  - A polynomial time strategy-proof mechanism for spectrum allocation
  - Subject to interference constraints

- Motivation:
  - Dynamic redistribution of FCC's long term licenses
  - Fair and open
  - Economic Efficiency



## **Graph Coloring**



- Conflict-free channel allocation = graph coloring
- Computationally, graph coloring intractable and in-approximable.





- If all we care about is truthfulness, a trivial solution:
  - Allocate channels to k highest bidders
  - Price: Bid of (k+1)th highest bidder



• Inefficient spectrum utilization: a<sub>3</sub> and a<sub>4</sub> left out

### Truthfulness with Maximal Utilization



- Always allocate a channel unless doing so precludes another user
- Desiderata:
  - Truthfulness
  - Pareto optimality
  - Computational efficiency
- VCG doesn't satisfy the computational efficiency requirement

#### First Attempt



- Sort and Greedily allocate channels
  - Allocate lowest available index
- Each winning bidder pays the bid of highest unallocated neighbor



VIOLATES TRUTHFULNESS !!!

### Another Attempt

- Greedily allocate channels
- For each Winning bidder a<sub>i</sub> determine neighbor a<sub>i</sub> s.t.
  - $-a_i$  loses when  $a_i$  is present, but
  - $-a_i$  wins when  $a_i$  is absent
- Charge a<sub>i</sub> the bid of a<sub>i</sub>





#### New Auction: Veritas



- Sort and Greedily allocate channels (lowest available first)
- Veritas-Pricing:
  - A winner i pays the bid of its critical neighbor C(i)
  - To determine Critical Neighbor for i
    - run greedy algorithm with B b<sub>i</sub>
    - Critical Neighbor of i is the first one to be denied a channel.





# of channels = 2

### **Proof of Veritas**



- Theorem: Veritas is truthful, achieves pareto optimality, and runs in O(n<sup>3</sup>k)
- Proof sketch
  - Criticality: Unique critical value for each winning bidder.
  - Monotonicity: A bid above the critical value always wins.
  - Truthfulness: If we charge every bidder its critical value, no incentive to lie.

#### **Bib and Collaborators**



- Joint work with
  - Buragohain, Gandhi, Toth, Zheng, Zhou, Zhou
- Papers
  - Selfish Load Balancing, Algorithmica, 2007
  - A game-theoretic analysis of wireless access points selection by mobile users, Computer Communication '08
  - Towards real-time dynamic spectrum auctions, Computer Networks, '08
  - eBay in the sky: strategy-proof wireless spectrum auctions, Mobicom '08



#### Thank You!