

Mechanism Design for Scheduling

Auction Protocols for Decentralized Scheduling

Wellman et al.

Elodie Fourquet

Electronic Market Design Presentation
School of Computer Science
University of Waterloo

November 22, 2004

Outline

- 1 Introduction
The Factory Scheduling Problem
- 2 Formal Model
Optimal Allocation & Equilibrium Solution Discussion
- 3 Ascending Auction (MM)
- 4 Combinatorial Auction (MM)
- 5 Generalized Vickery Auction (DRM)
- 6 Conclusions

Scheduling Problem Motivation

- Basic scheduling = hard problem
- Resource allocation problem

Scheduling Problem Motivation

- Basic scheduling = hard problem
- Resource allocation problem
- Essential to :
 - 1 computer science
 - 2 manufacturing & service industries

Scheduling Problem Motivation

- Basic scheduling = hard problem
- Resource allocation problem
- Essential to :
 - 1 computer science
 - 2 manufacturing & service industries
- In the Internet no time delivery guarantee

Scheduling Approaches

- 1 Distributed scheduling heuristics :
First-come first-served, priority-first, shortest-job-first

Scheduling Approaches

- 1 Distributed scheduling heuristics :
First-come first-served, priority-first, shortest-job-first
- 2 **Market mechanism** : price system
- 3 Direct revelation mechanism : GVA

An Application : Scheduling

- Goals
 - 1 Agents make effective decision
 - 2 **Pareto optimal solution** = resources are not wasted
 - 3 Reasonable communication, closure and computation

An Application : Scheduling

- Goals
 - 1 Agents make effective decision
 - 2 **Pareto optimal solution** = resources are not wasted
 - 3 Reasonable communication, closure and computation

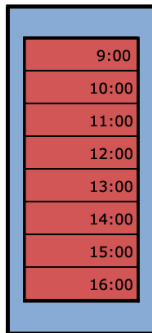
- Problems
 - 1 Equilibrium solution
 - 2 Sometimes hard problem = NP-complete
Discreteness & complementarity issues
 - 3 Combinatorial and Generalized Vickery Auction

An Application : Scheduling

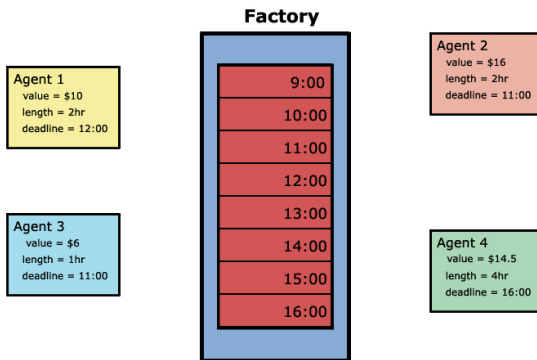
- Goals
 - 1 Agents make effective decision
 - 2 **Pareto optimal solution** = resources are not wasted
 - 3 Reasonable communication, closure and computation
- Problems
 - 1 Equilibrium solution
 - 2 Sometimes hard problem = NP-complete
Discreteness & complementarity issues
 - 3 Combinatorial and Generalized Vickery Auction
- Practical application of course theory

Factory Scheduling Example

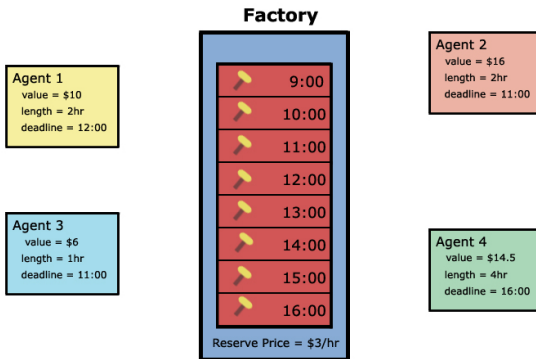
Factory



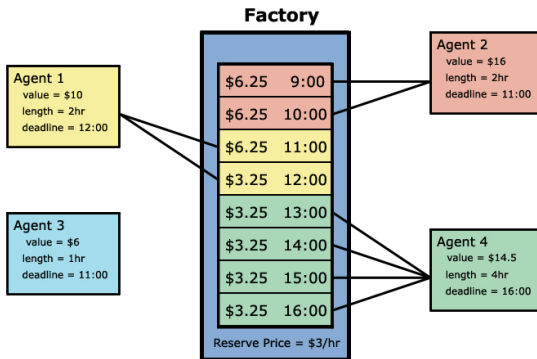
Agents' Jobs



Allocation with an Auction



Results



- Equilibrium solution
- Globally optimal allocation. Solution global value = \$40.5

Decentralized vs Centralized Scheduling

Decentralized	Each agent is self-interested
Decentralized	Each agent knows only private info
Decentralized	Each agent communicates relevant private info
<i>Decentralized</i>	<i>Market Mechanisms (MM) : AA and CA</i>

Decentralized vs Centralized Scheduling

Decentralized	Each agent is self-interested
Decentralized	Each agent knows only private info
Decentralized	Each agent communicates relevant private info
<i>Decentralized</i>	<i>Market Mechanisms (MM) : AA and CA</i>

Centralized	Every agent info is known
Centralized	Decision-maker controls resources
<i>Centralized</i>	<i>Direct Revelation Mechanism (DRM): GVA</i>

Decentralized vs Centralized Scheduling

Decentralized	Each agent is self-interested
Decentralized	Each agent knows only private info
Decentralized	Each agent communicates relevant private info
<i>Decentralized</i>	<i>Market Mechanisms (MM) : AA and CA</i>

Centralized	Every agent info is known
Centralized	Decision-maker controls resources
<i>Centralized</i>	<i>Direct Revelation Mechanism (DRM): GVA</i>

Decentralized vs Centralized information & computation

General Discrete Resource Allocation Problem

Definition

- G , a set of n discrete goods
- A , a set of m agents
- \perp , the seller
- $p = \langle p_1, \dots, p_n \rangle$, set of prices

General Discrete Resource Allocation Problem

Definition

- G , a set of n discrete goods
- A , a set of m agents
- \perp , the seller
- $p = \langle p_1, \dots, p_n \rangle$, set of prices

Valuations

- Agent j has utility $v_j(X)$ for holding set of goods X , $X \subseteq G$
- Seller has utility $q_i =$ reserve price, if good i is unallocated

Allocation Solution

A mapping, f , assigns discrete good to agents :

$$f : G \rightarrow A \cup \perp$$

Allocation Solution

A mapping, f , assigns discrete good to agents :

$$f : G \rightarrow A \cup \perp$$

	Allocated to agent j	Unallocated
Set of goods	$F_j \equiv \{i f(i) = j\}$	$F_{\perp} \equiv \{i f(i) = \perp\}$

Values Achievable

Maximum surplus value of agent j for holding set X at p

$$H_j(p) \equiv \max_{X \subseteq G} [v_j(X) - \sum_{i \in X} p_i]$$

Values Achievable

Maximum surplus value of agent j for holding set X at p

$$H_j(p) \equiv \max_{X \subseteq G} [v_j(X) - \sum_{i \in X} p_i]$$

Global value of solution f

Sum of agent values achieved + reserve value of goods not sold

$$v(f) \equiv \sum_{j=1}^m v_j(F_j) + \sum_{i \in F_{\perp}} q_i$$

Simple Scheduling

Definition

Each agent j has a job of :

- Length λ_j
- Deadlines $d_j^1 < \dots < d_j^{K_j}$
- Values $v_j^1 > \dots > v_j^{K_j}$

where $1 \leq K_j \leq n$, n total number of slots available

Several deadlines : higher values for earlier deadlines

Different Problems

1 Lengths of job :

Single-unit	$\lambda_j = 1$ for all j
Multiple-unit	$\lambda_j > 1$ for some j

2 Deadlines of job :

Fixed-deadline	$K_j = 1$ for all j
Variable-deadline	$K_j > 1$ for some j

Price Equilibrium

Definition

A solution f is in *equilibrium* at prices p iff :

- 1 All agents j get goods in allocation f that max his surplus at p

$$v_j(F_j) - \sum_{i \in F_j} p_i = H_j(p)$$

- 2 For all i , $p_i \geq q_i$
- 3 For all $i \in F_{\perp}$, $p_i = q_i$

Optimality of Equilibrium

Theorem

For the general discrete resource allocation problem, if there exists a p such that f is in equilibrium at p , then f is an optimal solution.

Optimality of Equilibrium

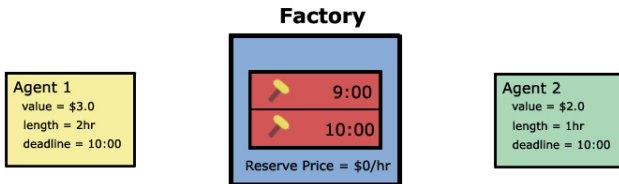
Theorem

For the general discrete resource allocation problem, if there exists a p such that f is in equilibrium at p , then f is an optimal solution.

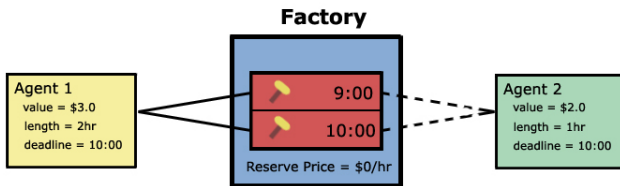
Proof (Main Idea).

Price forms a boundary between equilibrium and alternate solution. □

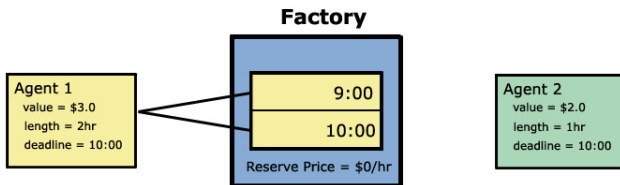
Agents' Jobs



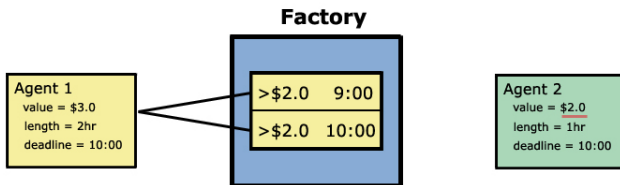
Agents' Interests



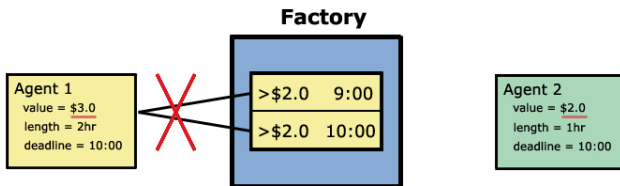
Optimal Solution



Price Equilibria Requirements



No Equilibrium Exists



Problem of **complementarities** in Agent1 preferences.

Equilibrium

- Single-unit scheduling problem always has at least one price equilibrium.
- But in general case, equilibrium may not exist.
- Single complementarity is sufficient to prevent a price equilibrium.

Market Mechanism Advantages

Considering decentralized scheduling :

- Markets are naturally decentralized
- Communication = exchange of bids & prices
- Mechanism can elicit info for Pareto & global optima
- Price is a common scale of value

Market Mechanism Advantages

Considering decentralized scheduling :

- Markets are naturally decentralized
- Communication = exchange of bids & prices
- Mechanism can elicit info for Pareto & global optima
- Price is a common scale of value
- Price system significantly simplifies resources allocation mechanism

Ascending Auction Protocol

Mechanism Bidding Rules

- *Bid price*, β_i = highest bid so far
- *Ask price*, $\alpha_i = \beta_i + \epsilon$ or q_i if undefined
- Agent must bid at least *ask price*

Agent Bidding Policies

Agent bids *ask prices* for the set of goods, maximizing his surplus.
No anticipation of other agents' strategies.

Ascending Auction Problems

- 1 Protocol may not find an equilibrium solution.

▶ AA Example 1

- 2 Protocol can produce a solution arbitrary far from optimal.

▶ AA Example 2

- 3 Protocol restricted to single-unit length job,
is still not guaranteed to reach equilibrium.

▶ AA Example 3

Incremental Auction Closing

- Sunk costs are considered
- Positive or negative effects on the solution
 - ▶ AAIC Example 1
- No effect for:
 - 1 Single-unit problem, no sunk costs
 - 2 If allocation represents a price equilibrium
- **Order** of reopening matters

Combinatorial Auction Needs

- Ascending auction mostly works well for single-unit problem.
- Ascending auction cannot always find existing equilibria in multiple-unit problem.

Combinatorial Auction Needs

- Ascending auction mostly works well for single-unit problem.
- Ascending auction cannot always find existing equilibria in multiple-unit problem.
- Combinatorial auctions help complementary issues.
But, computationally more complex.

Problem Allocation Reformulation

Definition

- G , a set of n discrete *basic goods*
- G' , a expanded set of *market goods*
 $good(y, z)$, denotes “bundle of y slots no later than slot z ”
- A , a set of m agents
- \perp , the seller
- P' , set of prices $p(y, z)$ for all market goods in G'

Scheduling Computational Tractability

Order

- No need to consider all 2^n combinations
- $\theta(l \cdot n)$ market goods in G' and prices in P'
where l is a bound on y , i.e. $y \leq l$ ($l \geq \max_{j \in A} \lambda_j$)
- Because additional structure (similar to Rothkopf et al. 1998)
Agents will want some number of slots before some deadline
- Goal is to preserve tractability

Market Good Allocation Solution

- A mapping, ϕ , assigns **market goods** to agents :

$$\phi : G' \rightarrow A \cup \perp$$

Market Good Allocation Solution

- A mapping, ϕ , assigns **market goods** to agents :

$$\phi : G' \rightarrow A \cup \perp$$

- Set of market goods allocated to agent j :

$$\Phi_j \equiv \{i | \phi(i) = j\}$$

Market Good Allocation Solution

- A mapping, ϕ , assigns **market goods** to agents :

$$\phi : G' \rightarrow A \cup \perp$$

- Set of market goods allocated to agent j :

$$\Phi_j \equiv \{i | \phi(i) = j\}$$

- A market allocation ϕ is **consistent with** a solution f if f gives each agent what is promised by ϕ

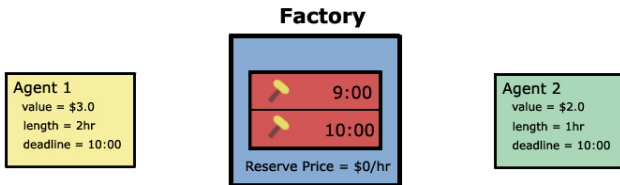
Combinatorial Price Equilibrium

Definition

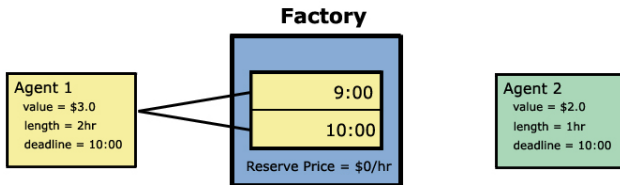
A solution ϕ is in *equilibrium* at prices p iff :

- 1 For all agent j , ϕ_j maximizes j 's guaranteed surplus at p
- 2 Market good price at least min consistent reserve price.
For all (y, z) , $p(y, z) \geq \min_B \sum_{i \in B} q_i$
- 3 There exists an implementing solution f , consistent with ϕ s.t.
 - 1 Allocated market good price \geq sum of basic good prices comprising market good in f
 - 2 When market good could be satisfied by basic goods unallocated, reserve prices of those goods define an upper bound on its price

Agents' Jobs



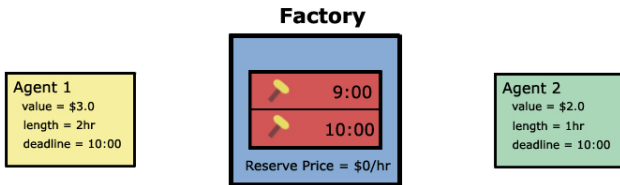
Optimal Solution



Combinatorial Auction



Combinatorial Auction



- Consider $l = 2$, $p(1, 9 : 00) = p(1, 10 : 00) = 2.1$ and $p(2, 10 : 00) = 2.9$

Combinatorial Auction



- Consider $l = 2$, $p(1, 9 : 00) = p(1, 10 : 00) = 2.1$ and $p(2, 10 : 00) = 2.9$
- Computed allocation $\Phi_1 = \{(2, 10 : 00)\}$, $\Phi_2 = \emptyset$
Satisfies combinatorial equilibrium conditions.

Optimal and Equilibrium

- Combinatorial equilibrium prices can support :
 - 1 Optimal solution
 - 2 But also non-optimal solution.
- Sub-optimality is not usefully bounded -even without reserve prices.
- Optimal solution supported by equilibria in original formulation are retained in the combinatorial one.
- Given monotone reserve prices, optimal solution can be supported with $\theta(l \cdot n)$ price system.

Generalized Vickery Auction

- Neither ascending nor combinatorial auction guarantee optimal solution to scheduling problem.

Generalized Vickery Auction

- Neither ascending nor combinatorial auction guarantee optimal solution to scheduling problem.
- GVA finds efficient schedules for all our scheduling problem.
- GVA is a direct revelation mechanism :
 - GVA is not a price system.
 - Rather GVA computes overall payments for agents' allocations.

VGA Protocol

Mechanism Bidding Rules

- Each agent j announces his alleged utility function \tilde{v}_j .
Not constrained to be truthful.
- Auction knows the reserve values, q_j .

Allocation Rules and Optimality

After receiving bids, GVA returns :

- 1 Allocation solution f^* ,
- 2 Payments to agents.

VGA Payments

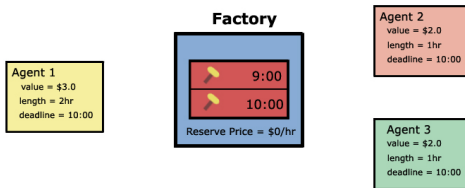
- Payments to agent j :

$$V_{-j} \equiv W_{-j}(f^*) - P_j(\tilde{v}_j)$$

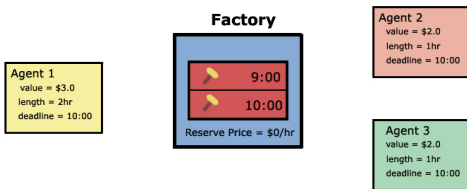
where :

- 1 W_{-j} = agents' total reported value at f^* , excluding j
 - 2 P_j = residual payment (function of other agent's reported valuations)
- Payments force truthful bidding as a dominant strategy. Optimal allocation is computed on truthful bids, therefore allocation is globally optimal.

VGA

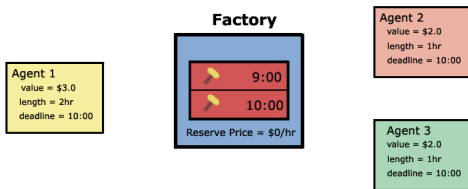


VGA



- Mechanism finds optimal solution : $f^*(9 : 00) = 2$ and $f^*(10 : 00) = 3$

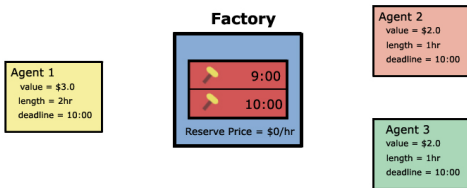
VGA



- Mechanism finds optimal solution : $f^*(9 : 00) = 2$ and $f^*(10 : 00) = 3$

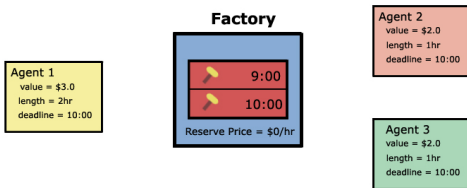
j	Agent 1	Agent 2	Agent3
W_{-j}	4	2	2
$v_j(F_j) + V_{-j}$	$0 + [4 - P_1]$	$2 + [2 - P_2]$	$2 + [2 - P_3]$

VGA



- For participation, received total value $v_j(F_j) + V_{-j} \geq 0$
 $P_j \leq 4$ for $j \in \{1, 2, 3\}$

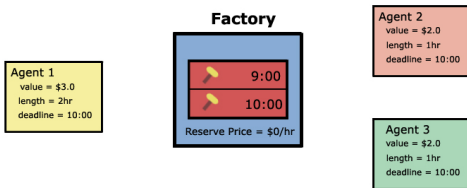
VGA



- For participation, received total value $v_j(F_j) + V_{-j} \geq 0$
 $P_j \leq 4$ for $j \in \{1, 2, 3\}$

j	Agent 1	Agent 2	Agent 3
$v_j(F_j) + V_{-j}$	$0 + [4 - P_1]$	$2 + [2 - P_2]$	$2 + [2 - P_3]$
P_j	4 (pays 0)	3 (pays 1)	3 (pays 1)

VGA



- For participation, received total value $v_j(F_j) + V_{-j} \geq 0$
 $P_j \leq 4$ for $j \in \{1, 2, 3\}$

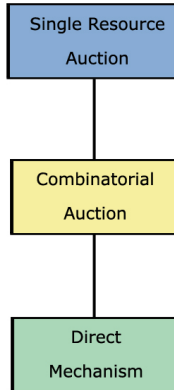
j	Agent 1	Agent 2	Agent 3
$v_j(F_j) + V_{-j}$	$0 + [4 - P_1]$	$2 + [2 - P_2]$	$2 + [2 - P_3]$
P_j	4 (pays 0)	3 (pays 1)	3 (pays 1)

- Net revenue \$2.0

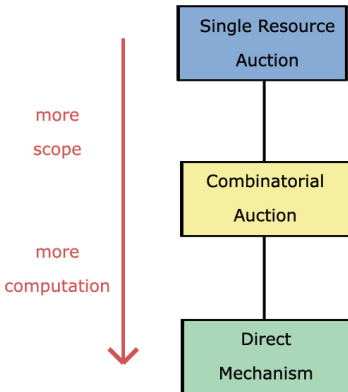
Performance

- Single-unit, fixed-deadline has optimal solution
Greedy algorithm running in $\theta(m \lg m)$
- VGA mechanism must solve multiple optimization problems :
 - 1 One to determine optimal solution
 - 2 One for each agent j with his bids removed to find P_jTherefore VGA adds a factor of m to the computation
- Single-unit, fixed-deadline has optimal VGA solution
With preference revelation needs $\theta(m^2 \lg m)$
- Multiple-unit scheduling problem is NP-complete

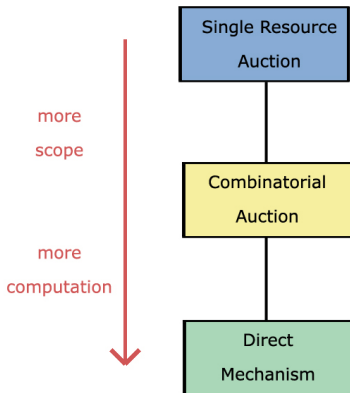
Scope and Computation Tradeoffs



Scope and Computation Tradeoffs



Scope and Computation Tradeoffs



But there exists more scheduling problems,
If we have time, for example....

Online Real-time Scheduling Problem

- Online scheduling of jobs on a single processor
Online = not all jobs are known in advance
- Jobs are owned by separate, self-interested agents
 - 1 Decide when to submit job after true release time
 - 2 Can inflate job's length
 - 3 Can declare arbitrary value and deadline for job
- Strategic agent can manipulate the system by announcing false characteristics of job, if beneficial for its completion
- Sellers schedule jobs and determine amount to charge to buyers

Online Real-time Scheduling Goals

- 1 Schedule needs to be constructed in real-time
- 2 Maximizing sum of job's values completed on time
- 3 Online algorithm needs to compare well against the optimal offline one
- 4 Preemption of a running job by a newly arrived job is possible

Online Real-time Scheduling Direct Mechanism

- Input : job declared by each agent
- Output : schedule and payment to be made by each agent to mechanism
- Goal = incentive compatibility
Agent's best interests :
 - 1 To submit job upon release
 - 2 To declare truthfully value, length and deadline of job
- Approximate solutions compare well with offline solutions

To Take Home

- Scheduling is important
- Many types of scheduling problem exist
- Most scheduling problems are hard, and most often NP-complete
- Price systems and auctions are a promising new approach for multiple scheduling problems
- Auction mechanisms encourage truth revelation about jobs
Crucial for distributed scheduling

Questions ?

Decentralized Scheduling Problem

AA may not find equilibrium solution

AA arbitrary far from optimal

AA single-unit may not find equilibrium solution

AAIC may do better

Challenges

Challenges

- 1 Message passing / closure / final schedule determination
Protocol problem : asynchronous communication

Challenges

- 1 Message passing / closure / final schedule determination
Protocol problem : asynchronous communication
- 2 Appropriate messages elicited
Mechanism design problem : socially desirable outcome

Combinatorial Price Equilibrium

Definition

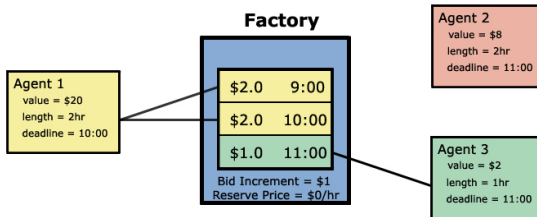
A solution ϕ is in *equilibrium* at prices p iff :

- 1 For all agent j , ϕ_j maximizes j 's guaranteed surplus at p
- 2 For all (y, z) , $p(y, z) \geq \min_{\{B \subseteq G_z : |B|=y\}} \sum_{i \in B} q_i$
- 3 There exists an implementing solution f s.t.
 - 1 For all j , $\sum_{(y,z) \in \phi_j} p(y, z) \geq \sum_{i \in F_j} q_i$
 - 2 For all "unallocated (y, z) ", $p(y, z) \leq \min_B \sum_{i \in B} q_i$

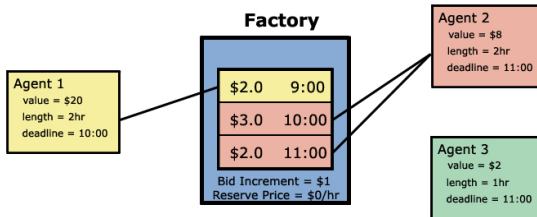
Agent jobs



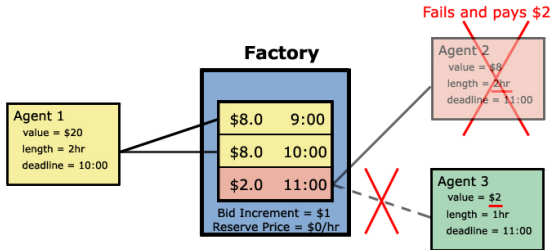
Bids



Bids

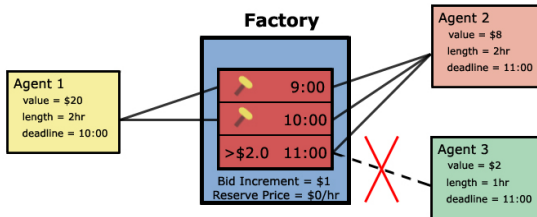


Bids

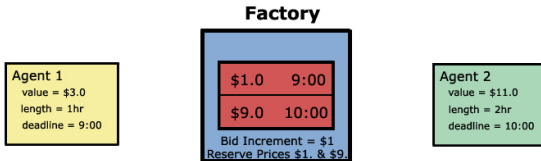


- Agent 2 wins slot 3 but cannot complete his job
- Agent 3 cannot get slot 3, $p_3 > 2$ blocked by Agent 2
- Not an optimal solution. Solution global value = \$20.0

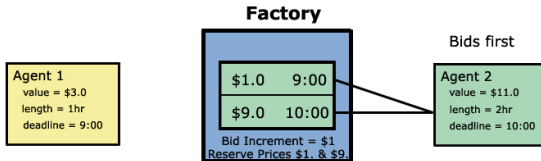
Problem



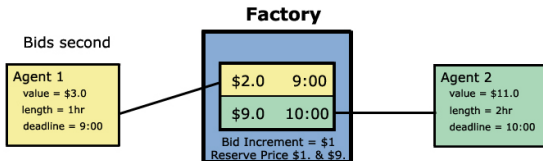
Agent jobs



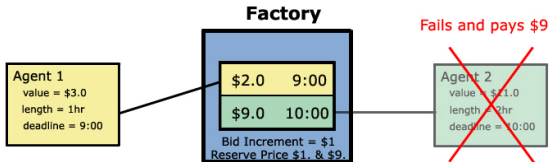
A2 Bids First



A1 Bids Second

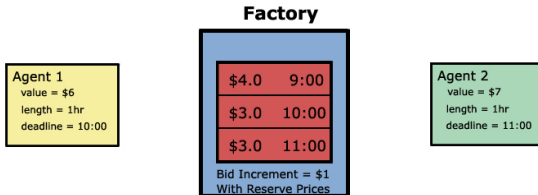


Allocation

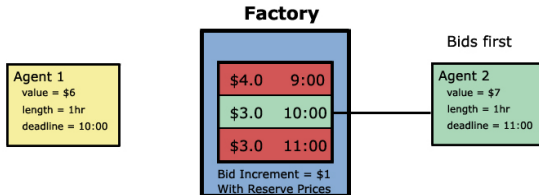


- Agent 2 wins slot 2 but cannot complete his job
- Solution global value = \$3.0

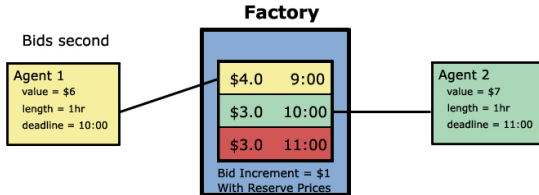
Agents' jobs



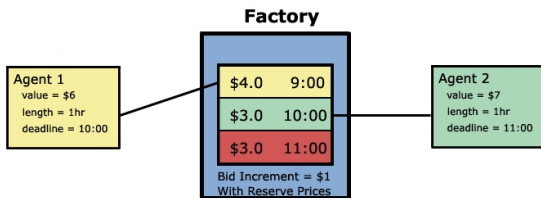
A2 Bids First



A1 Bids Second

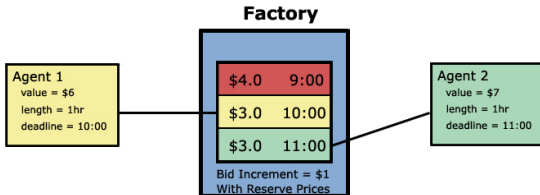


Allocation

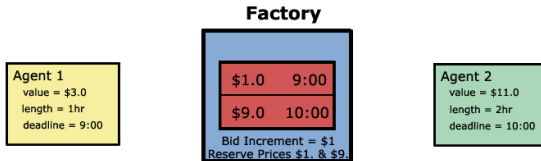


- But $p_2 = \$3 < p_1$ not an equilibrium
- Agent 1 would maximize his surplus by demanding p_2 at the final prices

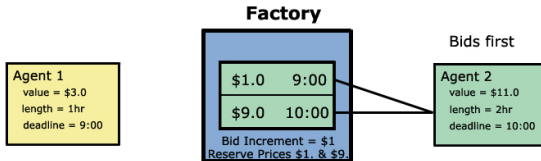
Equilibrium Solution

[Return](#)

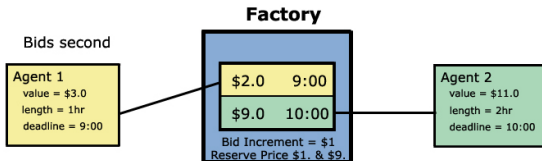
Agents' jobs



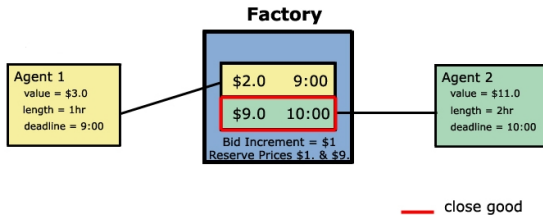
A2 Bids First



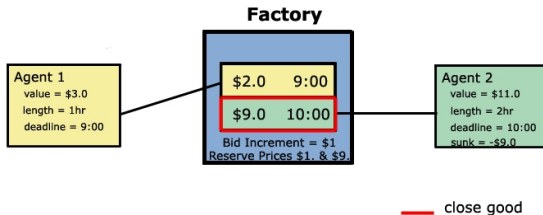
A1 Bids Second



Auction Closed for Slot 2



Agent 2 sunk cost



- Agent 2 treats his payment as sunk, and value slot 1 at \$11

