

# Equilibria in Electric Power Exchange Auction Markets

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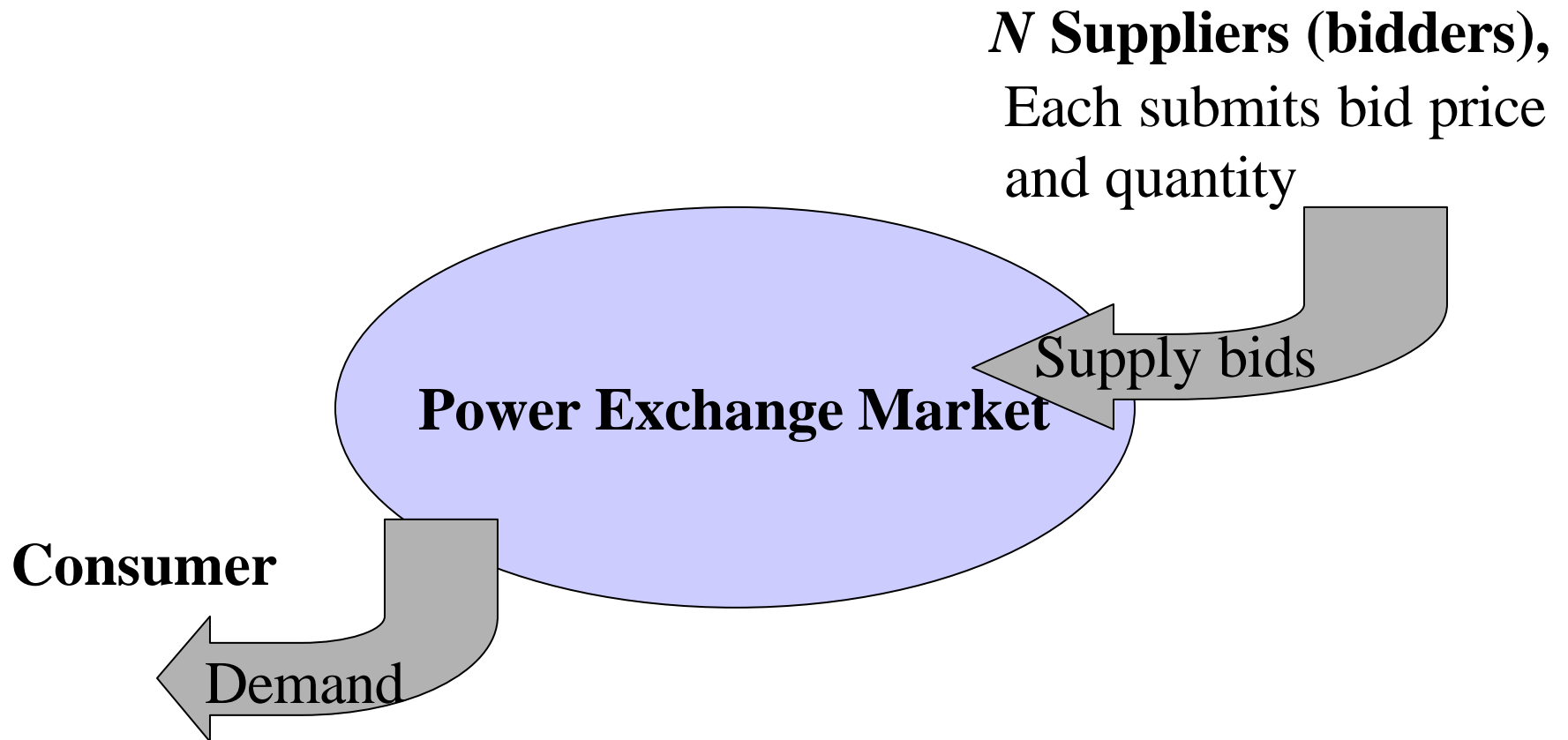
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# Outline

- I. Problem Overview
- II. Model
- III. Results
- V. Summary and Extensions

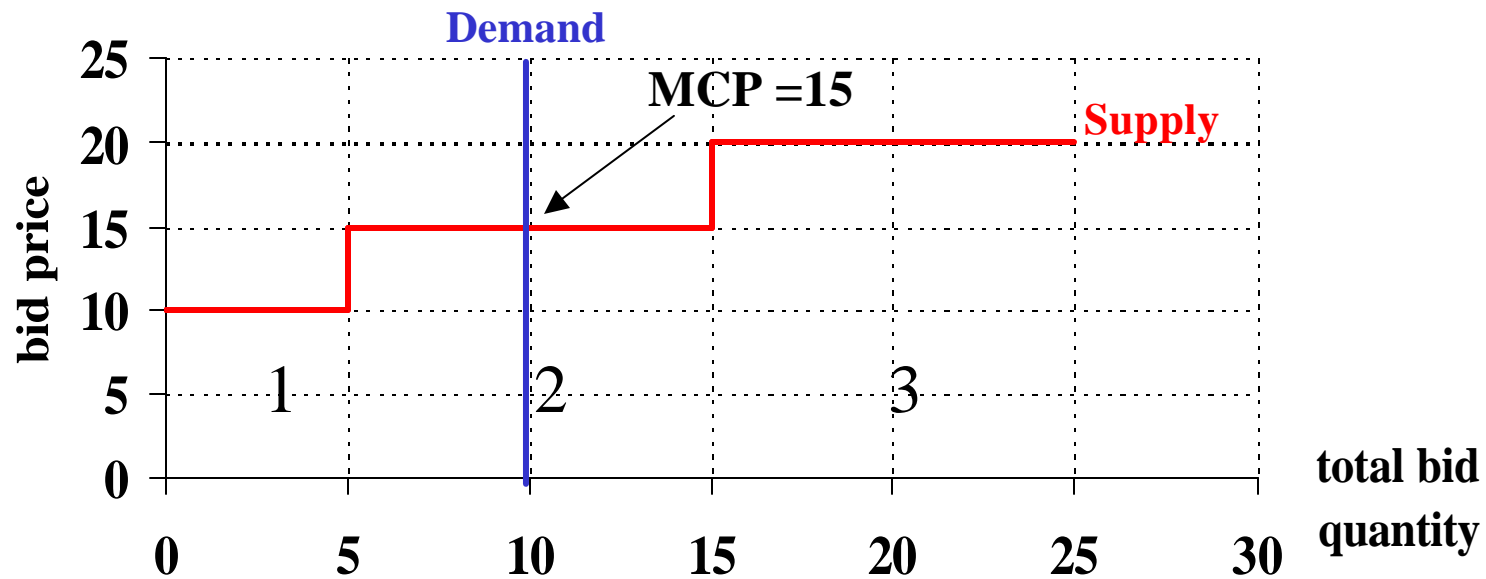
# Competitive Markets



# Market Clearing Process

Supplier 1 : 5MWh @ \$10  
Supplier 2 : 10MWh @ \$15  
Supplier 3 : 10MWh @ \$20

Demand is 10



**Problem:** finding optimal bidding strategies and the resulting MCP

# Market Overview

- **Non-sealed bid, Multi-round**
  - Bidders can see each other's bids and can adjust their prices as many times as they want
  - Market is closed when no bidder wants to adjust his/her bid price
- **Selling at spot:**
  - All dispatched units are traded at the same price

# Model

- Demand  $D$  (or  $d$ ) must be satisfied
- Bidder  $i$  has unit cost  $c_i$
- Single bid per bidder :  $(x_i, p_i)$ 
  - $x_i$  = bid quantity of bidder  $i$
  - $p_i$  = bid price of bidder  $i$  : (assume discrete)  
$$p_i \in \{l\mathbf{e} \mid l = 0,1,\dots,O\}, \quad i = 1,\dots,N$$
  - $\mathbf{b} = [(x_1, p_1), \dots, (x_N, p_N)]$

# Model

- Market clearing price

$$MCP(\mathbf{b}, d) = \left\{ \min_j p_j : \sum_{i \in I(j)} x_i \geq d, I(j) = \{i : p_i \leq p_j\} \right\}$$

- Dispatch quantity of Bidder  $i$

$$q_i(\mathbf{b}, d) = \begin{cases} 0 & \text{if } p_i > MCP(\mathbf{b}, d) \\ x_i & \text{if } p_i < MCP(\mathbf{b}, d) \\ \bar{q}_i(\mathbf{b}, d) & \text{if } p_i = MCP(\mathbf{b}, d) \end{cases}$$

\* Dispatch of marginal bidders follows order determined from submission time of bids

# Model

- Bidder  $i$ 's payoff

$$f_i(\mathbf{b}) = E_D [(MCP(\mathbf{b}, D) - c_i) q_i(\mathbf{b}, D)]$$

- Objective: Find Nash equilibrium  $\{p_i^*, i = 1, \dots, N\}$  such that

$$f_i \left( \left[ (x_1, p_1^*), \dots, (x_i, p_i), \dots, (x_N, p_N^*) \right] \right) \leq f_i \left( \left[ (x_1, p_1^*), \dots, (x_i, p_i^*), \dots, (x_N, p_N^*) \right] \right)$$

for all feasible  $p_i$ , for bidder  $i$ , and all  $i = 1, \dots, N$

# Model

- Distinct bidders:

$$|c_i - c_j| > 2\epsilon, \quad \forall i \neq j$$

- Fixed bid quantity: bidders can adjust only bid price
- Given a number  $x$ ,

$$\lfloor x \rfloor = \max\{\epsilon^i / \epsilon^i \leq x, i=0, \dots, O\} \text{ and}$$

$$\lceil x \rceil = \min\{\epsilon^i / \epsilon^i \geq x, i=0, \dots, O\}$$

# Market Stability Condition

$$\bar{D} \leq \sum_{\forall i \neq j} x_i \quad \text{for } j = 1, \dots, N$$

- $\bar{D}$  is the highest demand realization
- No one is guaranteed to be dispatched

# Results

- Multiple equilibria
- Known demand:
  - At the highest MCP equilibrium point, every bidder  $i$  bids at  $\lceil c_i \rceil$ , except the marginal bidder  $i$  who bids at  $\lfloor c_j \rfloor$
  - Unique marginal bidder if partially dispatched
- Stochastic demand:
  - Single marginal  $i$ 
    - At any demand, bids at  $\lceil p_j - \epsilon \rceil$
    - At the highest demand, bids at  $\lfloor c_j \rfloor$
  - Two marginal bidders,  $i, j$ 
    - They bid just above the cost of the bidder with a lower quantity  
 $p_i = p_j = \lceil c_j \rceil$  where  $x_j \leq x_i$

# Highest Equilibrium MCP

- Known demand: the highest possible equilibrium MCP must be in the set

$$\{\lfloor c_i \rfloor, i = 1, \dots, N\}$$

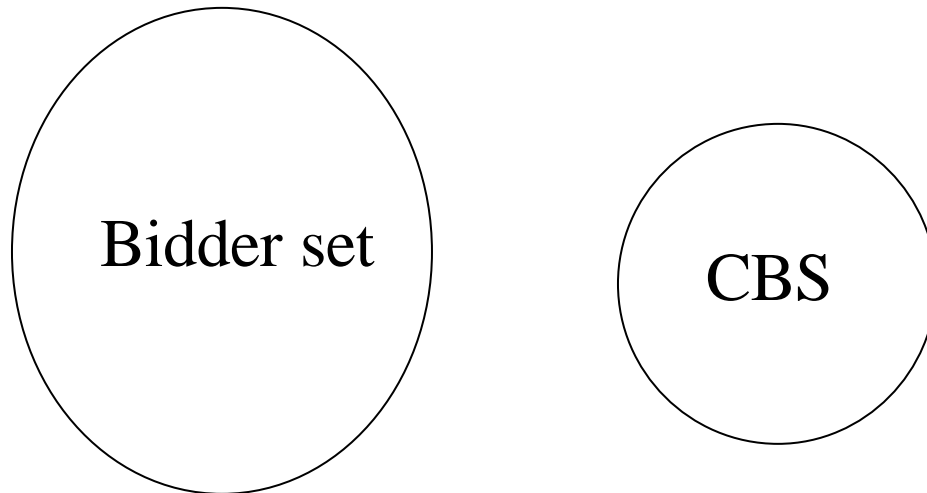
- Stochastic demand: the highest possible equilibrium MCP must be in

$$\{\lfloor c_i \rfloor, \lceil c_i \rceil, \lceil c_i \rceil + \mathbf{e}, i = 1, \dots, N\}$$

# Competitive Bidder Set (CBS)

- CBS: bidders with the lowest costs and satisfy the market stability condition

$$\bar{D} \leq \sum_{\forall i \neq j} x_i \quad \text{for } j = 1, \dots, N$$



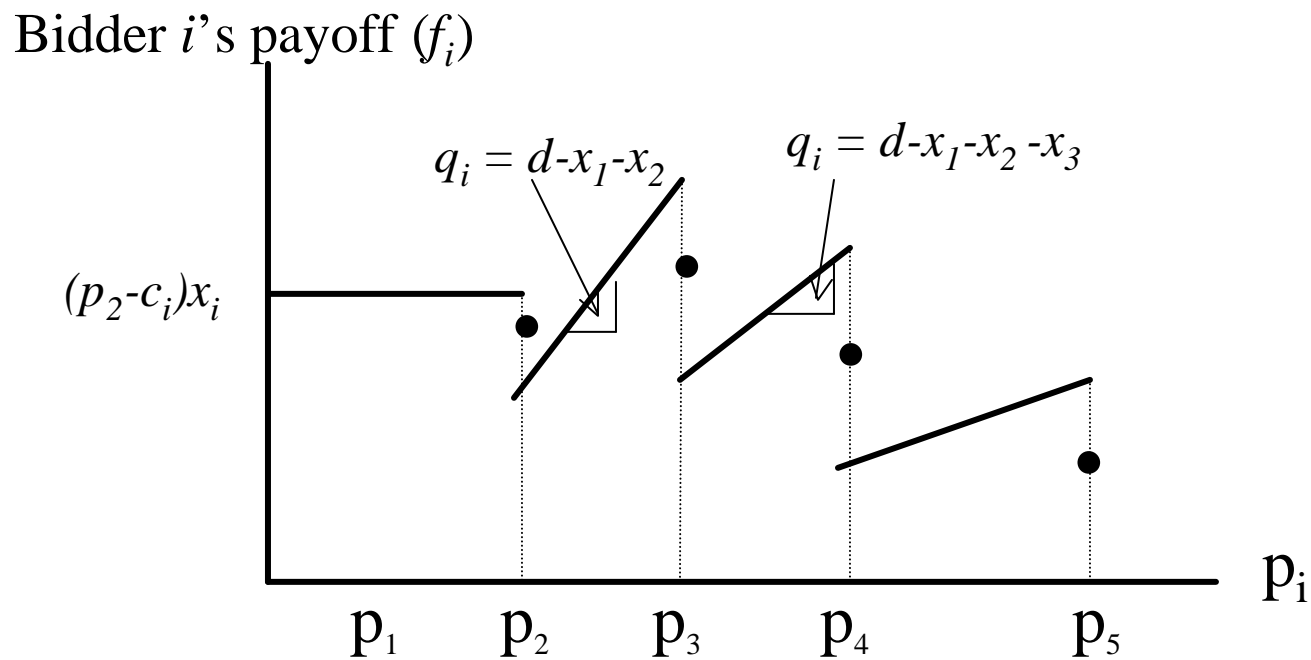
# Competitive Bidder Set (CBS)

At an equilibrium point,

- All bidders outside the CBS are not dispatched
- When demand is known, at least one bidder in the CBS is not dispatched
- $2\varepsilon$  plus the highest cost among the bidders in the CBS is an upper bound on the MCP

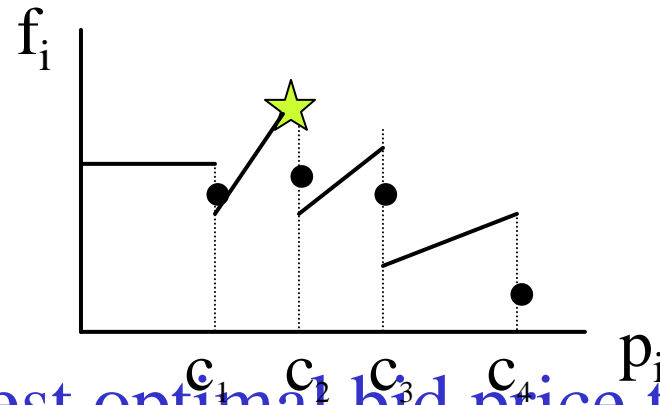
# Payoff function

- Given other bidders' bid prices and demand



# Algorithm for Finding the Highest MCP Equilibrium Point with Deterministic Demand

- Constructing CBS
- Condition on each bidder to be a marginal while others bid at cost
- Find the optimal bid price



- Pick the one with the highest optimal bid price to be the marginal bidder; others bid at costs

## A Special Case: Identical Quantity

- Assume  $x_i = x$  for all  $i = 1, \dots, N$
- Equilibrium point with the highest MCP

$$p_i^* = \lfloor c_{i+1} \rfloor \quad \forall i = 1, \dots, N-1$$
$$MCP = \lfloor c_k \rfloor$$

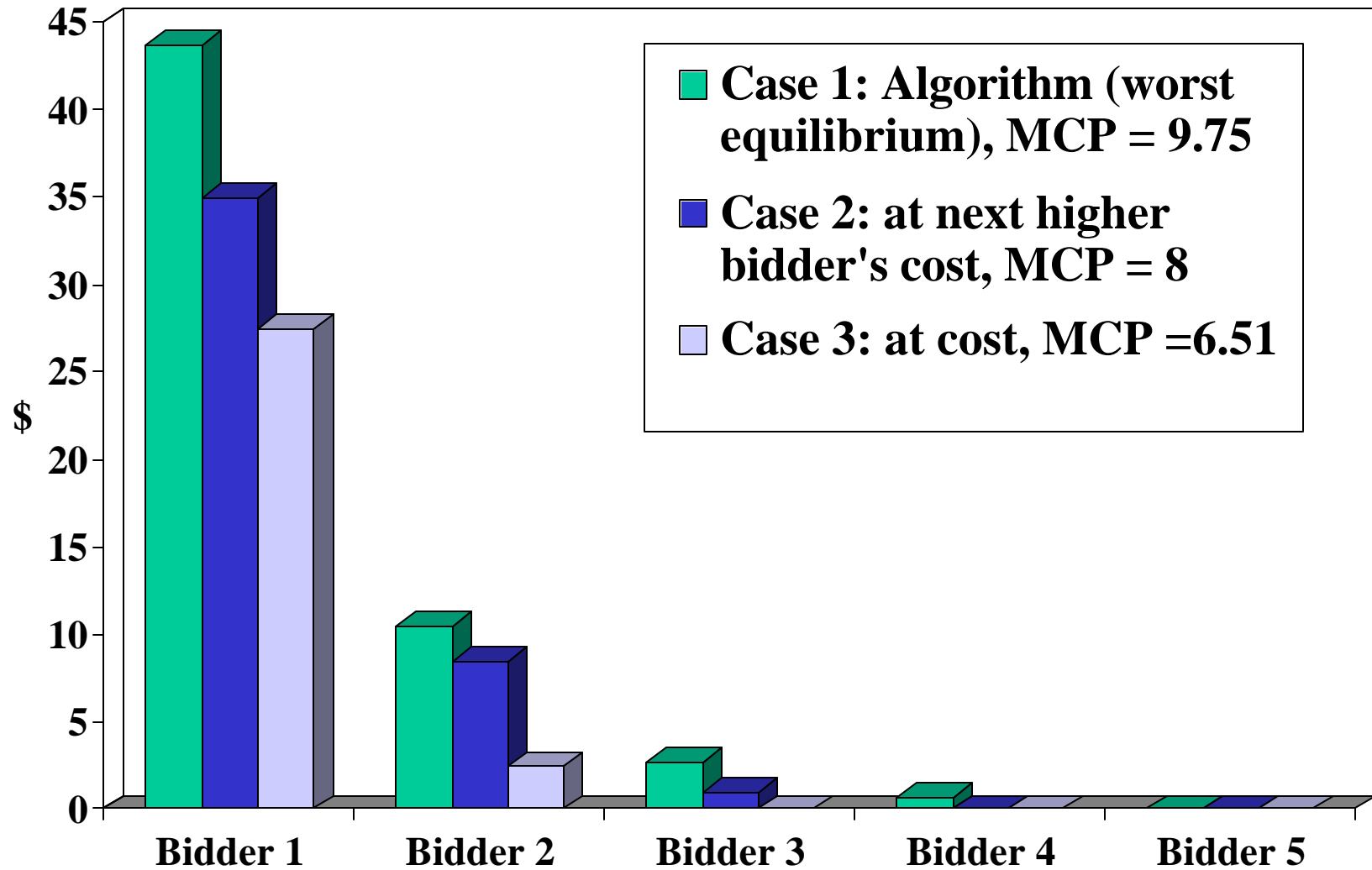
where  $k$  is the first undispached bidder, given all bidders bid at their costs.

# A Numerical Example

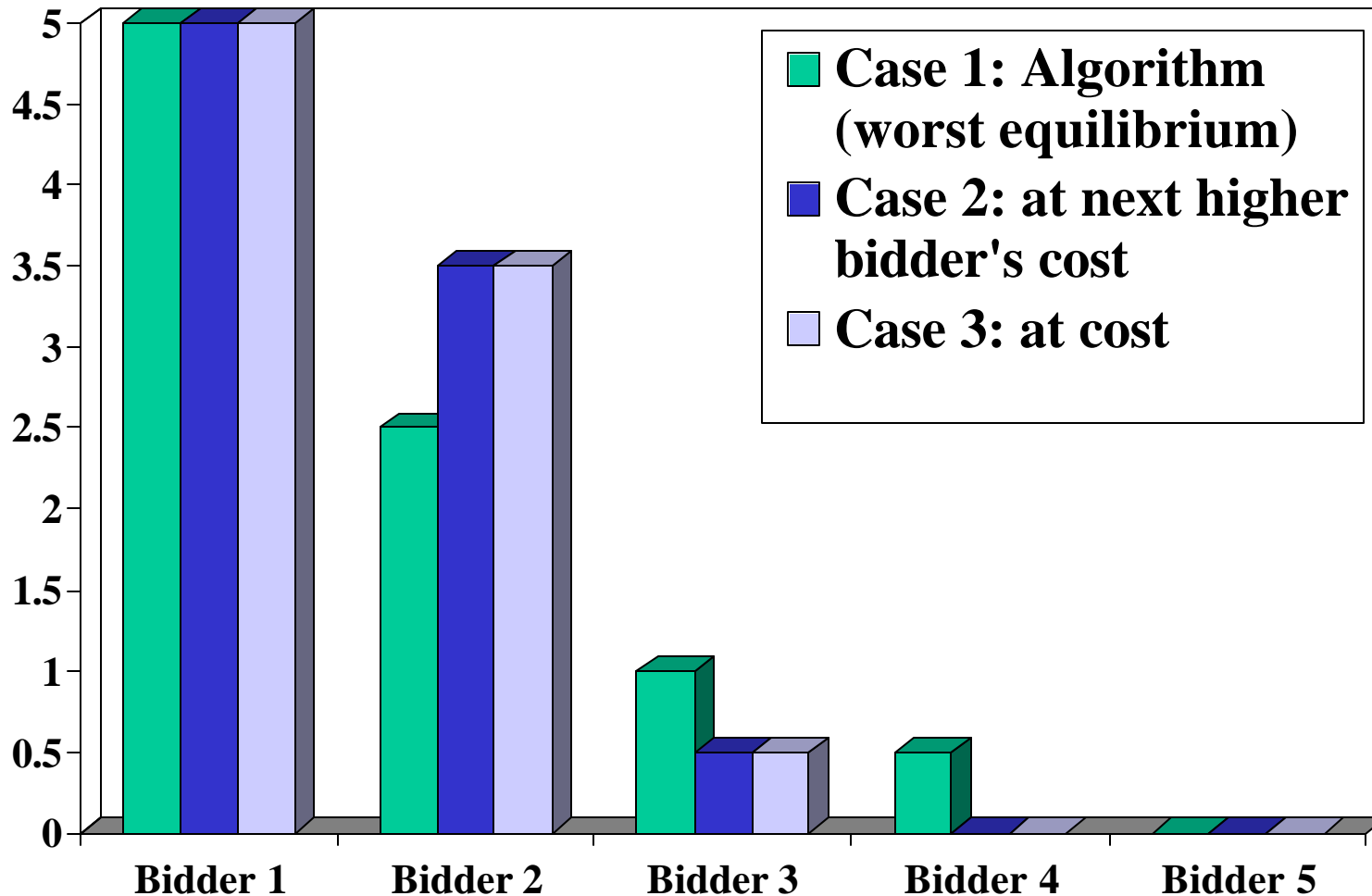
$i$	$C_i$	$x_i$
1	1.01	5
2	6.01	5
3	7.01	1
4	9.01	1
5	10.51	11

- $e = 0.01$ ,  $D = 7$  or  $11$  w.p.  $0.5$

# Comparison of Payoff



# Comparison of Dispatch Quantity



# Other Problems-Price Down as Demand Rises

## Demand

- Period 1 - demand=50
- Period 2 - demand=100 or 200 equally likely

## Capacities

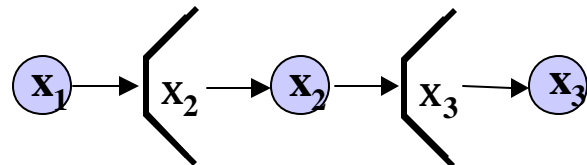
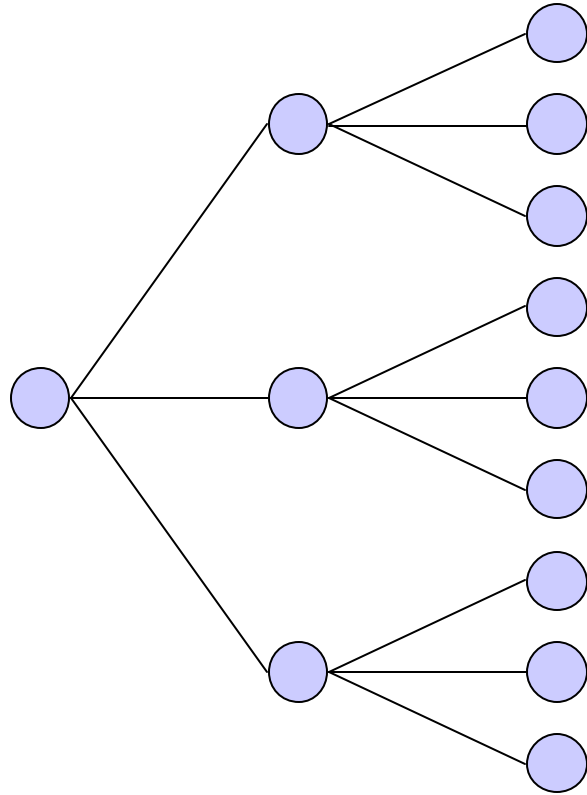
- Hydro - 100 total
- Thermal - 60 at once
- Backstop -  $\infty$

## Suppose Period 2 Demand = 100

- Hydro - Bid only in Period 2, 100 at  $5 - \epsilon$
- Thermal - Bid 5, Backstop - Bid 50

# General Equilibrium Solution: Multistage Stochastic Linear Program

Stage 1      Stage 2      Stage 3



$$\min \quad c_1 x_1 + Q_2(x_1)$$

$$s.t. \quad W_1 x_1 = h_1$$

$$x_1 \geq 0$$

$$Q_t(x_{t-1,a(k)}) = \sum_{\mathbf{x}_{t,k} \in \Xi_t} \text{prob}(\mathbf{x}_{t,k}) Q_{t,k}(x_{t-1,a(k)}, \mathbf{x}_{t,k})$$

$$Q_{t,k}(x_{t-1,a(k)}, \mathbf{x}_{t,k}) = \min \quad c_t(\mathbf{x}_{t,k}) x_{t,k} + Q_{t+1}(x_{t,k})$$

$$s.t. \quad W_t x_{t,k} = h_t(\mathbf{x}_{t,k}) - T_{t-1}(\mathbf{x}_{t,k}) x_{t-1,a(k)}$$

$$x_{t,k} \geq 0$$

- $Q_{N+1}(x_N) = 0$ , for all  $x_N$ ,
- $Q_{t,k}(x_{t-1,a(k)})$  is a piecewise linear, convex function of  $x_{t-1,a(k)}$

# Nested Decomposition

- In each subproblem, replace expected recourse function  $Q_{t,k}(x_{t-1,a(k)})$  with unrestricted variable  $\mathbf{q}_{t,k}$

– Forward Pass:

- Starting at the root node and proceeding forward through the scenario tree, solve each node subproblem

$$\hat{Q}_{t,k}(x_{t-1,a(k)}, \mathbf{x}_{t,k}) = \min_{\mathbf{x}_{t,k}} c_t(\mathbf{x}_{t,k})x_{t,k} + \mathbf{q}_{t,k}$$

$$s.t. \quad W_t \mathbf{x}_{t,k} = h_t(\mathbf{x}_{t,k}) - T_{t-1}(\mathbf{x}_{t,k})x_{t-1,a(k)}$$

$$E_{t,k} \mathbf{x}_{t,k} + \mathbf{q}_{t,k} \geq e_{t,k} \quad (\text{optimality cuts})$$

$$D_{t,k} \mathbf{x}_{t,k} \geq d_{t,k} \quad (\text{feasibility cuts})$$

$$x_{t,k} \geq 0$$

- Add feasibility cuts as infeasibilities arise

– Backward Pass

- Starting in top node of Stage  $t = N-1$ , use optimal dual values in descendant Stage  $t+1$  nodes to construct new optimality cut. Repeat for all nodes in Stage  $t$ , resolve all Stage  $t$  nodes, then  $t-1$ .

– Convergence achieved when



# Pereira-Pinto Method

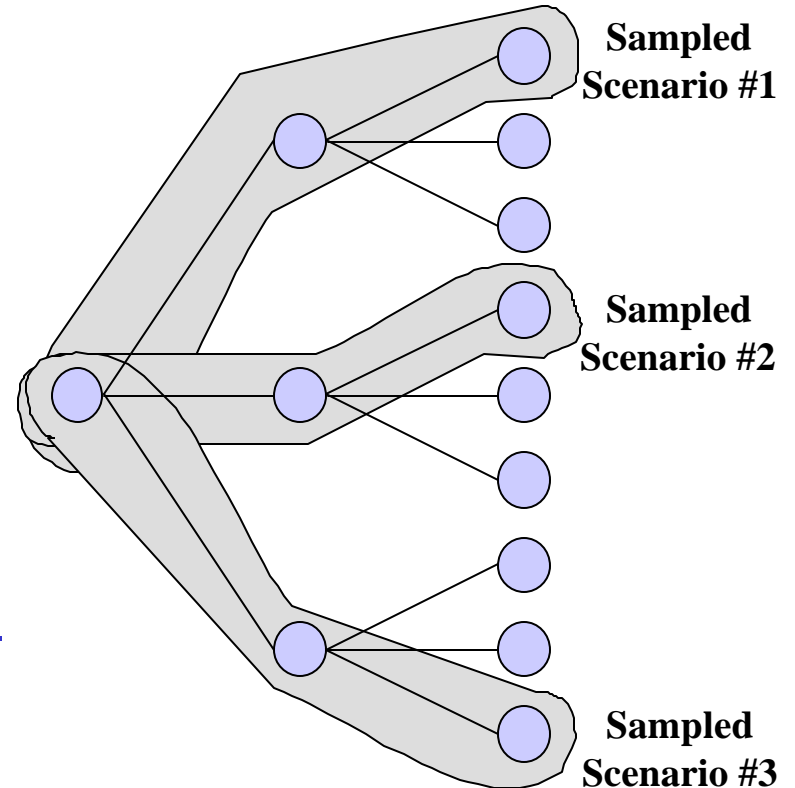
1. Randomly select  $H$   $N$ -Stage scenarios
2. Starting at the root, a forward pass is made through the sampled portion of the scenario tree (solving ND subproblems)

3. A statistical estimate of the first stage objective value  $\bar{z}$  is calculated using the total objective value obtained in each sampled scenario

the algorithm terminates if current first stage objective value  $c_1x_1 + \bar{q}_1$  is within a specified confidence interval of

4. Starting in sampled node of Stage  $t = N-1$ , solve all Stage  $t+1$  descendant nodes and construct new optimality cut.

Repeat for all sampled nodes in Stage  $t$ , then repeat for  $t = t - 1$



# Abridged Nested Decomposition

- Also incorporates sampling into the general framework of Nested Decomposition
- Also assumes relatively complete recourse and serial independence
- Samples both the subproblems to solve and the solutions to continue from in the forward pass

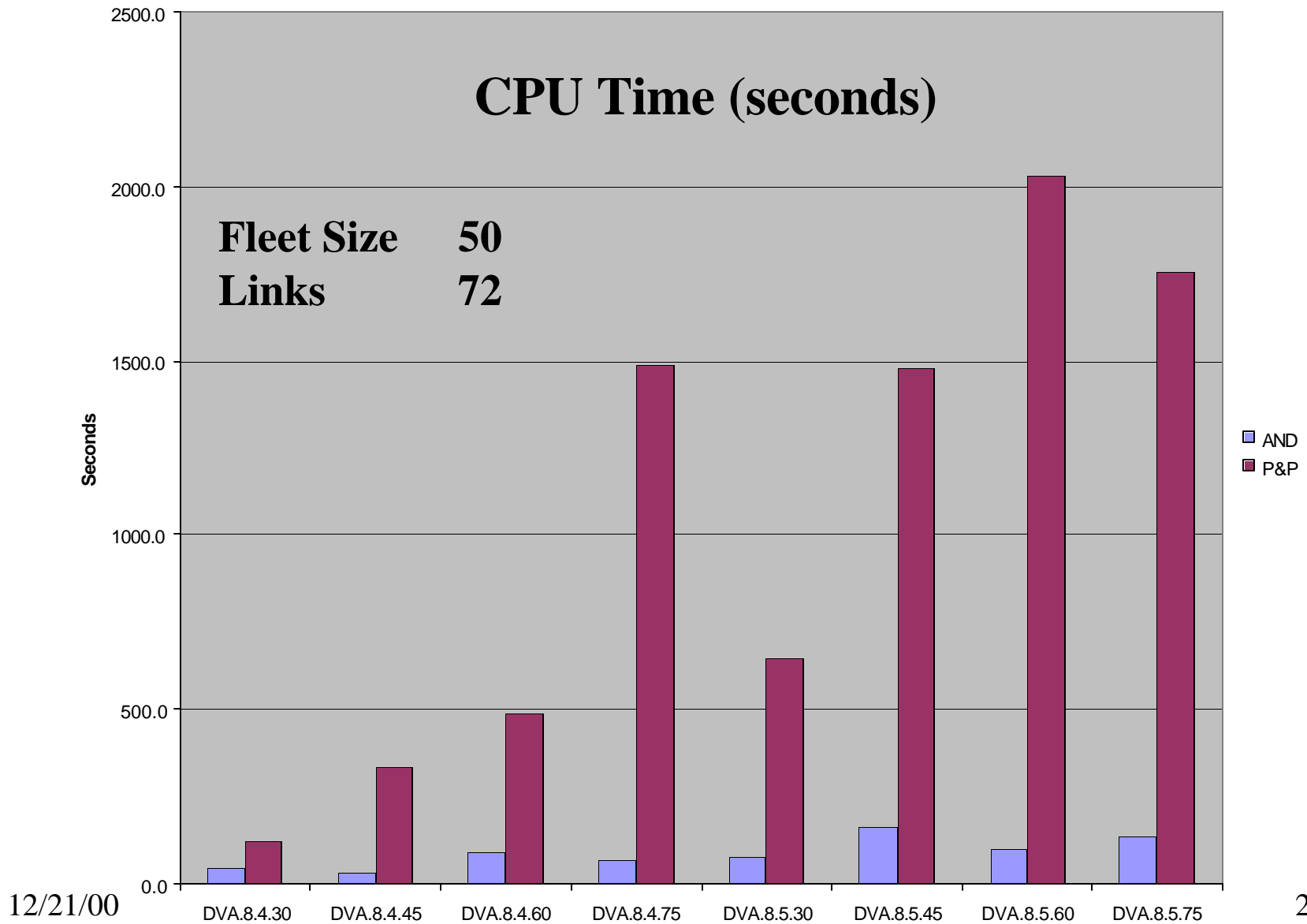
# Computational Results

- Implementation of Pereira & Pinto Method and Abridged Nested Decomposition
  - written in C, run on Sun SPARC 20 workstation
  - uses CPLEX to solve subproblems
- Pereira & Pinto Method
  - uses a sample size of 30 for each problem
- Abridged Nested Decomposition
  - number of Stage t subproblems solved from each Stage t-1 branching value: 15
  - initial number of Stage t branching values: 2
    - number of Stage t branching values increases with each failed convergence test
- Both methods terminate when first stage objective value is within one standard deviation of statistical estimate

# Computational Results

- Test Problems
  - Dynamic Vehicle Allocation (DVA) problems of various sizes
    - set of homogeneous vehicles move full loads between set of sites
    - vehicles can move empty or loaded, remain stationary
    - demand to move load between two sites is stochastic
  - DVA. $x.y.z$ 
    - $x$  number of sites (8, 12, 16)
    - $y$  number of stages (4, 5)
    - $z$  number of distinct realizations per stage (30, 45, 60, 75)
  - largest problem has  $> 30$  million scenarios

# Computational Results (DVA.8)



# Model for Colombia Power

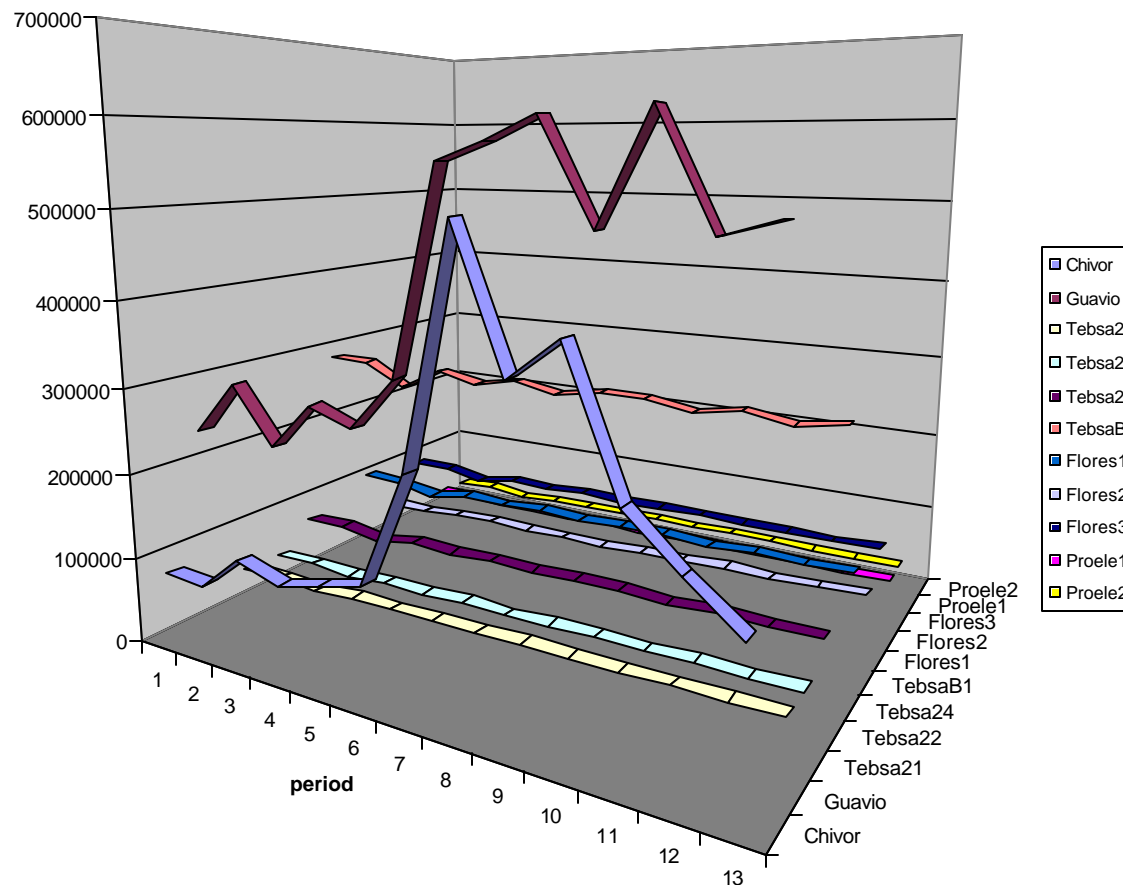
- Basic model in AMPL
- Solvable in CPLEX
- Assume serial correlation within a larger stage (may extend over several months)
- Single branch or multiple branches
- Abridged NDUM able to solve to optimality

# Basic Model

- Objective:
  - minimize total generation costs
- subject to the operating constraints:
  - meet load constraints
  - thermal capacity constraints
  - hydro maximum/minimum flow constraints
  - export/import capacity constraints
  - minimum/maximum reservoir level
  - penalty on alert levels and minimum levels

# Example Results (selected plants)

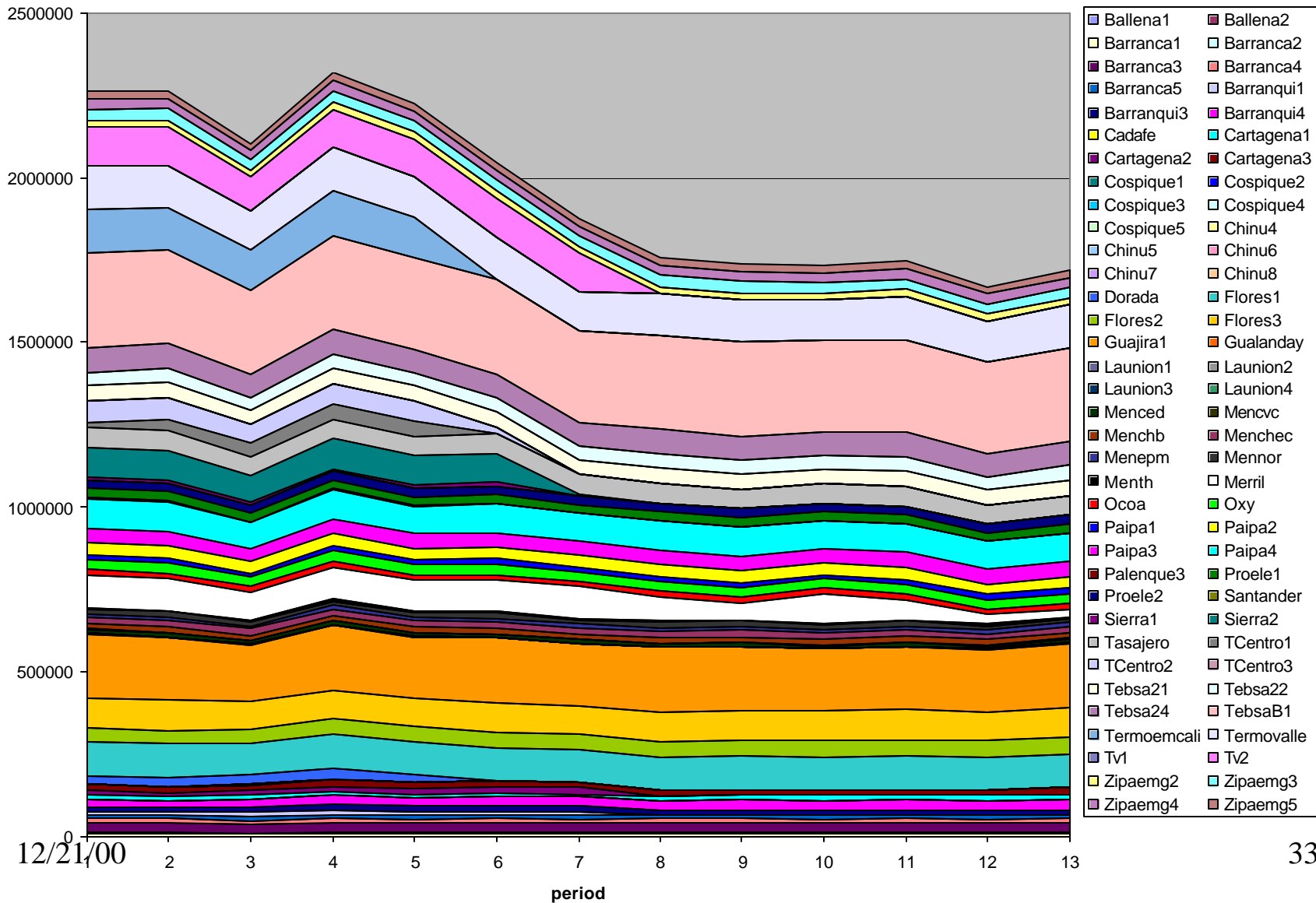
Interest plants





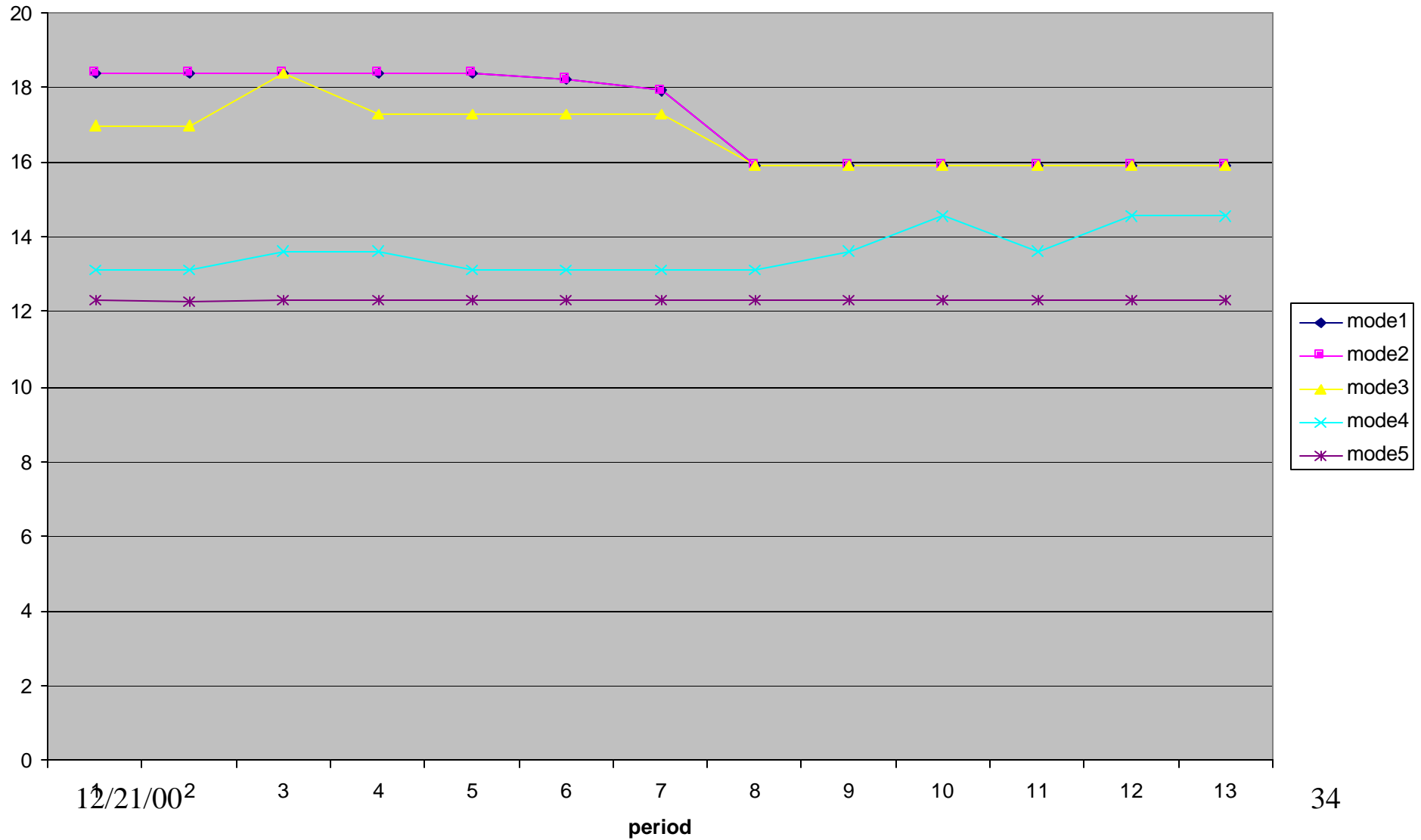
# Example: Thermal Generation

Thermal generation

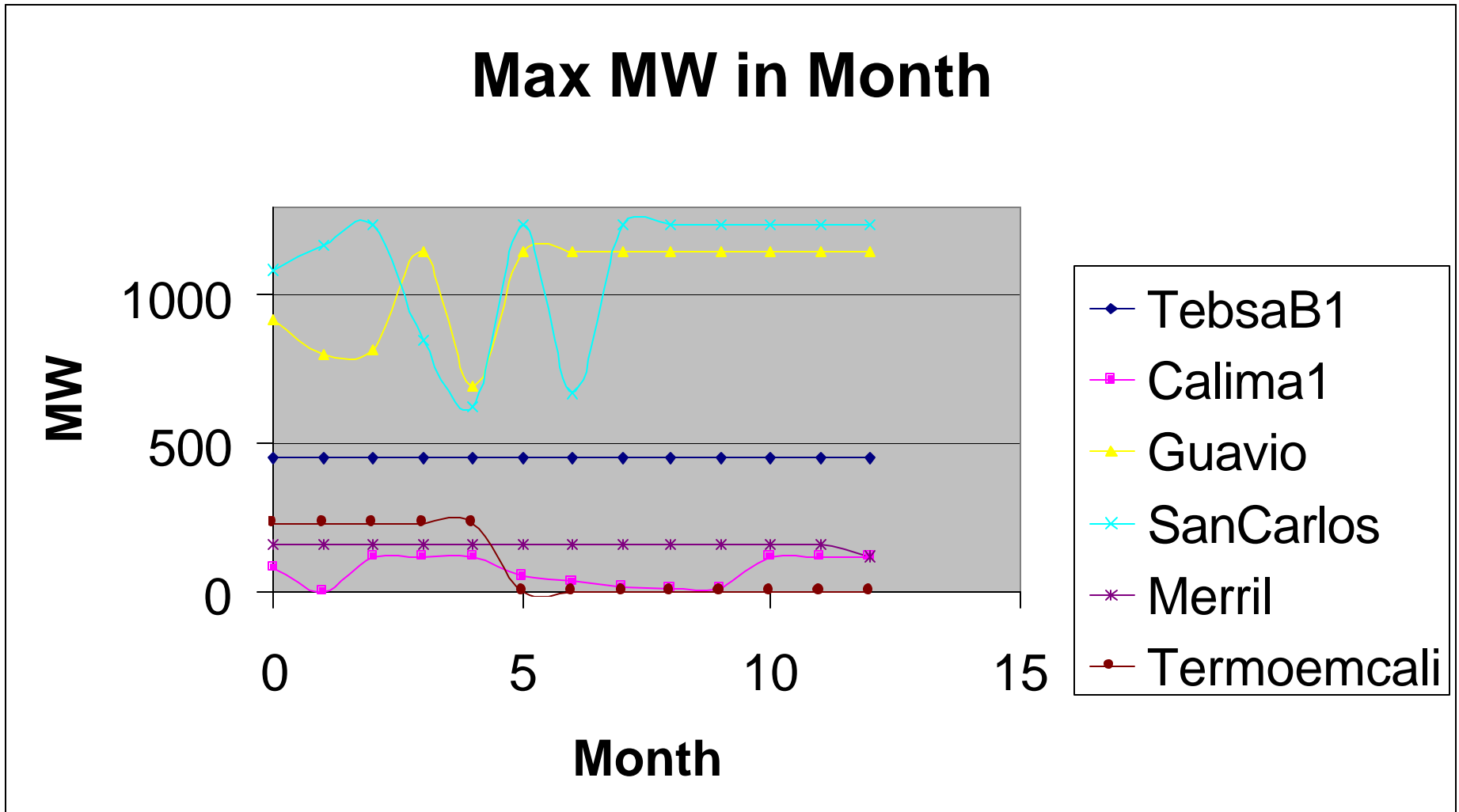


# Example: Dual Prices

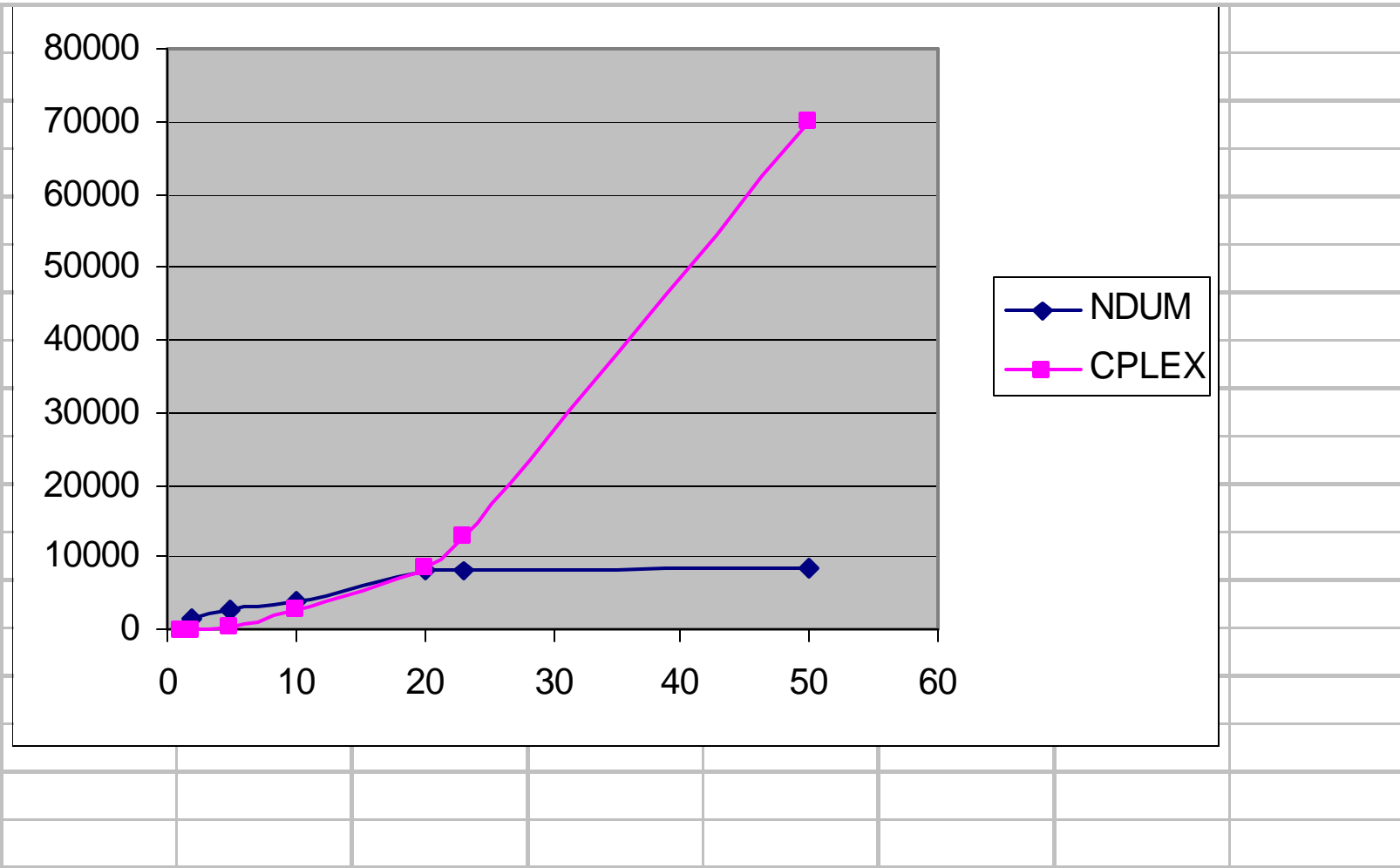
Meet load dual area 1



# Example: Max MW



# NDUM and CPLEX v. No. of Scenarios



# Summary

- Multiple equilibria
- Demand is known
  - $\text{MCP} \leq \lfloor c_N \rfloor$
  - marginal  $i$  at  $\lfloor c_j \rfloor$ , others  $i$  at  $\lfloor c_i \rfloor + \mathbf{e}$
- Demand is stochastic
  - $\text{MCP} \leq \lceil c_N \rceil + \mathbf{e}$
  - marginal  $i$  at  $p_j - \mathbf{e}$ ,  $\lceil c_k \rceil$ , or  $\lceil c_k \rceil + \mathbf{e}$ , others  $i$  at  $\lfloor c_i \rfloor + \mathbf{e}$
- Algorithm to find highest MCP equilibrium point
- Abridge Nested Decomposition for Solving Individual Problems

# Extensions

- Include start-up cost of generation
- Analyze multi-period problems: cost depends on the dispatch of the previous period
- Allow multiple bids per bidder

# Worst Equilibrium point

- Worst equilibrium MCP = highest possible bid price

$$p_i^* = c_i \quad \text{for } i = 1, \dots, N-1$$

$$\sum_{i=1}^{N-1} x_i^* = d - \mathbf{e}$$

$$p_N^* = O\mathbf{e}$$

$$x_N^* \geq \mathbf{e}$$