
Agent-based system planning tools

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Goals

Good designs of complex artifacts cannot be obtained without good measures of performance, and good tools for synthesis, analysis, simulation and especially, verification.

Our goals are:

- 1. a suite of agent-based tools for the simulation and verification of market designs.**
- 2. new performance measures for market designs (traditional measures, such as HHI, are inadequate).**
- 3. insights and guidelines for market designs.**



Why software-agent-based tools?

1. Fidelity.

Actual markets contain autonomous agents: sellers, ISOs, buyers, Autonomous software agents are needed to model their interactions.

2. Packaging.

Autonomous agents are convenient packaging devices. They make it possible to write modular, robust and expandable software.

3. Efficiency.

Experimental work is easier with software agents than human agents



Tools for the conceptual design of quasi-repetitive auctions, such as day-ahead and hour-ahead auctions

Work done at CMU



What do we need to include in our models?

Markets seek balances between the needs of sellers and buyers. These balances emerge as a result of:

- **Competition**
- **Cooperation**
- **Learning**
- **Spatial and temporal distributions (network conditions)**



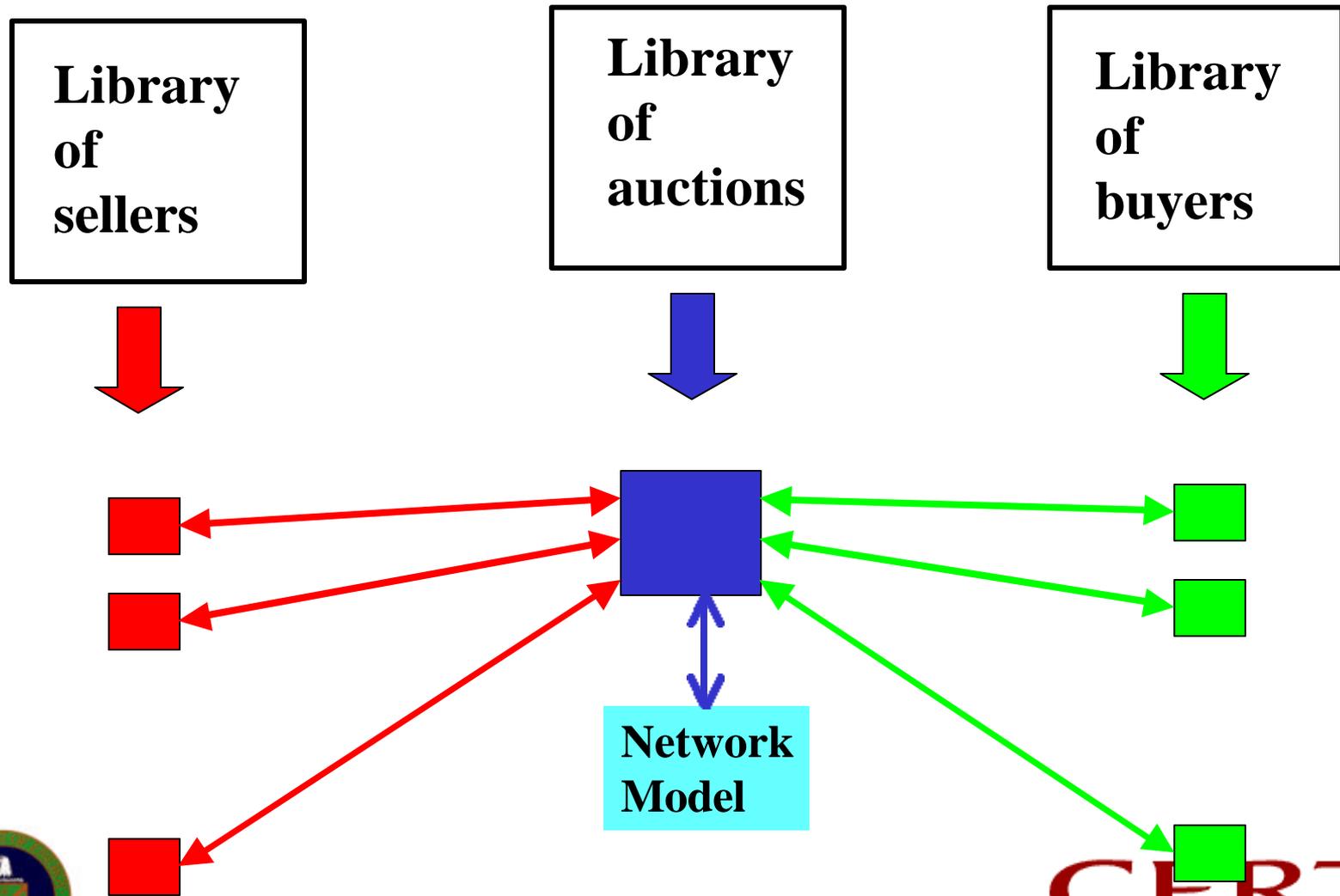
What do we want market designs to achieve?

**In a multi-agent-system that is distributed in space and time,
we want:**

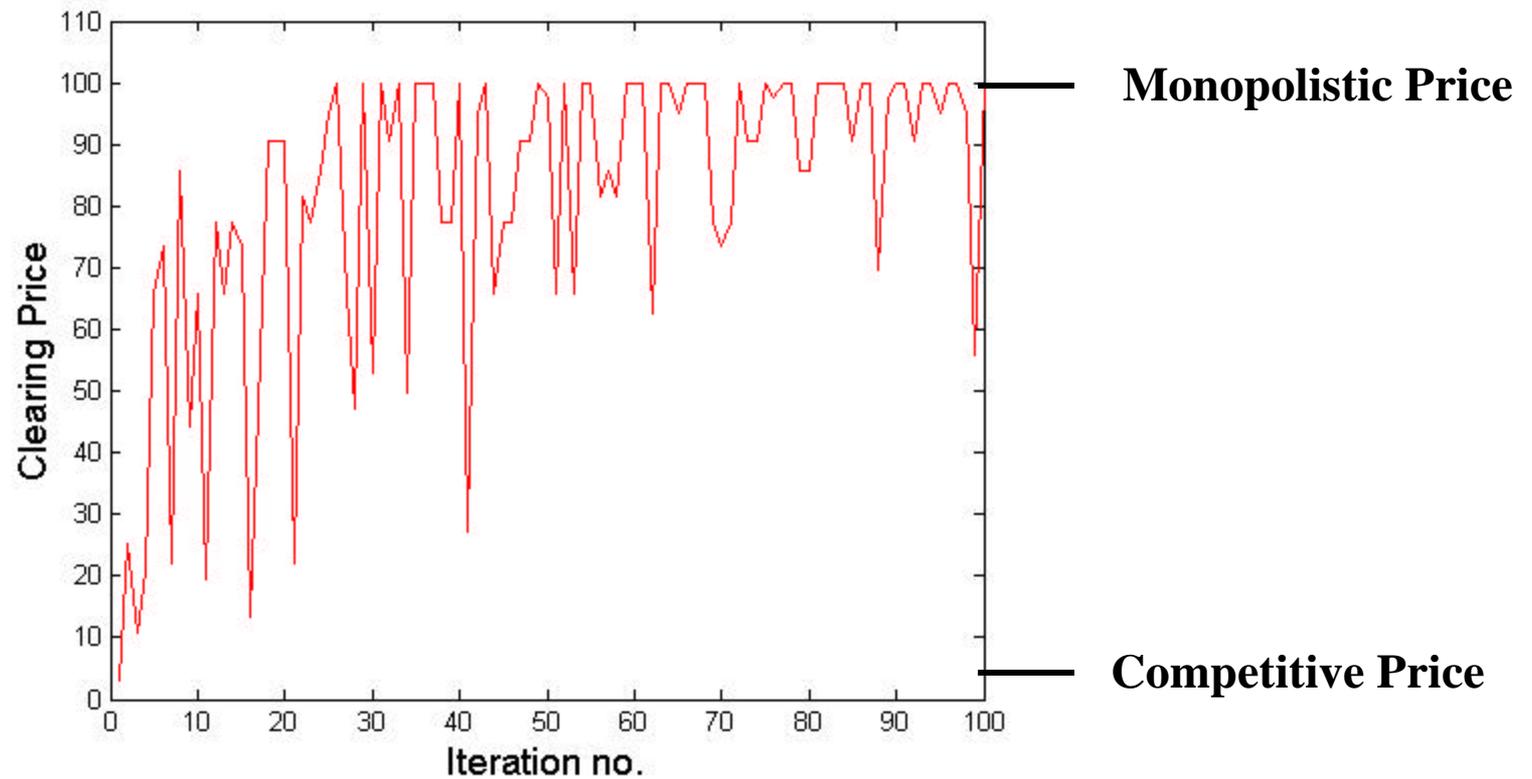
Local optimality → global optimality



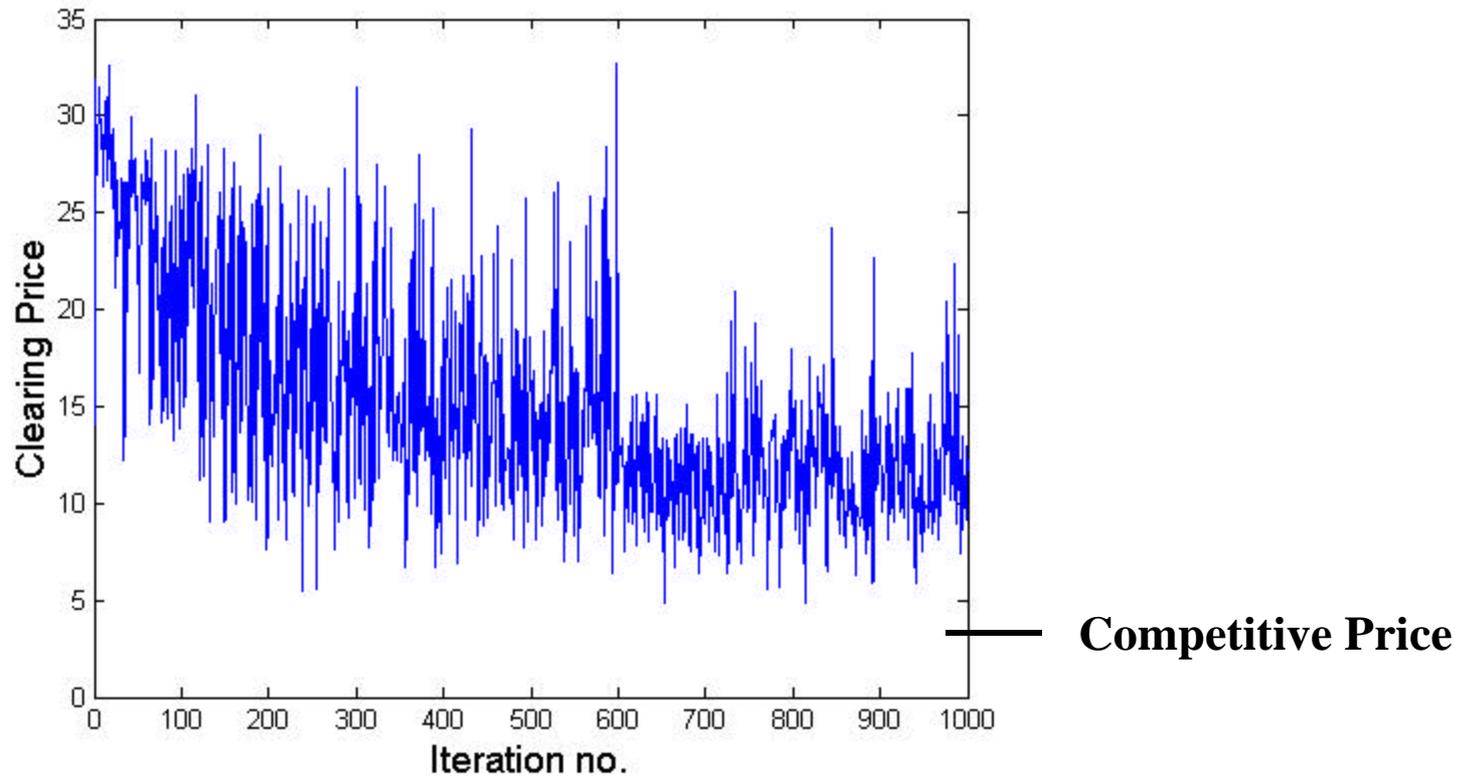
THE SIMULATOR



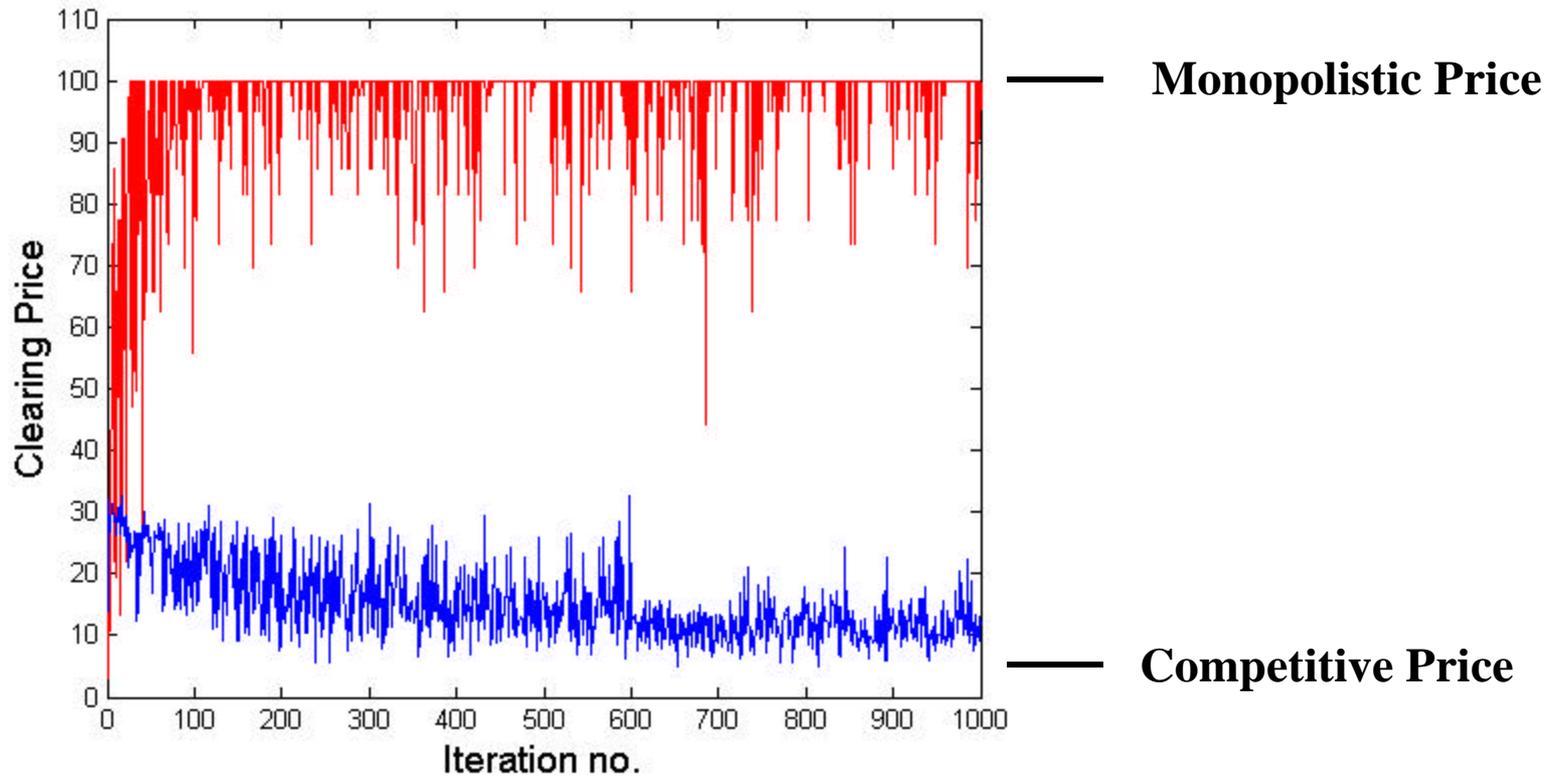
Intelligent sellers and fixed demand



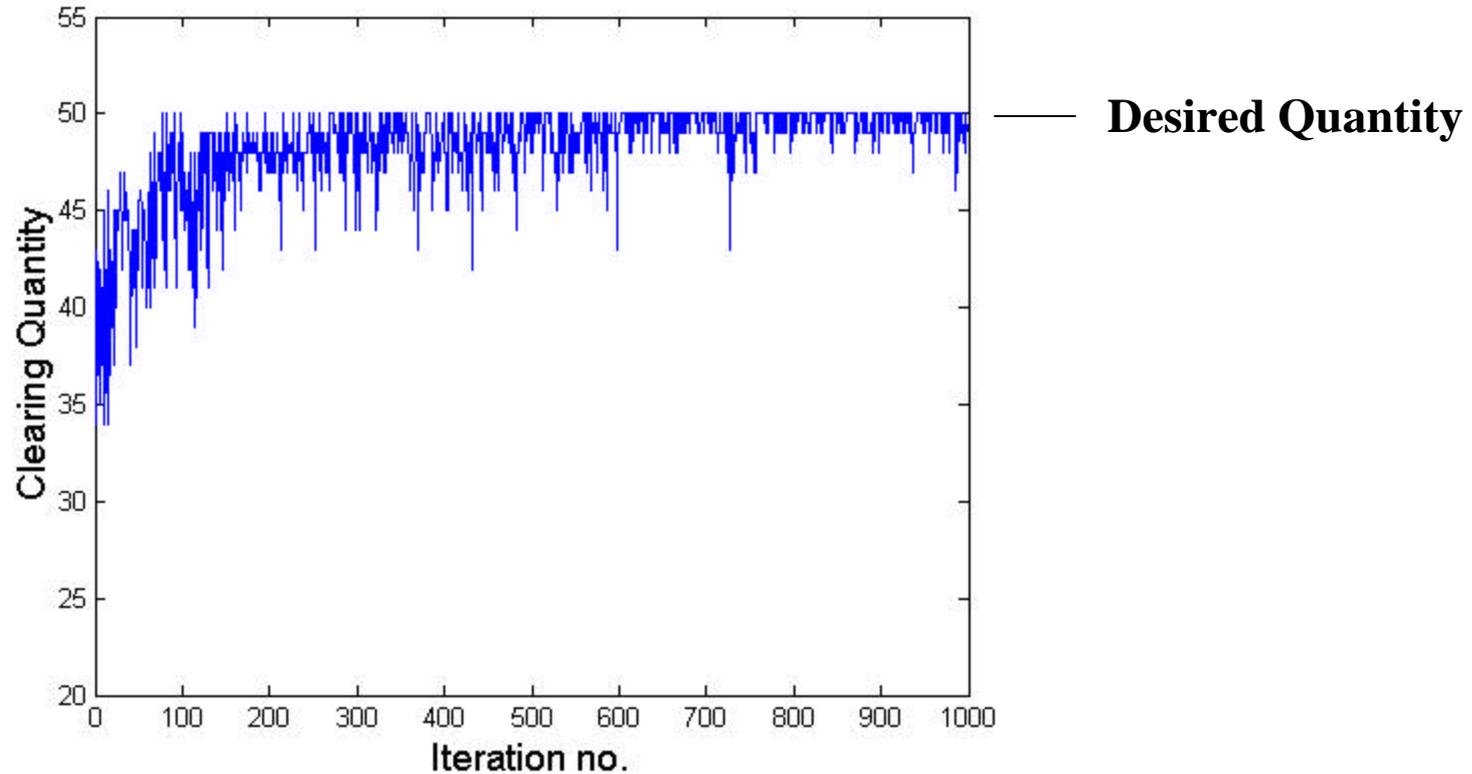
Intelligent sellers and intelligent buyers



- Intelligent sellers and fixed demand
 - Intelligent sellers and intelligent buyers
-



Intelligent sellers and intelligent buyers



Insights

1. Intelligent sellers and passive buyers with fixed or elastic demands
 - Sellers Maximize (profit) → Competitive Price
 - Sellers Maximize (profit + price) → Monopolistic Price
2. Intelligent buyers → Competitive Price

Elastic (price-sensitive) demand is not enough. To drive prices down to competitive levels requires intelligent buyers capable of learning (adjusting their buying strategies).

Learning ↔ collusion among the sellers



Two new performance measures for quasi-repetitive markets

1. P_L/P_M : the ratio of the price-with-agent-learning to the monopolistic price
2. P_L/P_C : the ratio of the price-with-agent-learning to the competitive price

The performance of quasi-repetitive markets is critically affected by agent-learning. Measures that do not take learning into account cannot be good indicators of actual performance.



N-ABLE: a simulator for the detailed evaluation of real-time pricing effects on Transco load, pricing and profitability

Work done at Sandia



Research

- Investigate potential effects of residential real-time price contracts on power system loads, prices, and stability
 - Formulate model of residential power usage and its re-scheduling under RTP contracts
 - Formulate model of Transcos that sell both uniform and RTP contracts
 - Run agent-based simulations based on different contract and consumption scenarios



Model of Residential Power Consumption

- **Desired** hourly household power consumption is the sum of
 - **Immovable use** - e.g., heating, A/C
 - **Movable use** - e.g., washers, dryers
 - **Optional use** - e.g., lighting

$$\hat{\mathbf{E}}_i^h = \begin{bmatrix} \hat{e}_i^{h,1} & \hat{e}_i^{h,2} & \dots & \hat{e}_i^{h,24} \end{bmatrix}$$

$$\hat{\mathbf{E}}_m^h = \begin{bmatrix} \hat{e}_m^{h,1} & \hat{e}_m^{h,2} & \dots & \hat{e}_m^{h,24} \end{bmatrix}$$

$$\hat{\mathbf{E}}_o^h = \begin{bmatrix} \hat{e}_o^{h,1} & \hat{e}_o^{h,2} & \dots & \hat{e}_o^{h,24} \end{bmatrix}$$

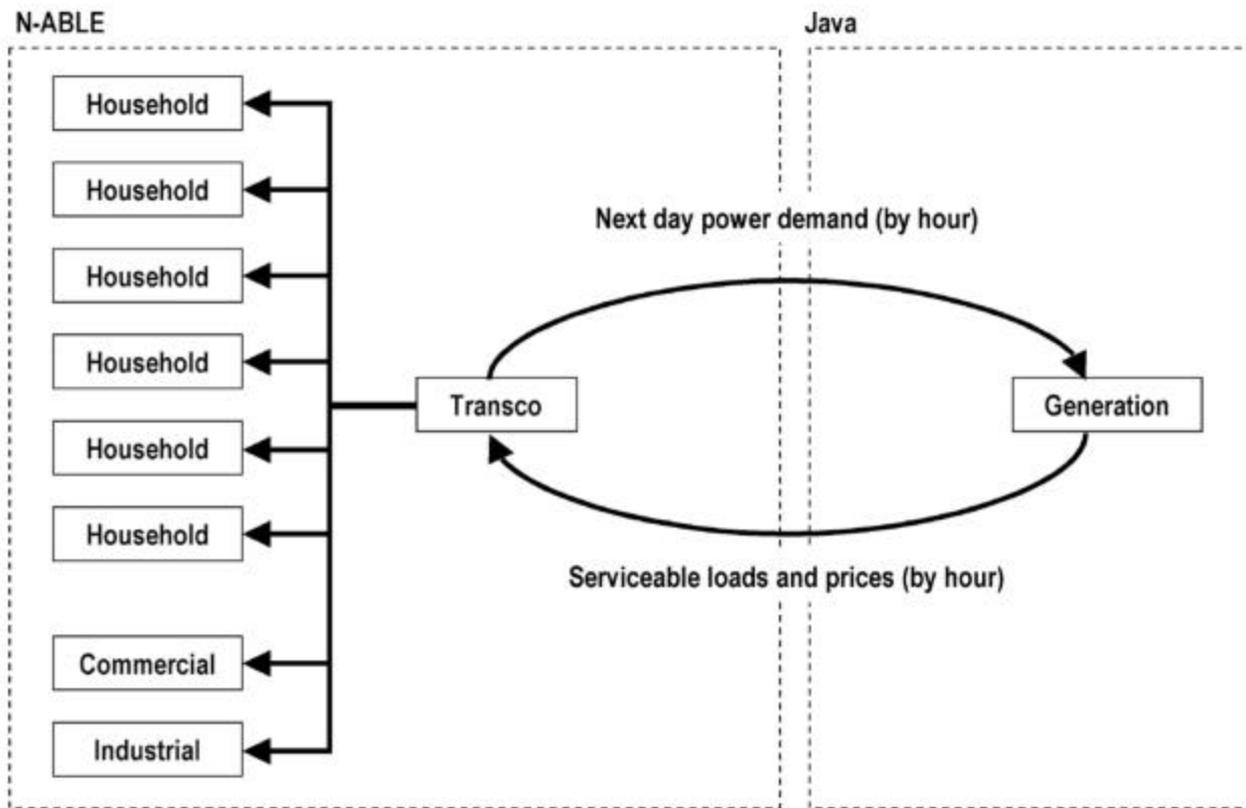


Rescheduling Power Consumption

- Under uniform contracts, households use desired levels.
- Under RTP contracts, some households will reschedule moveable and optional usage, based on
 - **Hourly RTP price distribution,**
 - **Personal monthly/daily budget for power expenses, and**
 - **Inconvenience caused by rescheduling their usage.**
 - **The fraction of households that are very frugal will also discard optional use.**



Simulation Model



Baseline Industry Data

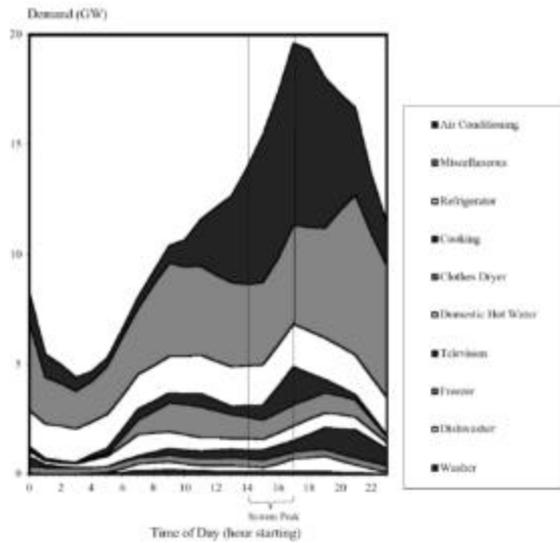
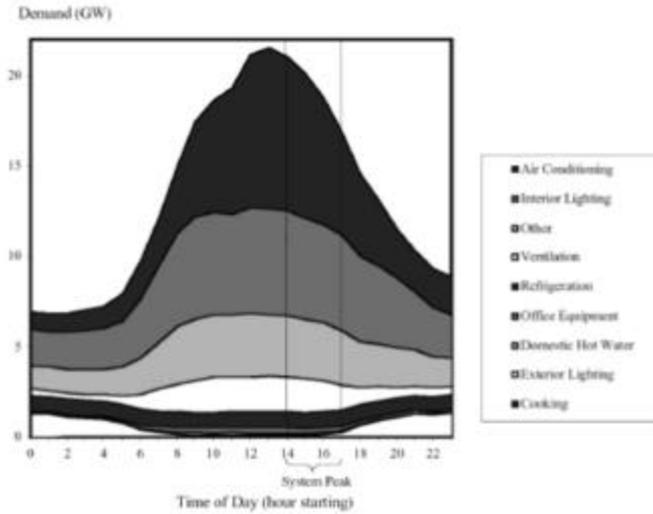


Figure 3: Household Hourly Use Profile (from Bro



se Profile (from Brown and Koomey[6])

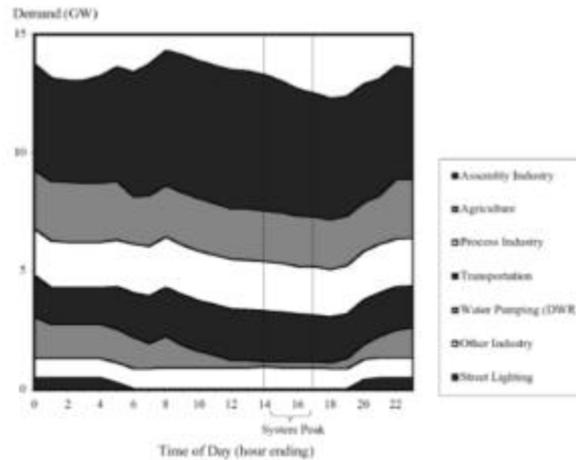


Figure 5: Industrial Hourly Use Profile (from Brown and Koomey[6])



Results: system usage under uniform-only contracts

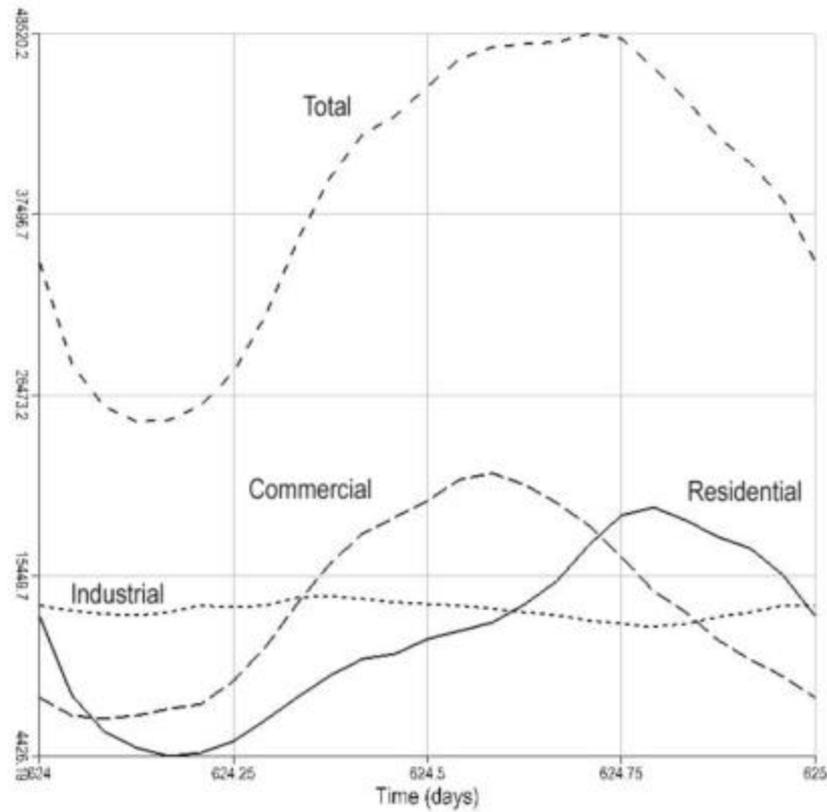


Figure 7: Hourly Usage: Residential, Commercial, and Industrial Usages



Results: RTP Contracts displace loads from peak-price hours

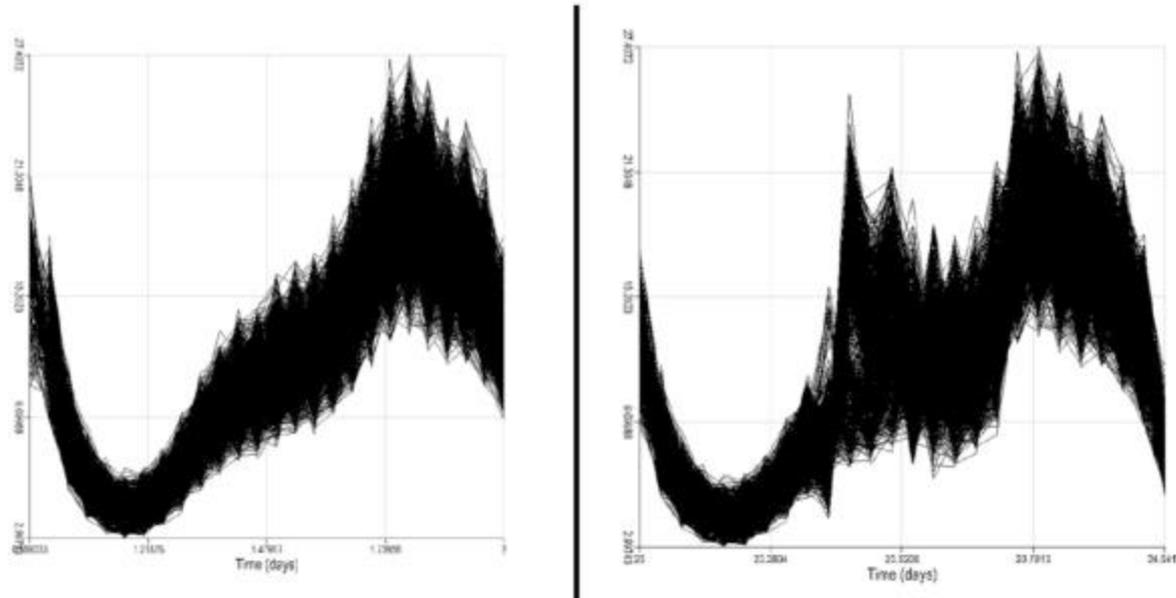
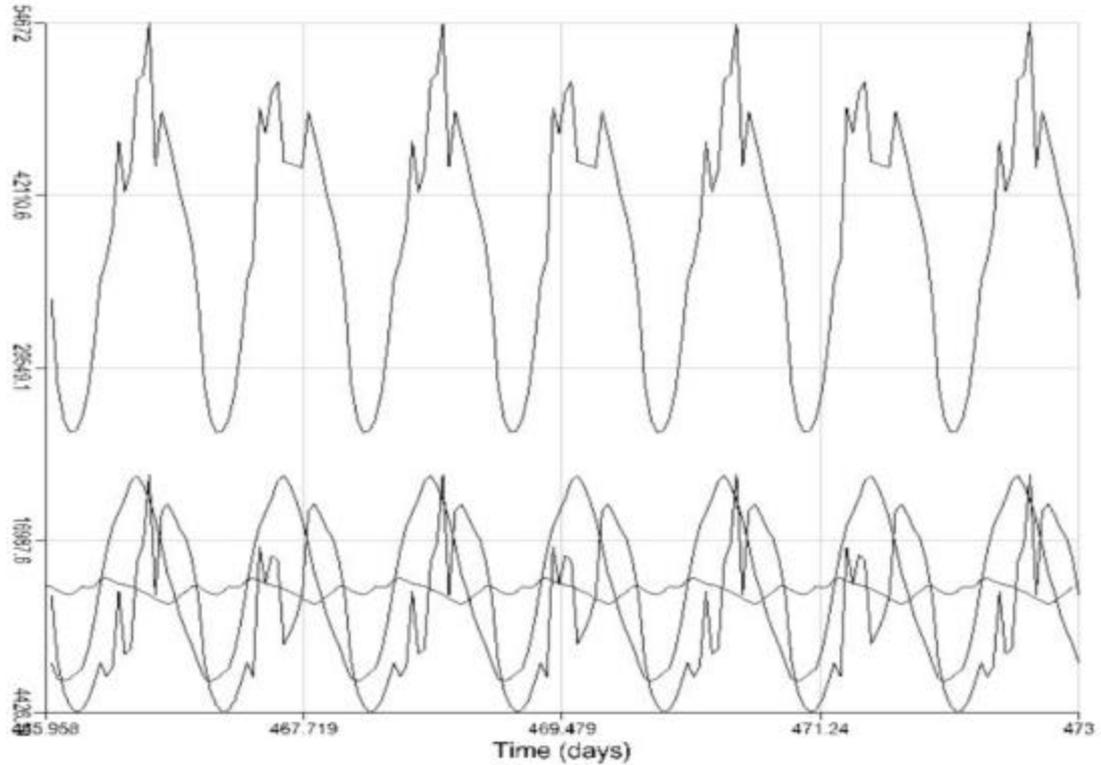


Figure 8: Hourly Usage: Uniform Price-Only (Left) and RTP-Only Contracts (Right)



Results: system loads under RTP-only contracts



Results: RTP market share as a function of households' willingness to experiment with contracts, budgets, rescheduling inconvenience

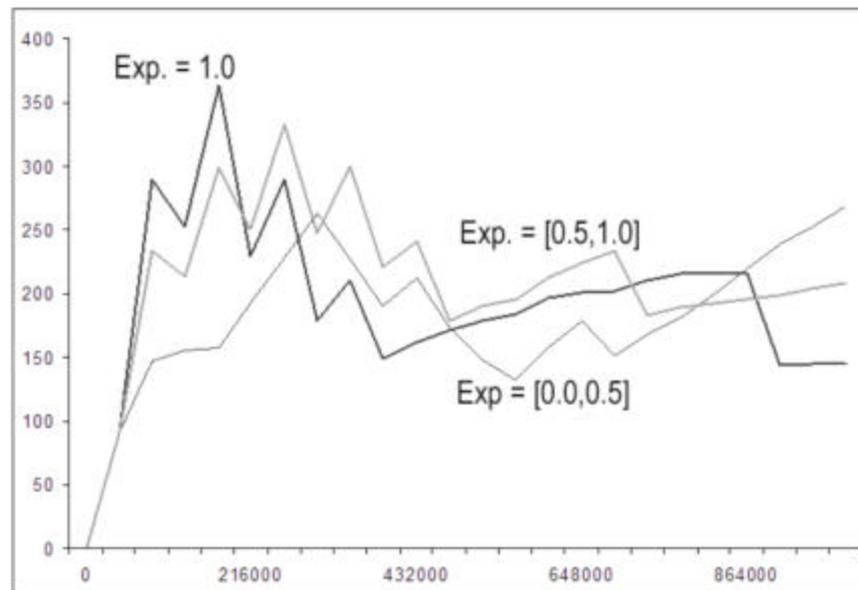


Figure 10: Number of RTP Contracts Over Time: Various Experimenting



Summary

- **RTP contracts increase the potential excess transmission capacity a Transco can sell in short-term (lucrative) markets.**
- **Transcos that aggressively market RTP contracts can have that capacity before other Transcos have it.**
- **Drastic defection of households from RTP contracts to uniform contracts creates needs for transmission capacity that will likely have to be purchased in short-term (expensive) markets.**
- **Model has sufficient fidelity to do future analysis of other Transco and residential power use policy issues.**

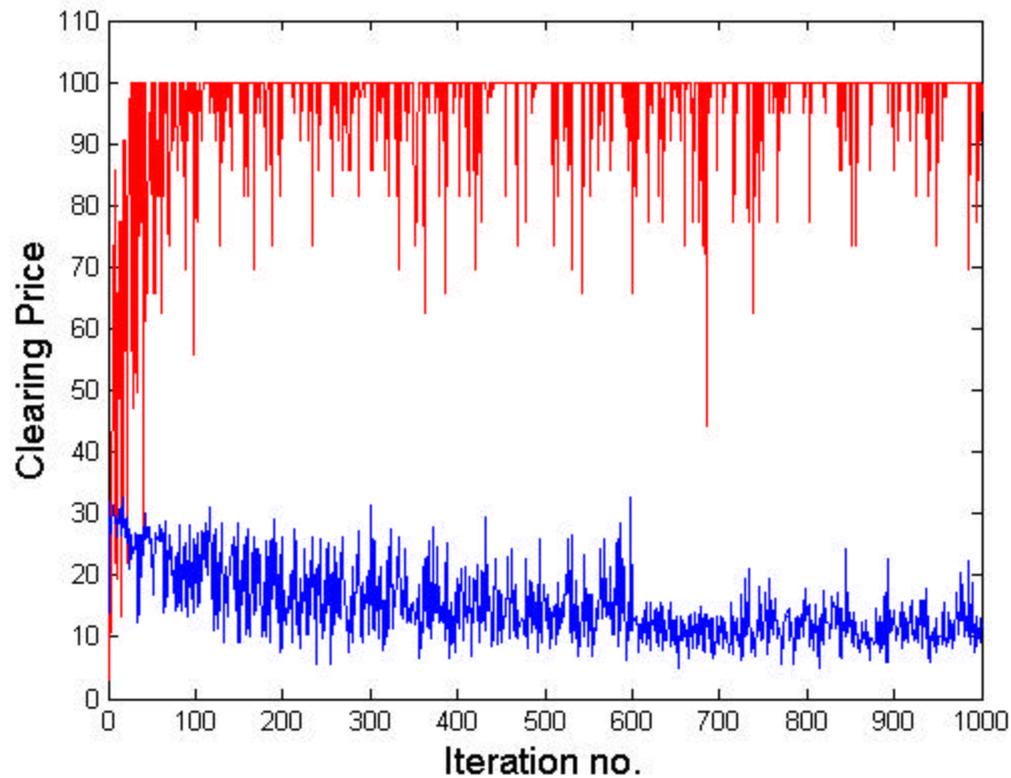


Measures of market consistency: fractal dimensions of price data

Work done at Cornell

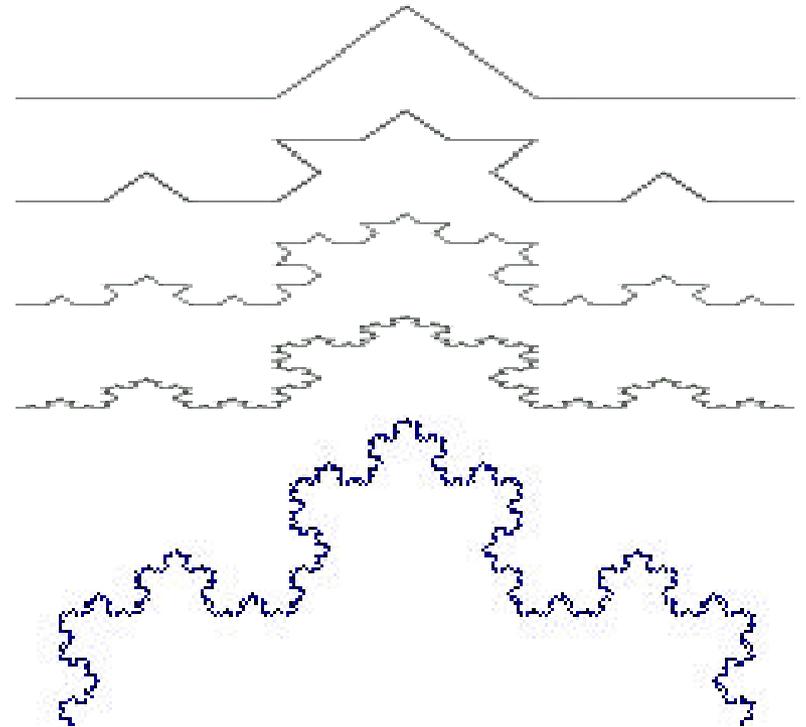


What information is contained in price fluctuations?



Koch Curve

Fractal dimension = 1.2619



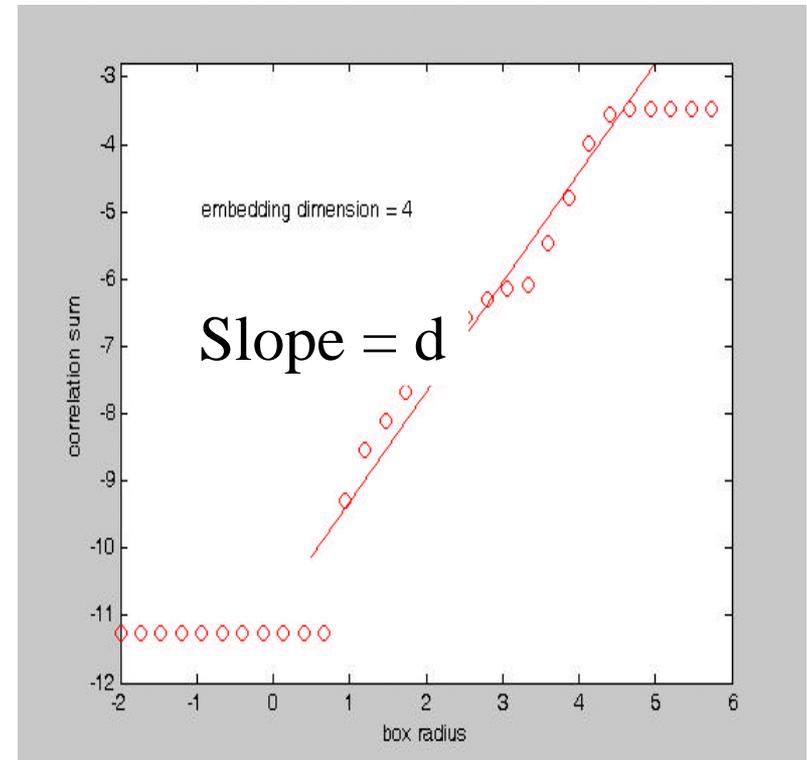
A conjecture

Changes in a fractal dimension of price data can be useful in assessing market performance



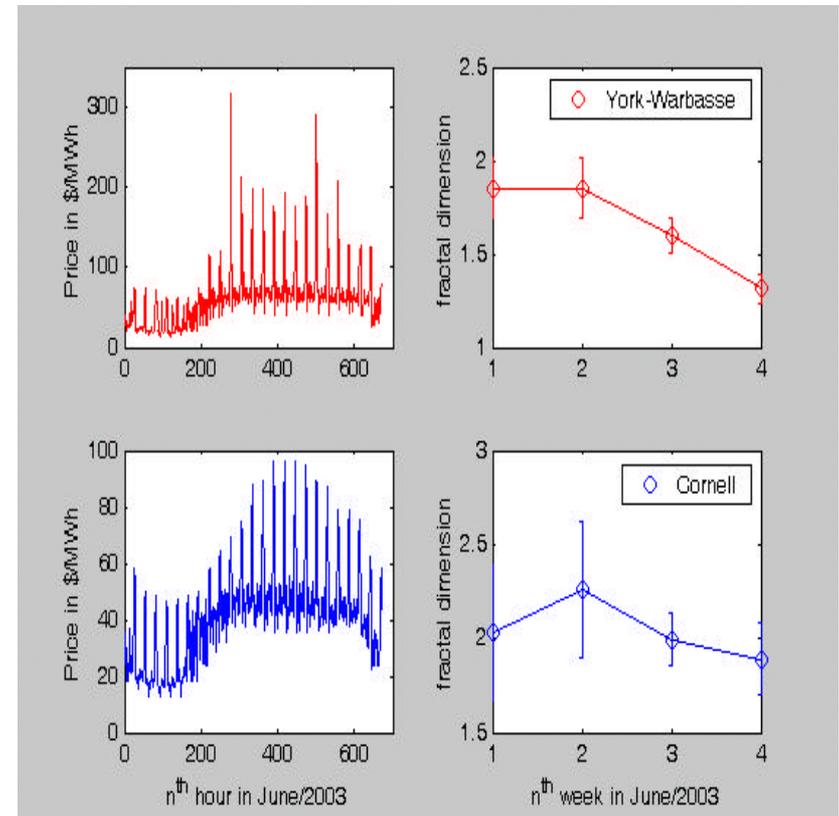
A fractal dimension

- Correlation sum, $C(\epsilon)$;
overall frequency of visit to ϵ -neighborhood
- For a very small ϵ ,
 $C(\epsilon)$ contains one point
- For a very large ϵ ,
 $C(\epsilon)$ engulfs the whole state
- Only from intermediate ϵ ,
 $d = \text{slope in log-log plot} \rightarrow$



NYISO DATA IN JUNE 2003

- High demand period starts
- York-Warbase (locational benefit)
 - change in dimension
- Cornell (no locational benefit)
 - no change in dimension
- Something happened at York-Warbase



CONCLUSIONS

- Price is a good measure of the state of a market
- Some fractal dimensions reflect the state of the market
- Changes in the values of these dimensions indicate changes in seller strategies, and could prove useful to the ISO.



Major accomplishments

1. Tools for the identification of flaws at the conceptual stage of designing quasi-repetitive markets
2. Tools for the detailed simulation of real-time-pricing
3. New measures of market performance
 P_L/P_M , P_L/P_C , fractal dimension
4. Design insights



Future Work

Tools for seams-issues

Tools for bi-lateral trades

Tools for designing customer strategies

Continuing work on measures and insights

Refinement of current tools



Linkages

PSERC

CEIC

ABB

NYISO

