Pragmatic Algorithmic Game Theory



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Algorithmic Game Theory

- "Research at the interface of CS, game theory, and economic theory, largely motivated by the Internet."
- Particular topics of concern
 - Design of new mechanisms, esp. auctions and market algorithms
 - Analysis of existing mechanisms
 - Equilibrium computation
- Typically, very general settings attacked using theoretical tools
- This approach has yielded impressive
 - impossibility
 - optimality

Introduction

- approximation results.
- However, sometimes it is very difficult to obtain clean theoretical results that address complex, realistic problems



Algorithmic Game Theory

Edited by Noam Nisan, Tim Roughgarden, Éva Tardos, and Vijay V. Vazirani

Foreword by Christos H. Papadimitriou

CAMBRIDGI

Pragmatic AGT

- More traction on realistic problems by leveraging one or both of the following forms of pragmatism:
 - Aiming to achieve good performance only on problems of interest, rather than in relatively unconstrained settings



- 2. Adopting statistical rather than analytical methods, thereby defining problems of interest implicitly via a dataset and/or appealing to data-driven measures of performance.
- I'll describe work that attacks "core AGT" problems but is pragmatic in both senses
 - I'm far from the only one working in this vein
 - Indeed, EC is a great place for interaction between those who adopt "pure" and "pragmatic" approaches

REPURPOSING RADIO SPECTRUM VIA AN "INCENTIVE AUCTION"

[Frechette & LB, unpublished]

Also draws on [Hutter, Hoos & LB, 2011]; [Xu, Hutter, Hoos & LB, 2009]

Introduction

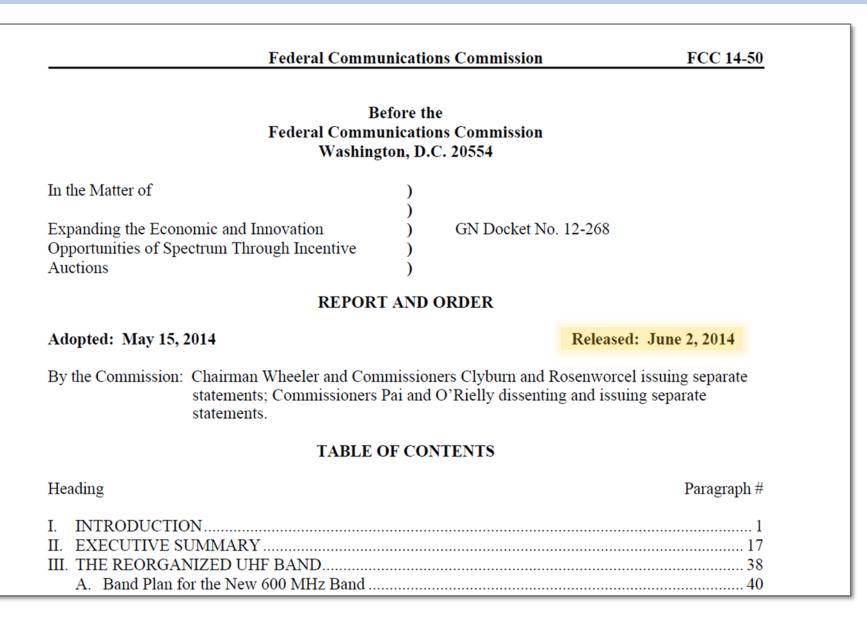
Incentive Auctions Resources

>

FCC's "Incentive Auction"

deral Communications Commission				A	Display Options	-	
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Home / Our Work / Incentive Auction	IS						
					Find out about business opportunities and the Incentive Auctions process		
Incentive Auctions	ALT	20		LEAR	I' PROGRAM		
Unleashing spectrum to meet America broadband	's demand for mobile	V					
The United States leads the wor	rld in key areas			RC Quick Links			
of wireless infrastructure and in including being the first country Long-Term Evolution (LTE) tech	y to have 4G	xplore the Broadca elevision Spectrum		Learning Everythin Reverse-Auctions f			
networks at scale and to enable	(Th C)	uction Rulemaking		Report and Order S	Staff Summary		
of white space spectrum. Meanwhile, demands on both licensed and unlicensed				Incentive Auctions	NPRM		
spectrum are increasing dramat	tically.						

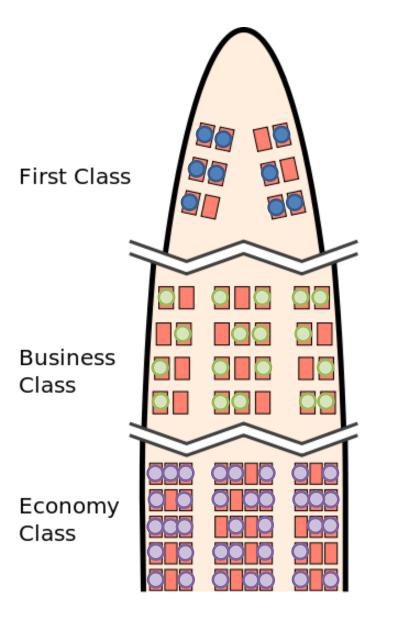
Approved and Ready to Go



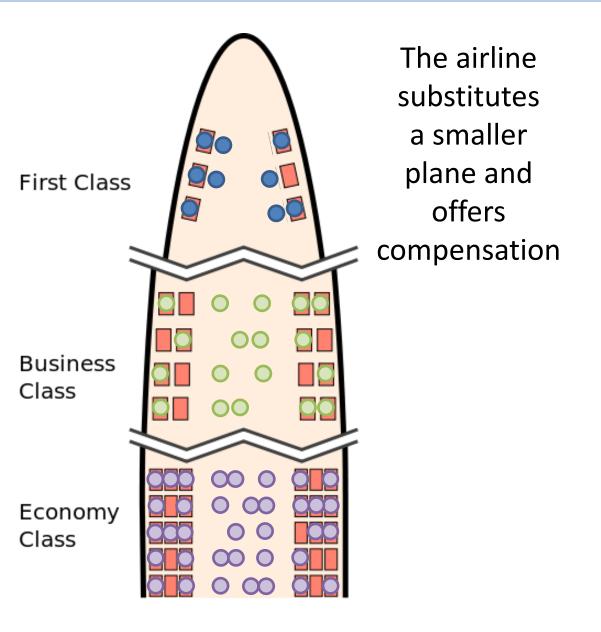
The FCC's "Incentive Auction"

- Reverse (descending-price) auction for broadcasters
 - stations declare they're willing to stop broadcasting at a given, initially high, price
 - price descends as long as stations can feasibly be "repacked" into the reduced band, given interference constraints
- Forward (ascending-price) auction for telecom firms
 - prices in each region increase while demand exceeds supply
- Auctions linked to ensure revenue target is met
 - if not, clearing target reduced and the auctions continue

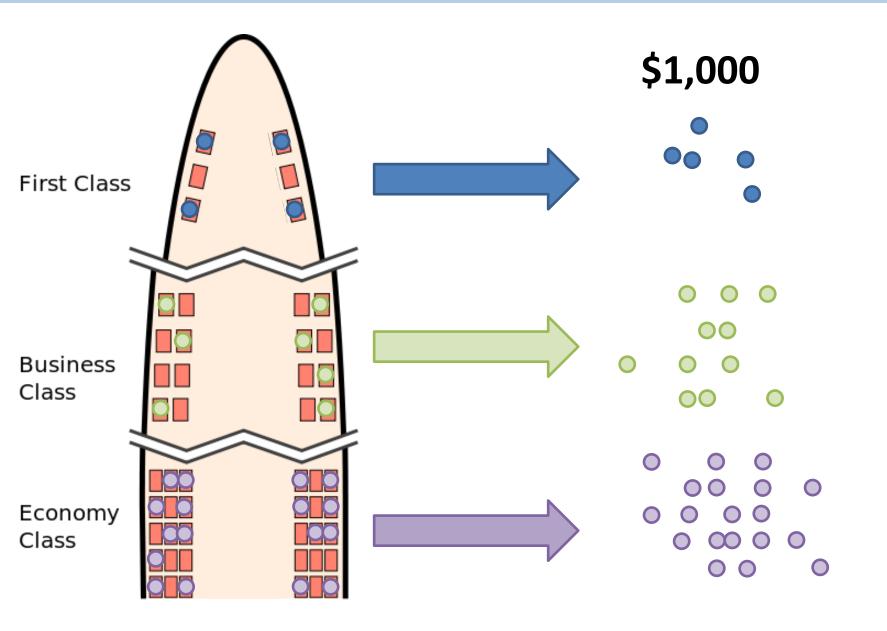
How Does the Reverse Auction Work?

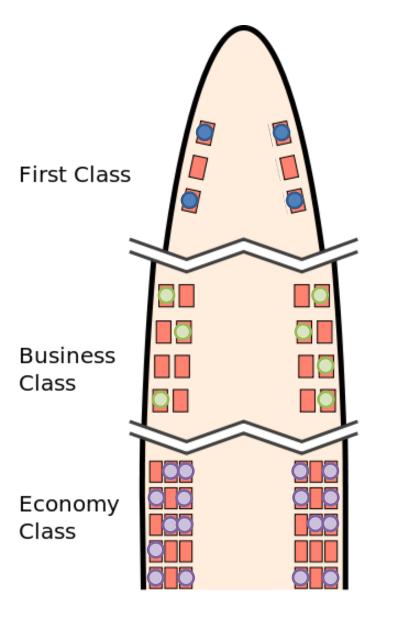


- Let's consider the example of airline overbooking, where passengers either fly in their assigned cabin or are compensated to give up their seat
- Thus, the feasibility constraint is (# passengers in cabin) ≤ (# seats)
- We'll use a **descending clock** auction to set compensations
- Let's start with a plane big enough to hold everyone...

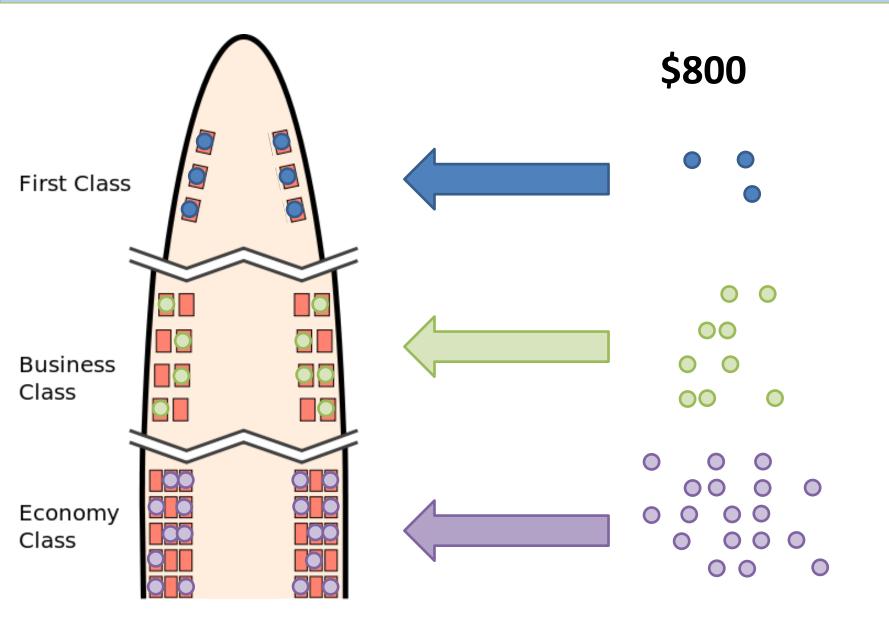


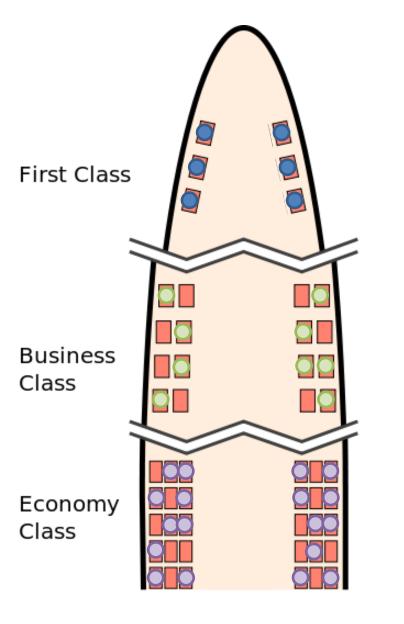
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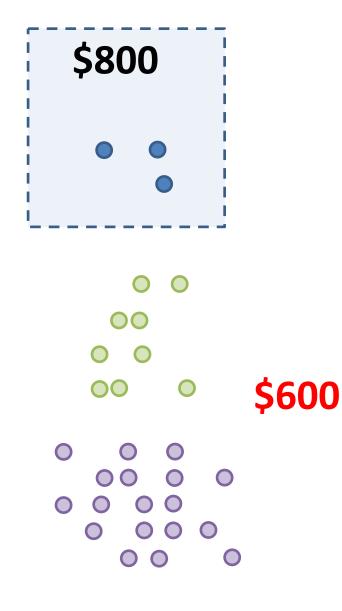


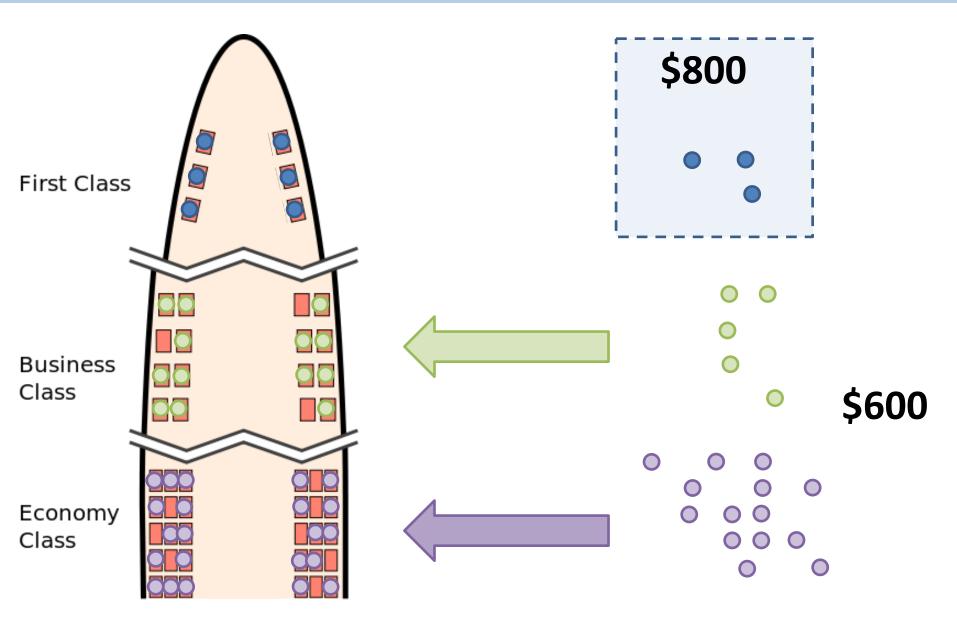


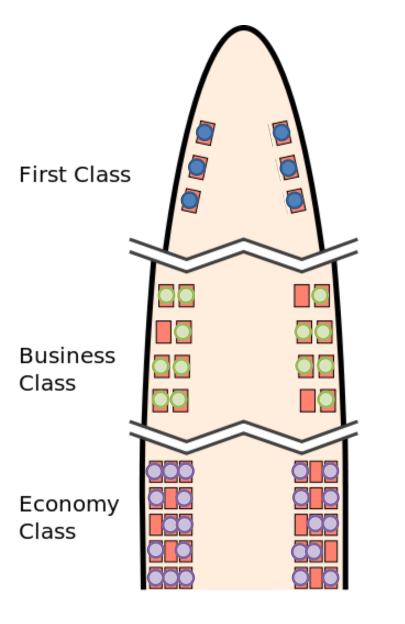
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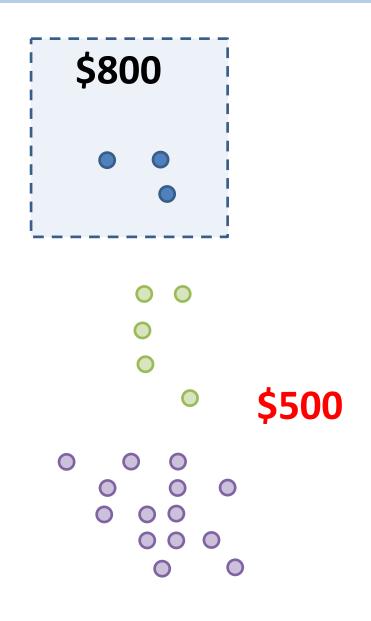


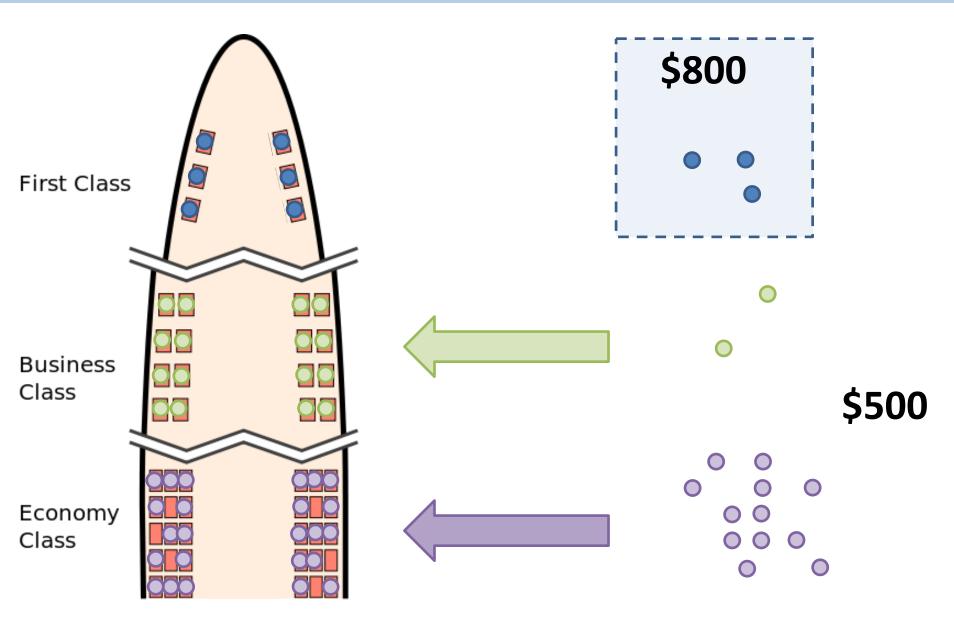


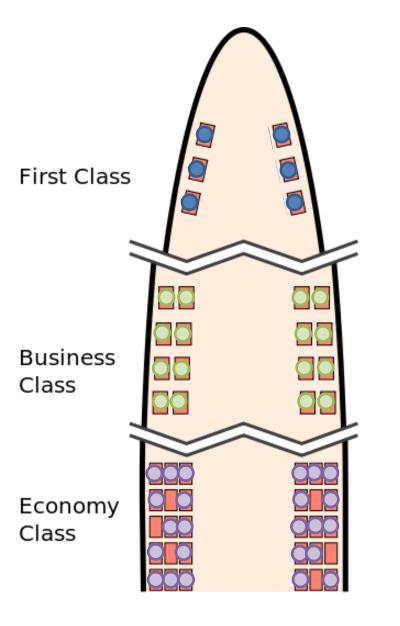


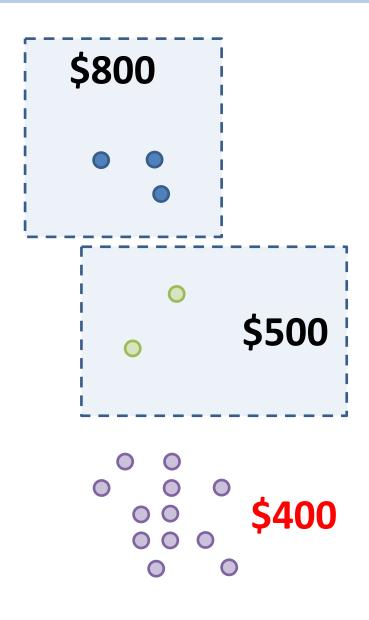


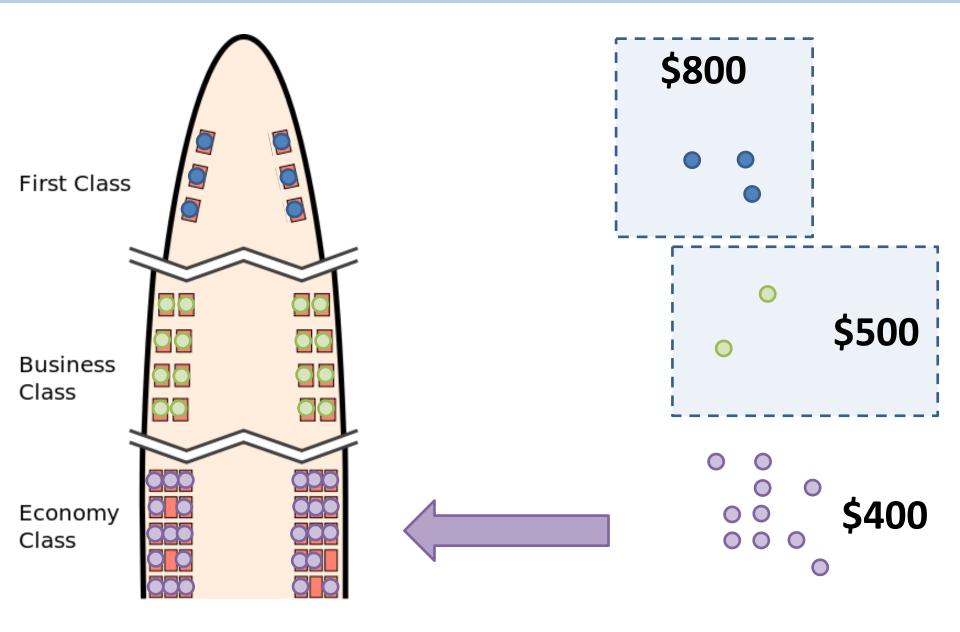


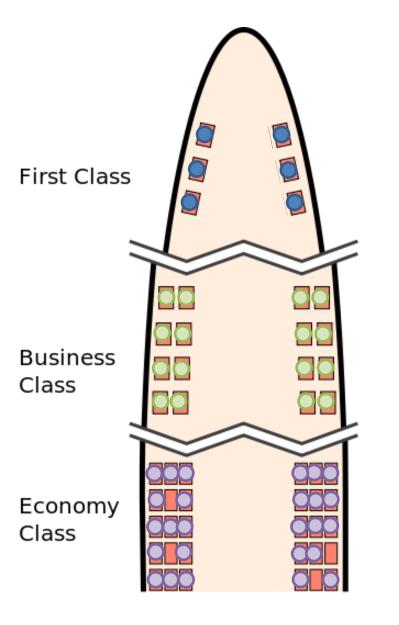


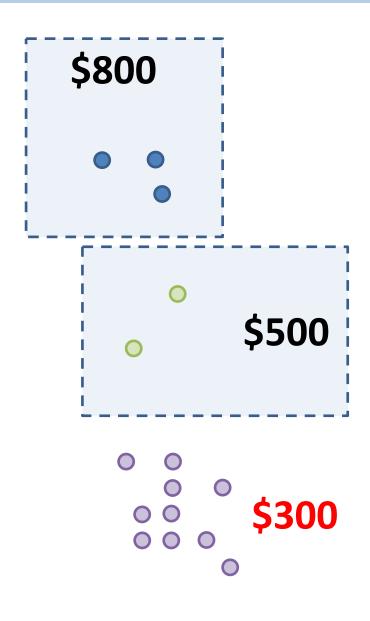


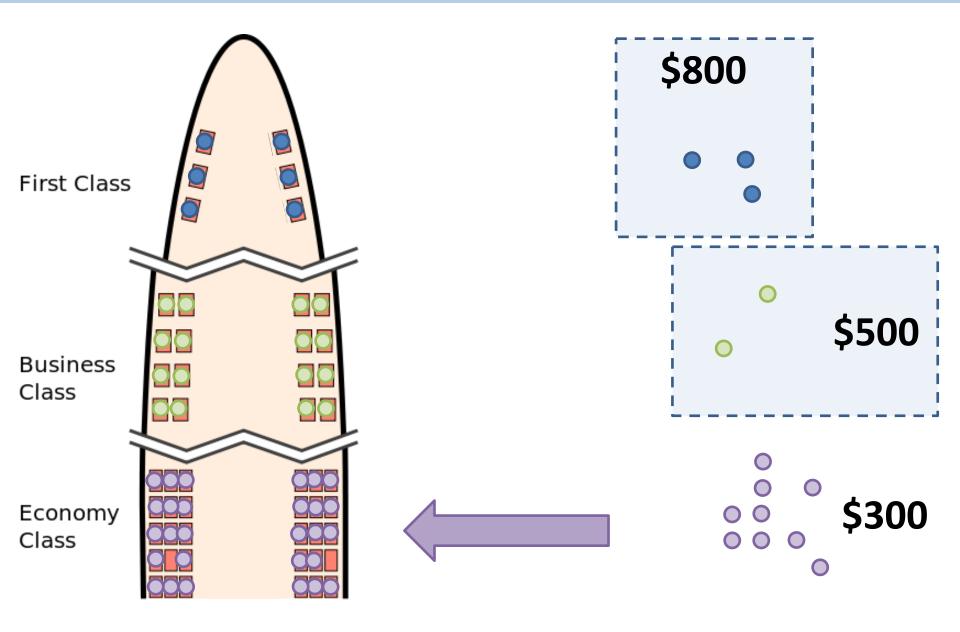


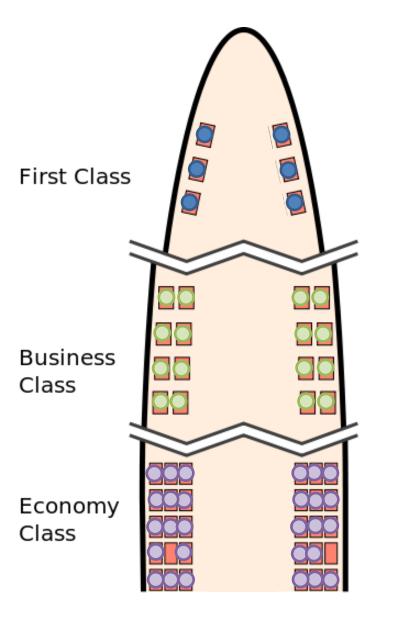


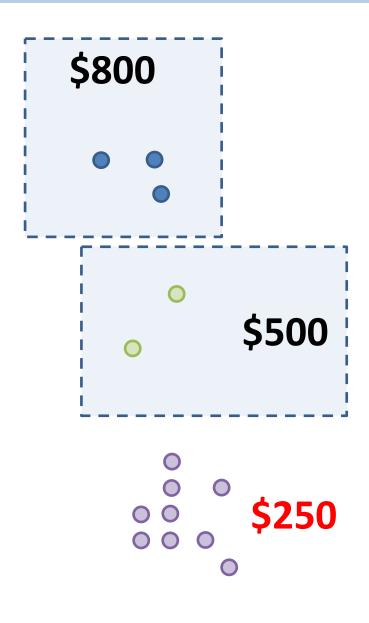


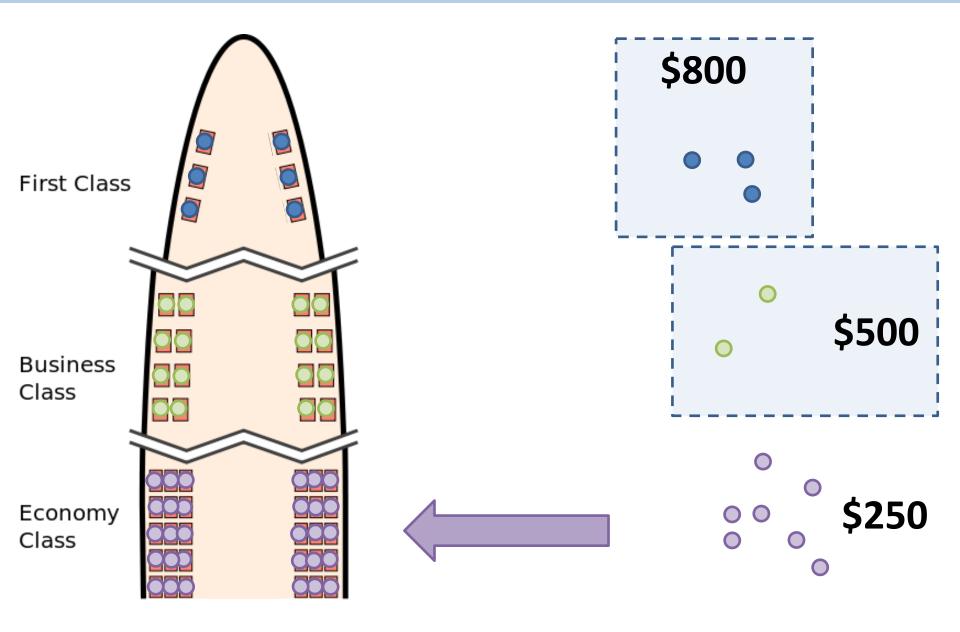


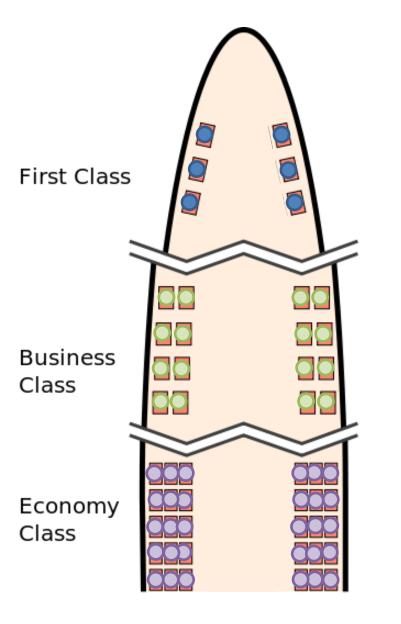


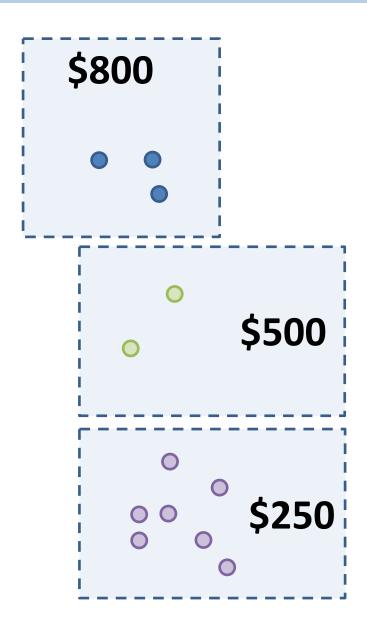


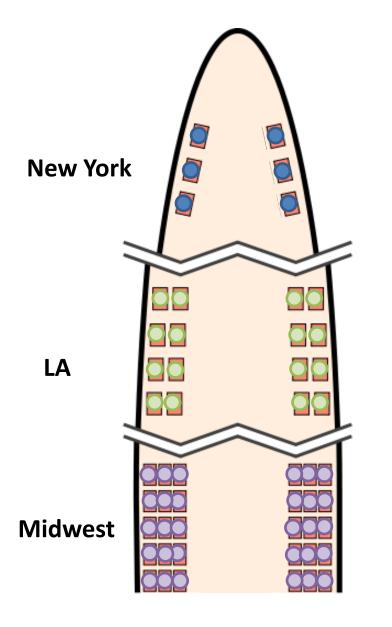


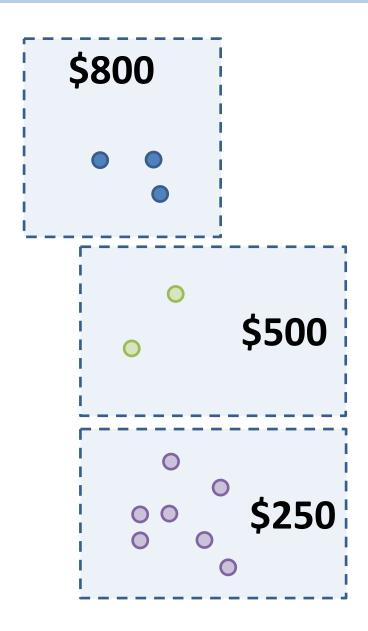












Real Constraints are Messier



The feasibility constraints are not uniform

 nearby stations can freeze at different times

Introduction

Feasibility Testing

Key computational problem: testing the feasibility of a given repacking, based on interference constraints

- Basis of "frozen test": millions per auction
- A hard graph-colouring problem
 - 130,000 constraints
 - Initial skepticism about whether this problem could be solved exactly at a national scale
- We're doing it, using tools from empirical algorithmics
 - SAT encoding
 - automatic algorithm configuration
 - algorithm portfolios

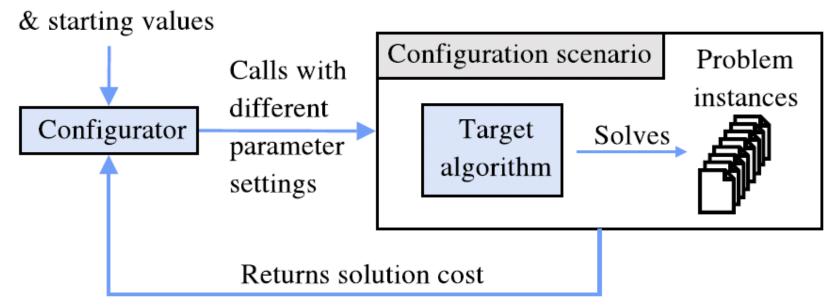
SAT Encoding

- $x_{s,c}$: the proposition that station *s* is assigned to channel *c*
 - one such variable for every station s and channel c
- Station *s* must broadcast on one of its allowable channels
 - For every station s and set of allowable channels $\{c_1, \dots, c_n\}$, create a clause $(x_{s,c_1} \lor \cdots \lor x_{s,c_n})$
- Station *s* may broadcast on at most one of these channels
 - For every pair of channels c_1 and c_2 allowed for station s, create a clause $(\neg x_{s,c_1} \lor \neg x_{s,c_2})$
- The repacking does not cause harmful interference
 - For every interference rule stating that s_1 cannot broadcast on c_1 while s_2 broadcasts on c_2 , create a clause $(\neg x_{s_1,c_1} \lor \neg x_{s_2,c_2})$
- Note: mostly 2-clauses
 - good for unit propagation: implies clique constraints

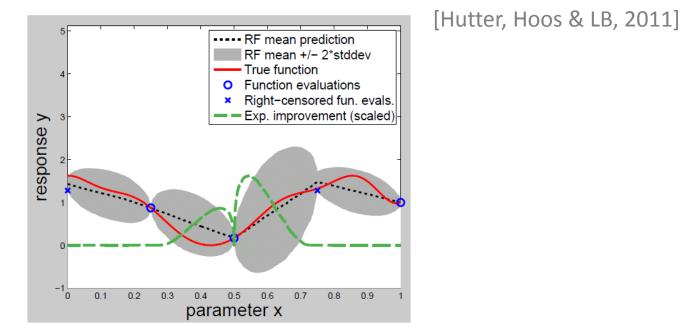
Algorithm Configuration

- High-performance solvers for NP-complete problems like SAT are typically parameterized
 - which branching heuristic, variable ordering, preprocessing strategy, clause learning technique, ...
- Address with algorithm configuration

Parameter domains



Sequential Model-based Algorithm Configuration (SMAC)



Initialize with a single run for the default configuration **repeat**

Learn a random forest model $m: \Theta \times \Pi \to \mathbb{R}$ from data so far

Marginalize out instance features: $f(\theta) = \mathbb{E}_{\pi}[m(\theta, \pi)]$

Find θ that maximizes expected improvement in $f(\theta)$ over incumbent Compare θ to the incumbent, updating if it's better.

until time budget exhausted

Algorithm Portfolios

- Often different solvers perform well on different problem instances
- Idea: build an algorithm portfolio, consisting of different algorithms that can work together to solve a problem
- **SATzilla**: state-of-the-art portfolio developed by my group (2003-2013)
- some key ideas:

Introduction

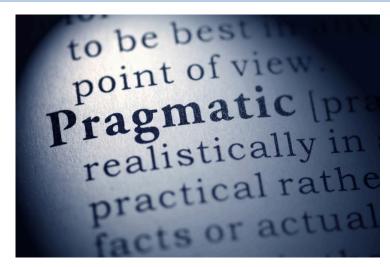
presolver scheduling

- machine learning to choose algorithm on a per-instance basis
- constituent solvers can be automatically configured ("Hydra")

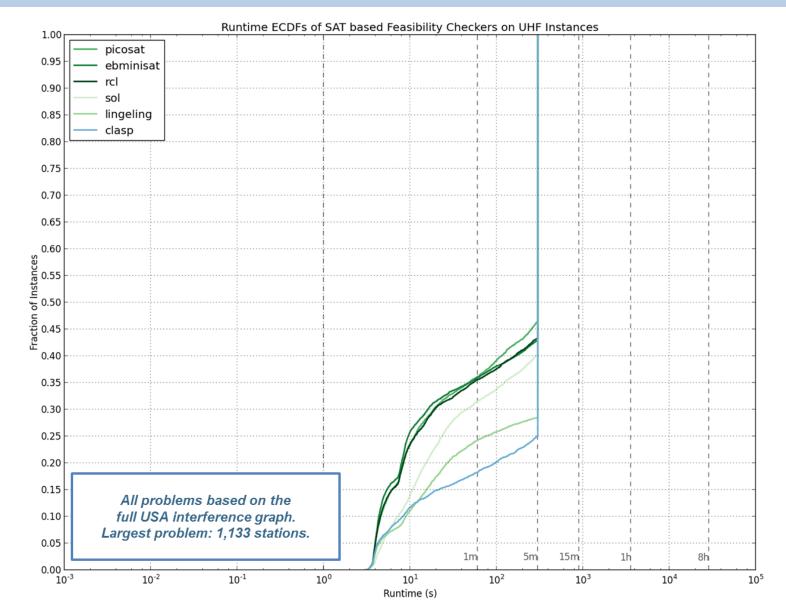
Introduction

How These Techniques are Pragmatic

- Target distributions of interference graphs that arise in practice
 - subsets of the whole-country graph
- Achieve good performance most of the time, but tolerate occasional failures
 - treat timeouts as UNSAT

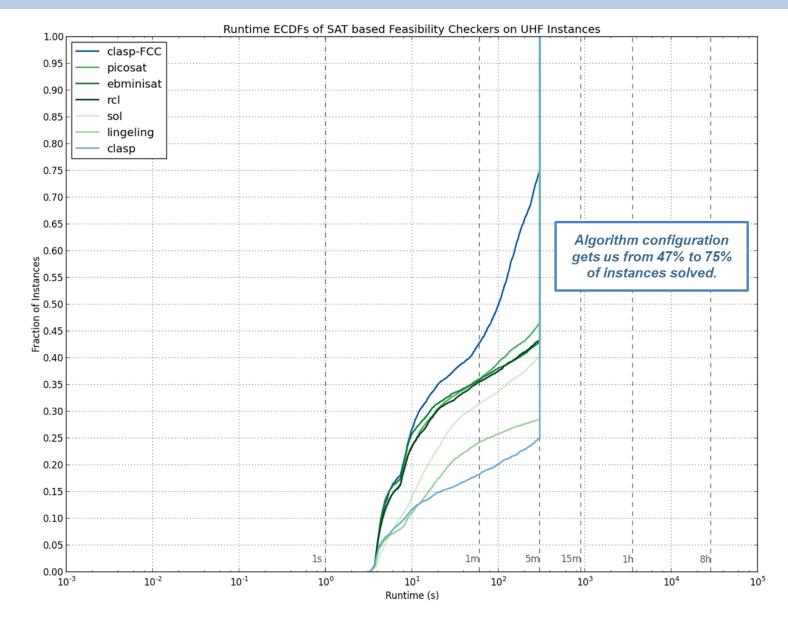


Comparing off-the-shelf SAT solvers (5 min cutoff)



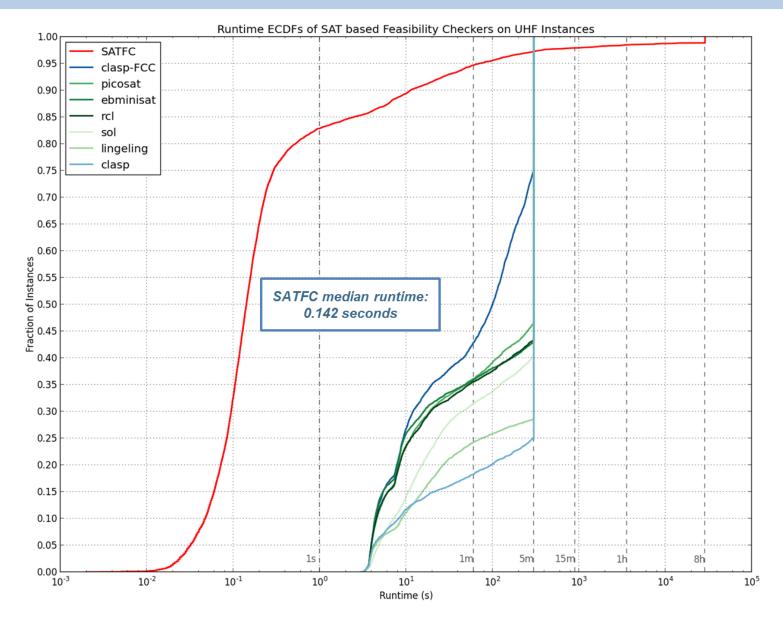
Introduction

Adding our specially-configured version of clasp

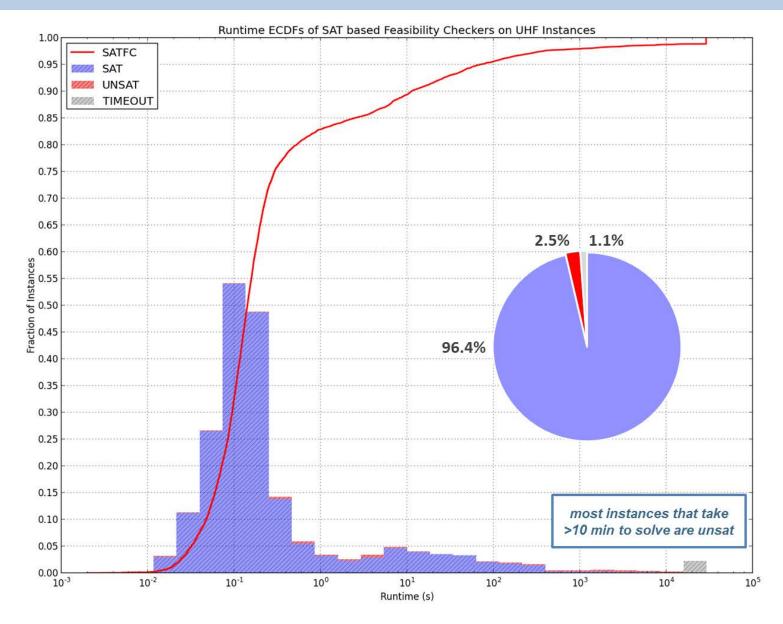


Introduction

Adding presolvers, other optimizations (8h cutoff)



SATFC performance and SAT/UNSAT breakdown



Including VHF Bands

- So far we've considered mechanisms for compensating stations for going off air
 - what if UHF stations can also be paid to move to a lower spectrum band (VHF)?
 - we now potentially face multi-minded bidders
 - willing to go to VHF at one price; off air at another
 - theory gives us less guidance here
 - An area of active study; multiple proposals have been made

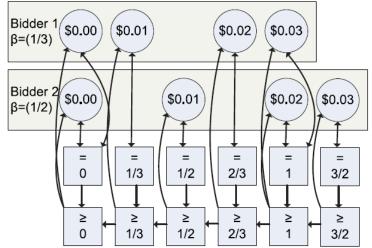


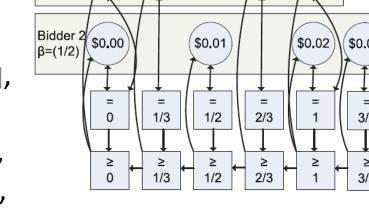
EQUILIBRIUM COMPUTATION AND COMPUTATIONAL MECHANISM ANALYSIS

[Jiang & LB, 2011] [Thompson & LB 2009; 2013; unpublished] Introduction

Action-Graph Games

- **Compactly represent games** exhibiting context-specific independence, anonymity or additivity [Jiang & LB, 2011]
 - the sort of structure exhibited by typical market settings
 - **pragmatic:** target typical case, not worst case
- **Fast algorithms** for computing quantities of interest
 - Nash equilibrium [Jiang & LB, 2011], correlated eq [Jiang & LB 2013], pure-strategy eq [Jiang & LB, 2010], all equilibria [Thompson & LB, 2011], *e-equilibrium* [Daskalakis, Schoenebeck, Valiant & Valiant, 2009]





Computational Mechanism Analysis

- What happens in equilibrium of real-world mechanisms, under given valuation distributions?
 - go beyond theoretical analysis of mechanism properties
 - answer quantitative questions (e.g., "which gives higher revenue?")
 - gives answers even in complex domains
 (reserve prices; messy valuation distributions; general eqm concepts)
- How it works:
 - repeatedly sample games from the valuation distribution
 - represent these games as AGGs
 - solve them using general AGG solvers
 - obtain statistics on economic quantities of interest
- **Pragmatic**: statistical, data-driven



CMA Application: Ad Auction Evolution

[Thompson & LB, 2009]

- Search engines used **different auctions** over the years
 - GFP: Yahoo! and Overture 1997-2002
 - uGSP: Yahoo! 2002-2007
 - wGSP: Google, Microsoft, Yahoo! 2007-present

Question: Is wGSP better than GFP and uGSP?

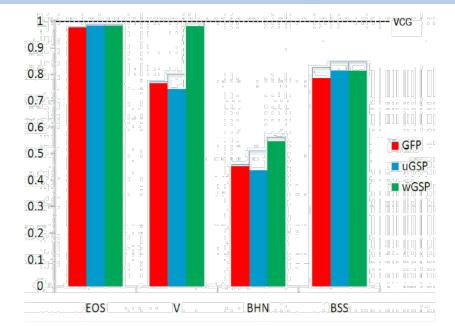
- better revenue?
- better efficiency?

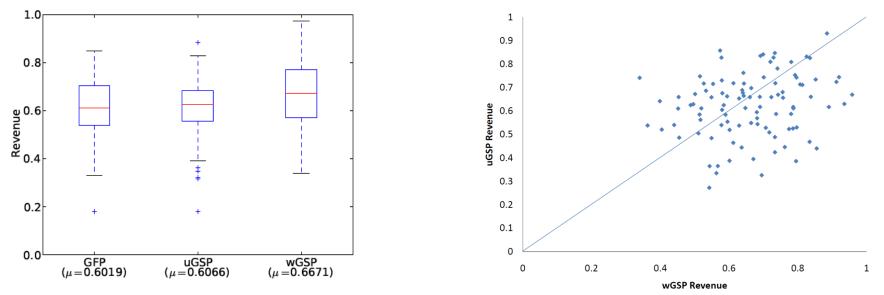
Approach:

- Construct perfect information AGGs sampled from widely studied valuation distributions
- Compute revenue/welfare optimal/pessimal equilibria

[Thompson & LB, 2009]

Analyzing Ad Auctions: Efficiency, Revenue





How should reserve prices be set in GSP?

[Thompson & LB, 2013]

- Goal: maximize (short term) revenue in GSP.
 - 6 GSP variants; parameters control the way reserve prices are set
- Represent as AGG, varying parameters on a fine grid
- Find all equilibria
- Find parameters that optimize best/worst equilibrium revenue

Revenue-Pessimal Equilibrium			Revenue-Optimal Equilibrium		
Auction	Revenue	Parameter(s)	Auction	Revenue	Parameter(s)
Vanilla GSP	3.814	_	Vanilla GSP	9.911	_
Squashing	4.247	s = 0.4	\mathbf{QWR}	10.820	r = 5.0
QWR	9.369	r = 9.0	Squashing	11.534	s = 0.2
Anchoring	10.212	r = 13.0	UWR	11.686	r = 11.0
QWR+Sq	10.217	r = 15.0, s = 0.2	Anchoring	12.464	r = 11.0
UWR	11.024	r = 15.0	QWR+Sq	12.627	r = 7.0, s = 0.2
\mathbf{UWR} + \mathbf{Sq}	11.032	r = 15.0, s = 0.6	UWR+Sq	12.745	r = 9.0, s = 0.2

Takeaway: unweighted reserves are both robust and outperform weighted reserves.

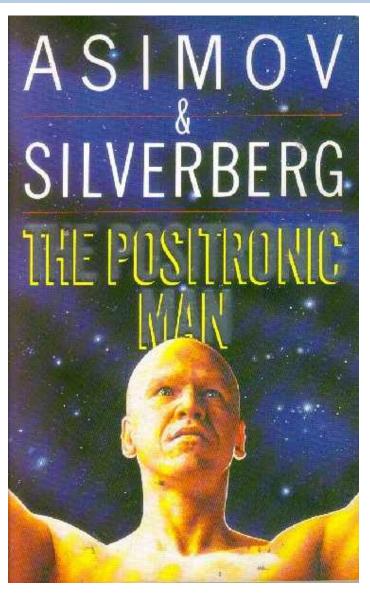
Introduction

Positronic Economist

- AGGs provide general tools for equilibrium computation
 - but not straightforward to encode domains of interest
- Positronic Economist

[Thompson & LB, unpublished]

- A general tool for taking manual effort out of mechanism analysis
- Leverages existing algorithms for computing (general; pure-strategy; all; Bayes-Nash) equilibria
- Generates compact AGGs from natural descriptions of mechanisms and settings.



Example: Independent Private Values

• Preferences (in math):

$$u_i(x,p_i) = \begin{cases} v_i - p_i & x = i \\ -p_i & \text{otherwise} \end{cases}$$

• Preferences (in posec):

```
def u(i, v, o, a_i):
    if o.i_win:
        return v[i]-o.my_payment
        return -o.my_payment
```

Note: we can study more complex settings too

 budgets, common values, etc.

Example: Single-Unit Auction

- Mechanism:
 - First-price auction with bids {0, ..., 9}
 - uniform random tie-breaking
- Mechanism (in posec):

def M(setting,i,theta_i,a_N):
 # i loses if anyone bids more than him
 if a_N.any([a for a in a_N.actions if a>a_N[i]]):
 return ProjectedOutcome(i_win=False,my_payment=0)

i wins in a t-way tie (where t can be 1)

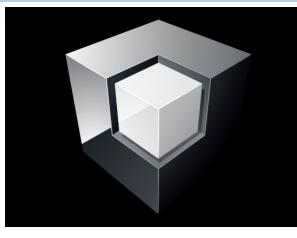
t = a_N.count(a_N[i])

return PosEc.Distribution(0,[1.0/t,1-1.0/t])

Introduction

posec can build an AGG in two ways

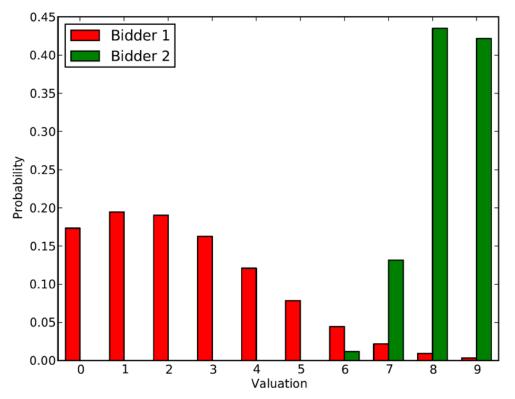
- Black-box structure inference
 - Probe every joint action × joint type in the simultaneous-move Bayesian game.



- White-box structure inference
 - Reason from the structure of the mechanism definition to detect independencies without probing everything.
 - Facilitated via Python's operator overloading and Positronic Economist's accessor functions (e.g., count, distribution).

posec Performance

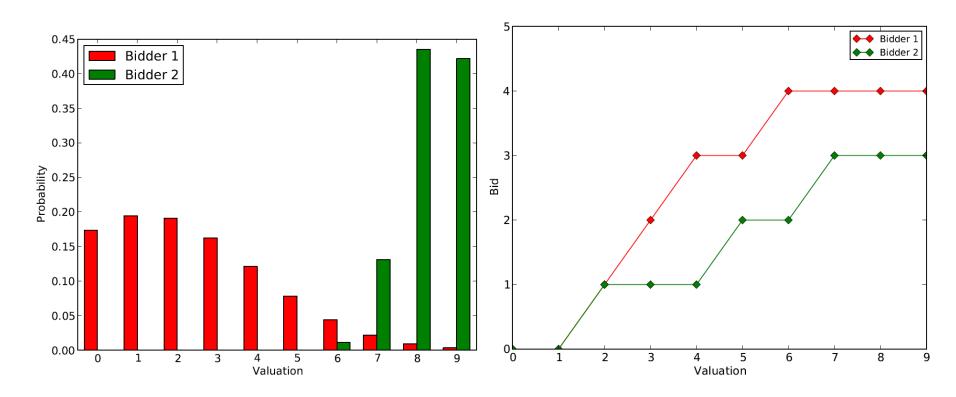
- Asymmetric, single-unit, first-price auction
- n = 2 players, t = 10 types each, a = 10 actions each



- Induced normal form: $(a^t)^n \ge 10^8 \text{ TB}$
- BAGG found by posec: **27kb file**, produced in ~6s

Introduction

posec Performance

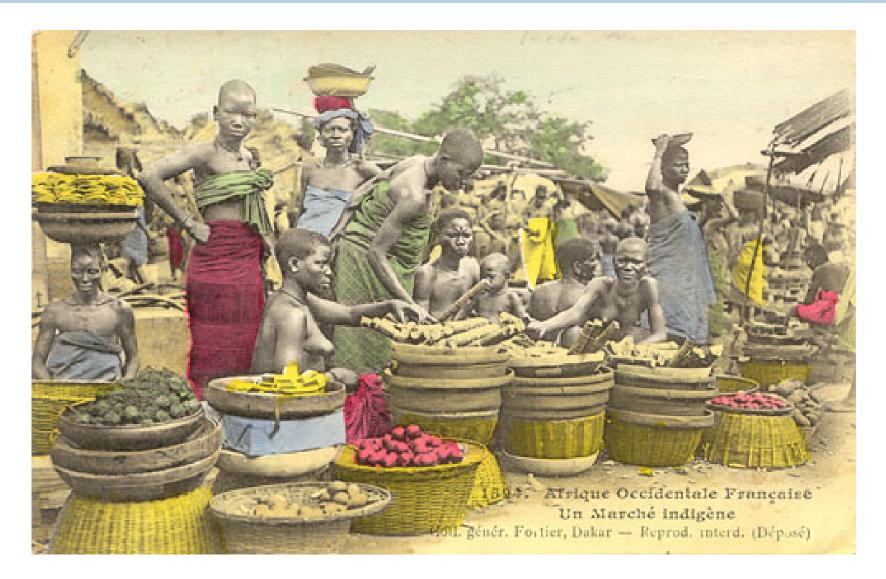


- GNM finds a (pure strategy) BNE in ~53s
- Gives us a price of anarchy lower bound: 1.068
- Try it out: https://github.com/davidrmthompson/positronic-economist

KUDU: A MOBILE MARKET FOR AGRICULTURAL COMMODITIES IN DEVELOPING COUNTRIES

[Ssekibuule, Quinn & LB, 2013]

African produce market circa 1900



Ugandan produce market circa 2011



Sometimes the scale is a bit bigger...



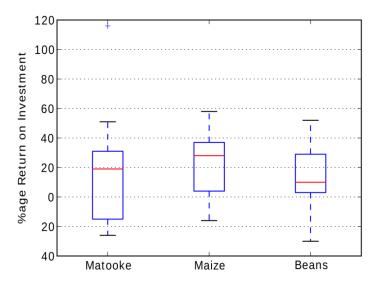
Sometimes the scale is a *lot* bigger...



Problem: Market Inefficiency

- Subsistence agriculture is the main occupation in Uganda
- Farmers waste a lot of time transporting produce; waiting by the road
- Buyers and sellers have trouble finding each other
- Sporadic food shortages in urban centers
- Robust arbitrage opportunities





The wave of the future?

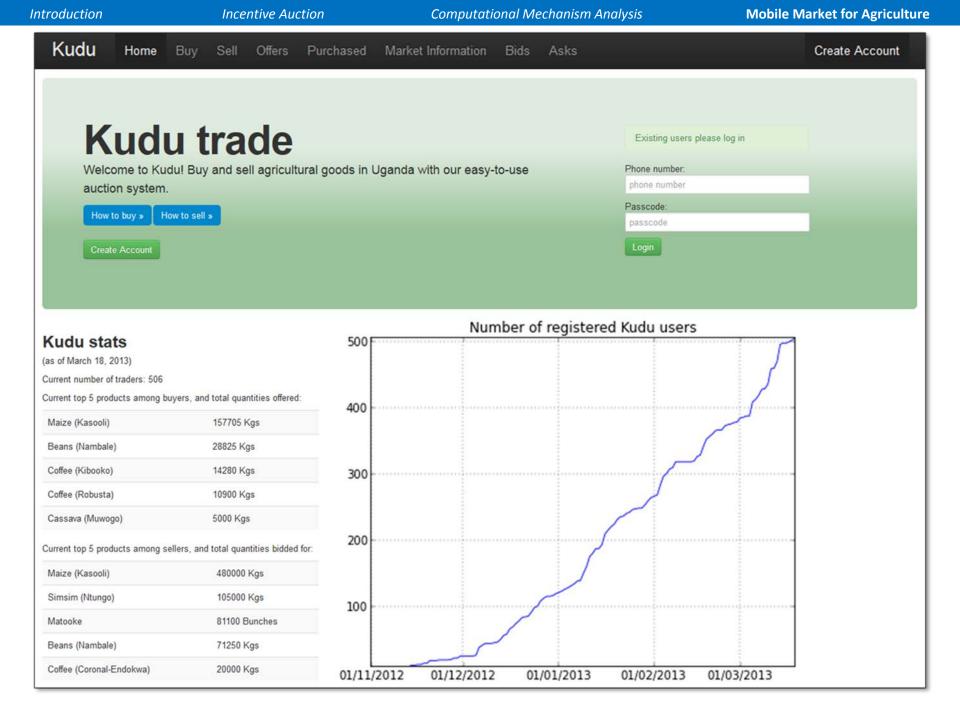


Kudu: an SMS-based market for agricultural commodities [Ssekibuule, Quinn & LB, 2013]

- **bids** consider price, reputation, quality, geographic location
- market clears daily
 - posted prices for farmers
 - second-pricing for buyers
- can ban specific traders

Really works!

- Field trial Jan July 2013
- http://www.kudu.ug



Statistics from our Field Trial

- 1024 traders and farmers registered
- 520 asks (USD \$1,700,000); 285 bids (USD \$960,000)
- 219 users used Kudu only to learn commodity prices
- Market activity highly dependent on radio adverts
- Largest bid: 120,000 Kg of maize (verified genuine)
 - 53 bids, 94 asks exceeding 10,000 Kg

Produce type	Total ask quantity
Peanuts	$512,375 { m ~Kg}$
Maize	$1,711,935 { m ~Kg}$
Beans mixed	$114,900 { m ~Kg}$
Coffee (Robusta)	$36,800 { m ~Kg}$
Sweet Potatoes	2,221 Sacks

Table 3: Quantities of the five categories of produce with the highest aggregate ask value.

Produce type	Total bid quantity
Maize	$917,\!300~{ m Kg}$
Sesame	$110,000 { m ~Kg}$
Beans mixed	$179,\!050~{ m Kg}$
Soya	$40,000~{ m Kg}$
Peanuts	$35,050~{ m Kg}$

Table 4: Quantities of the five categories of produce with the highest aggregate bid value.

Next Steps for Kudu

National scale

Introduction

- More work on quality, reputation
- More sophisticated matching

Incentive Auction

- Dealing with malformed bids
- Mostly, advertising and labor
- Looking for **funding**—ideas?



Pragmatic AGT

I argued for the benefits of pragmatic AGT, which

- 1. measures performance on specific **problems of interest**;
- 2. adopts statistical rather than analytical methods.

Today I told you about:

- **Spectrum repacking:** computational issues are at the heart of the FCC's upcoming radio spectrum redistribution. We're helping to build high-speed feasibility checkers and investigating novel auction designs.
- **Computational Mechanism Analysis:** can leverage compact game representations to enable quantitative, statistical analysis of existing mechanisms like ad auctions.
 - **Positronic Economist** will make it easier for you to use these tools.
- Kudu, an SMS-based market for agricultural commodities, leverages practical ideas from AGT to help Ugandan farmers.

Thanks to my students and collaborators!



- Alexandre Fréchette
 - -Spectrum repacking



- David R.M. Thompson
 - -Ad auctions; Positronic Economist



- Albert Xin Jiang
 - –AGGs



John Quinn, Richard Ssekibuule
 – Kudu