

Agent-Based Modeling for Electric Power Markets

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Modeling & Measuring Societal Resilience**

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Presentation Outline

- * The complexity of electric power systems
- * Can agent-based modeling (ABM) help?
- * Adventures in ABM testbed development for U.S. restructured electric power markets:

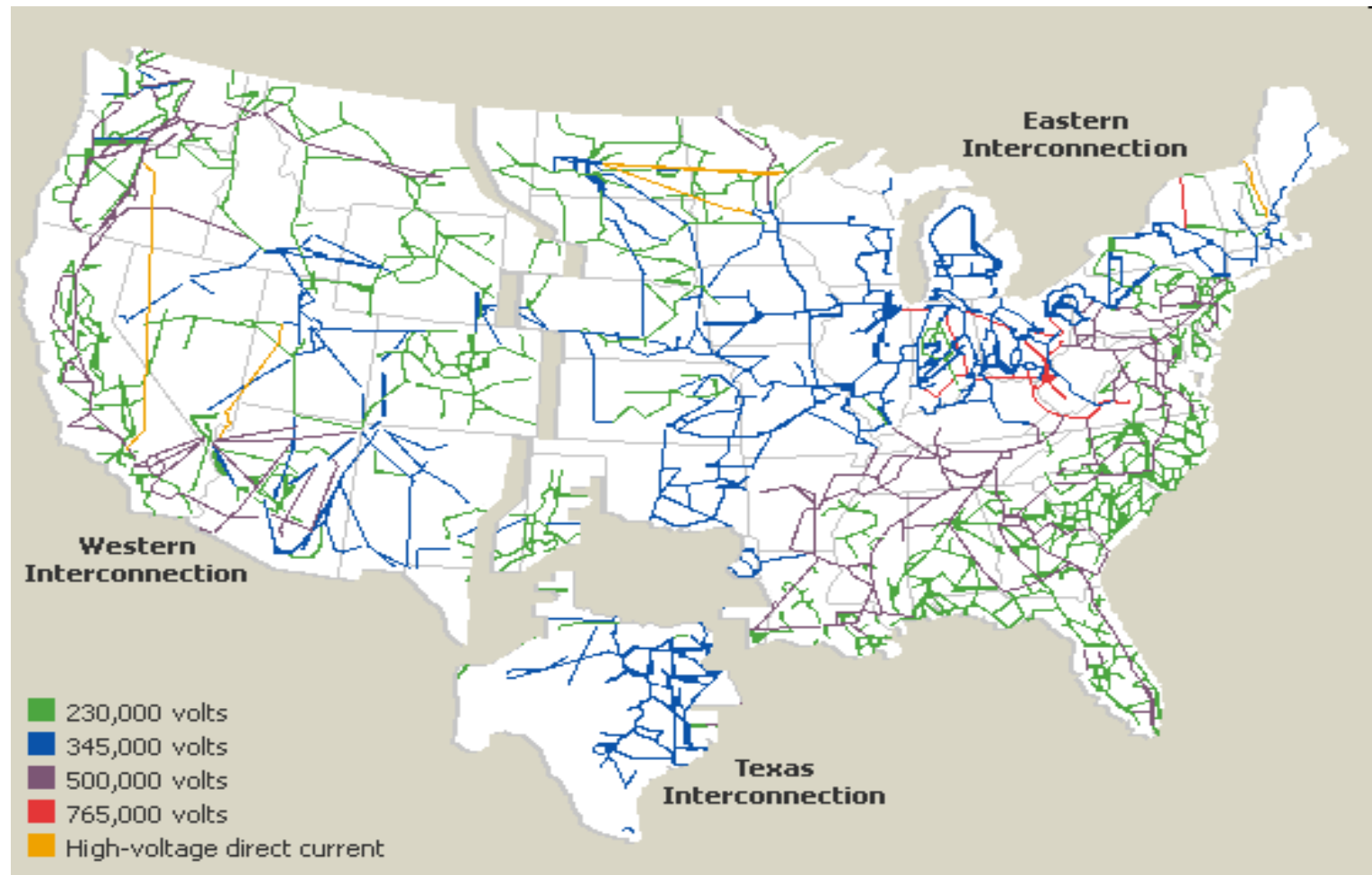
AMES = Agent-based Modeling of Electricity Systems

- * Illustrative experimental findings

The Complexity of Power Systems

- ◆ Many industrialized economies are in the process of restructuring the way power is produced & distributed
- ◆ These restructured power systems are immensely complicated, encompassing
 - Physical constraints
 - Institutional arrangements
 - Behavioral dispositions of human participants
- ◆ To be useful and informative, power system studies need to take proper account of all three elements.

U.S. Wholesale Electric Power Transmission Grid

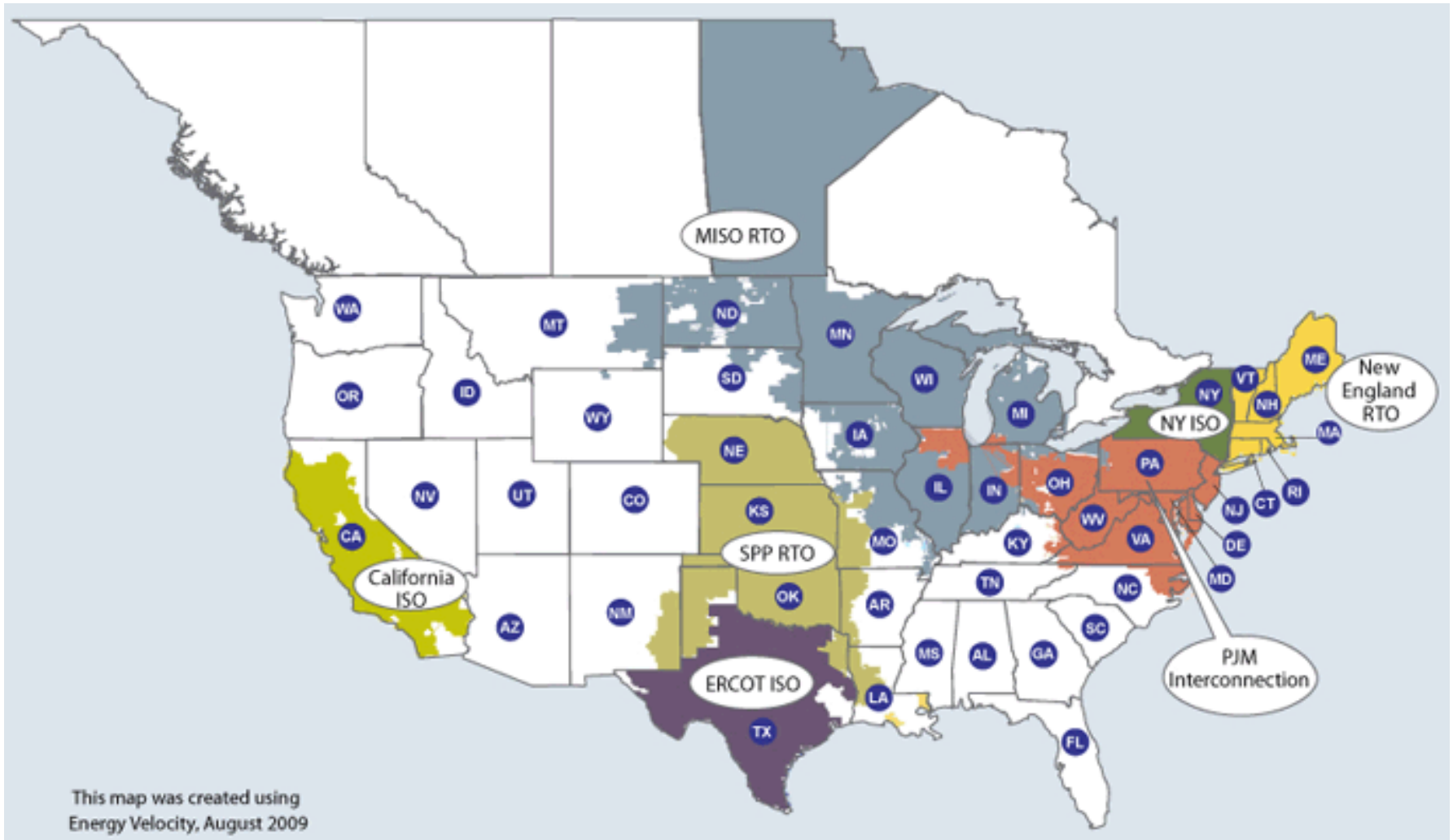


North American Restructuring of Wholesale Power Markets

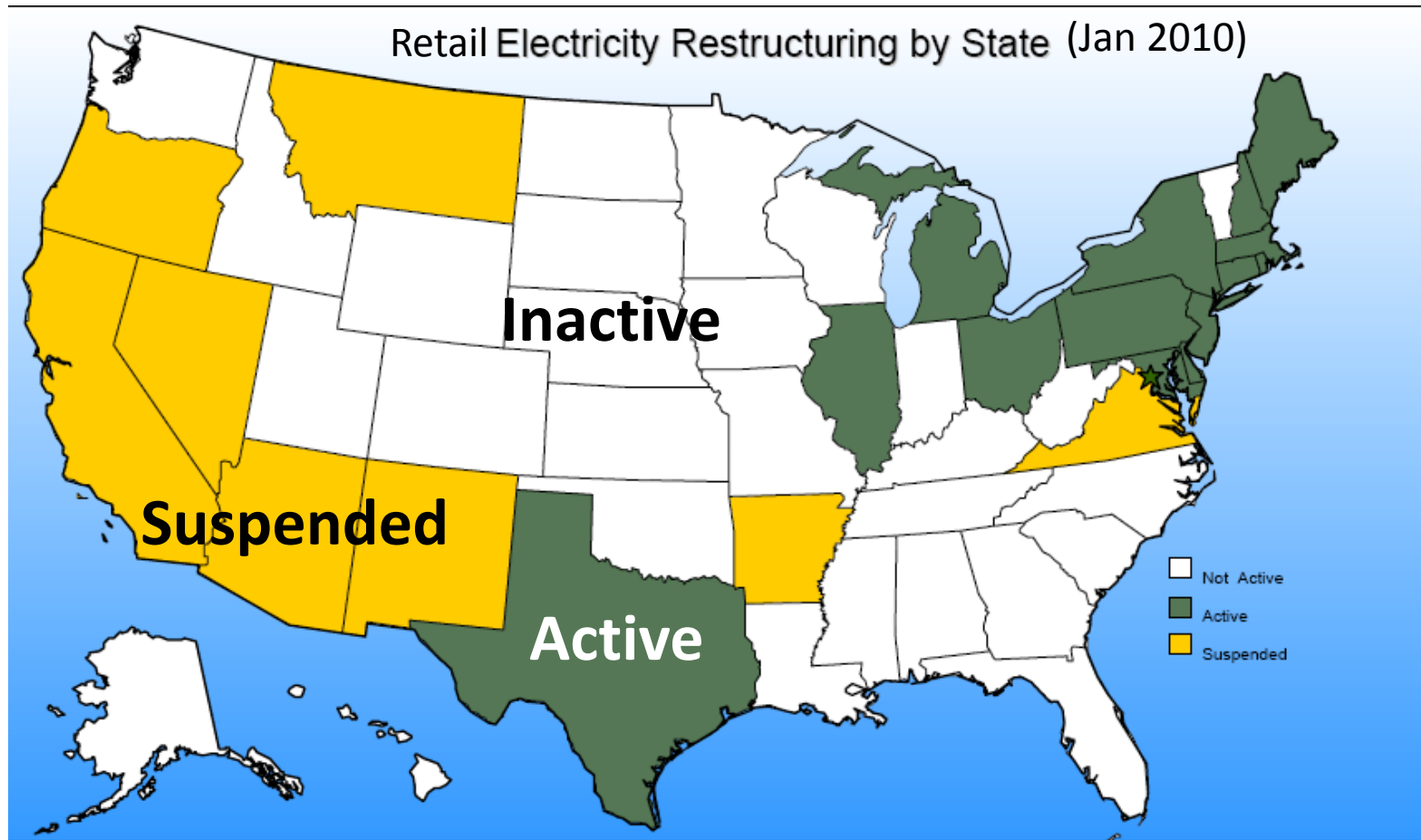
- ❑ In April 2003 the U.S. Federal Energy Regulatory Commission (FERC) proposed adoption of a wholesale power market design with particular core features.
- ❑ Over 50% of North American generation now operates under some variant of the FERC design.
- ❑ **FERC Design Adopters to Date:**
New York (NY-ISO), mid-Atlantic states (PJM), New England (ISO-NE), Midwest/Manitoba (MISO), Texas (ERCOT), Southwest (SPP), and California (CAISO)

FERC Wholesale Power Market Design Adopters to Date

<https://www.ferc.gov/industries/electric/indus-act/rto/rto-map.asp>



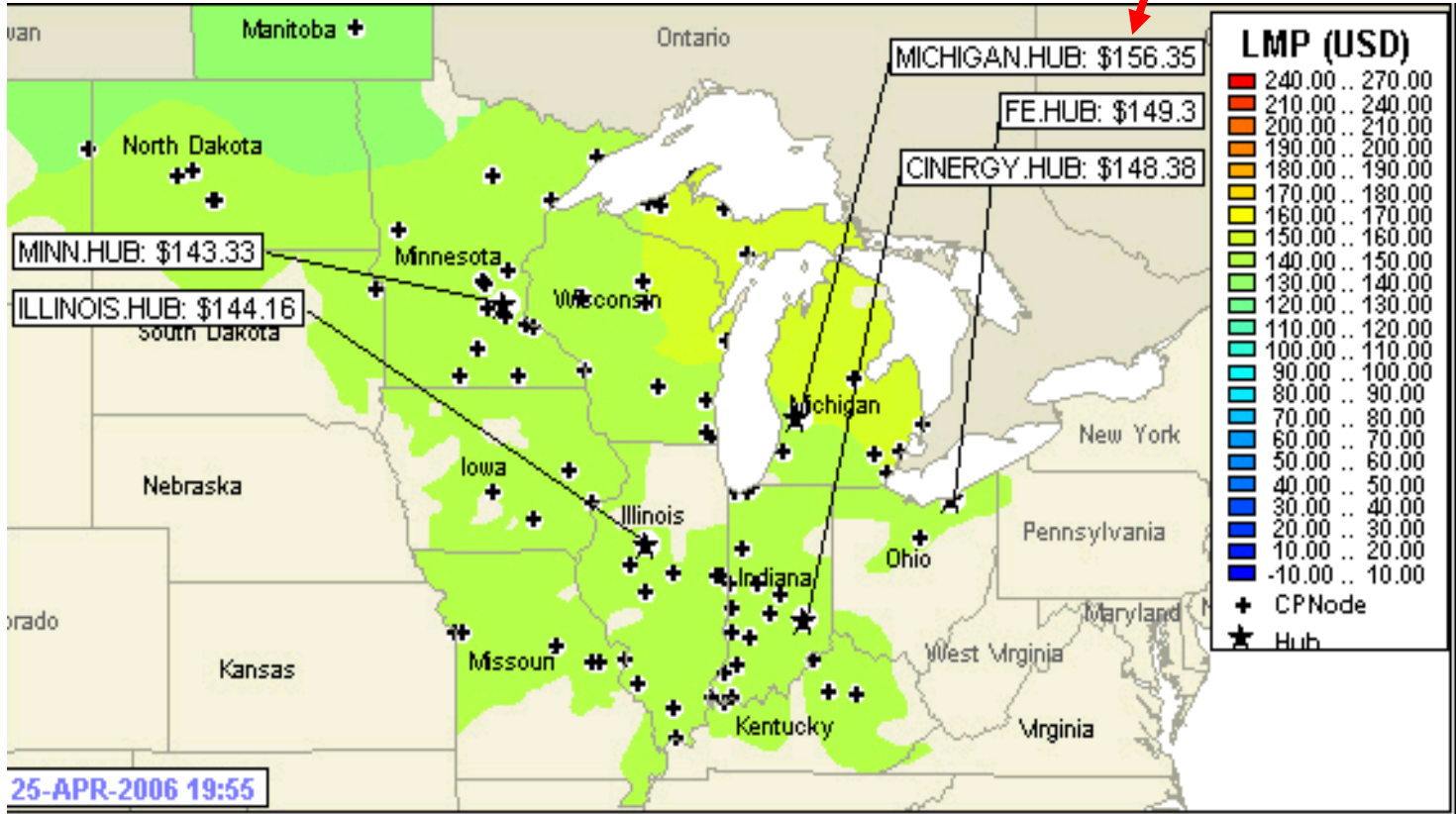
Since 2000 Cal/Enron scandal, retail restructuring has slowed/stopped



Actual Electricity Prices in Midwest ISO (MISO)

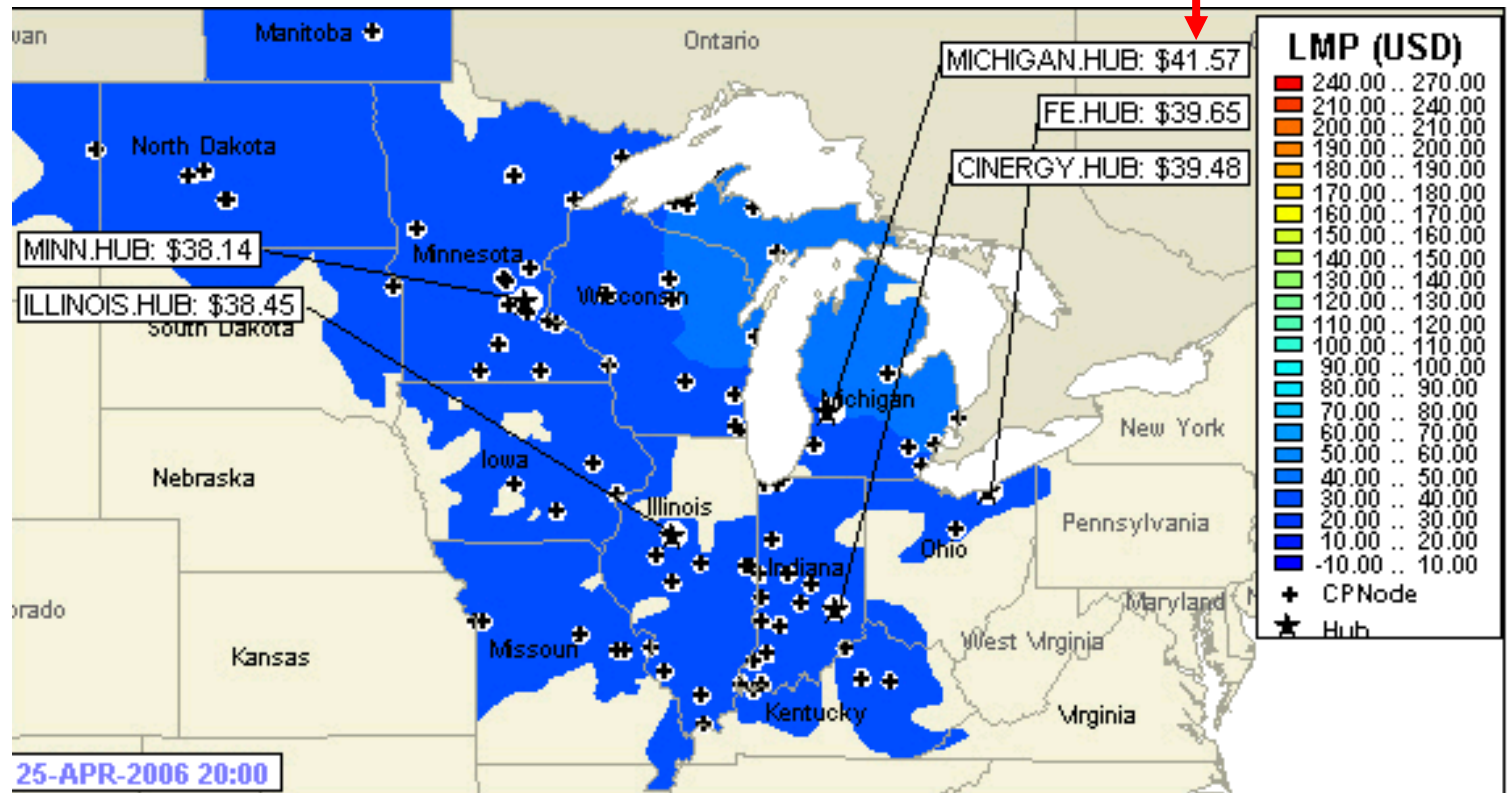
April 25, 2006, at 19:55

Note this price, \$156.35



Five Minutes Later...

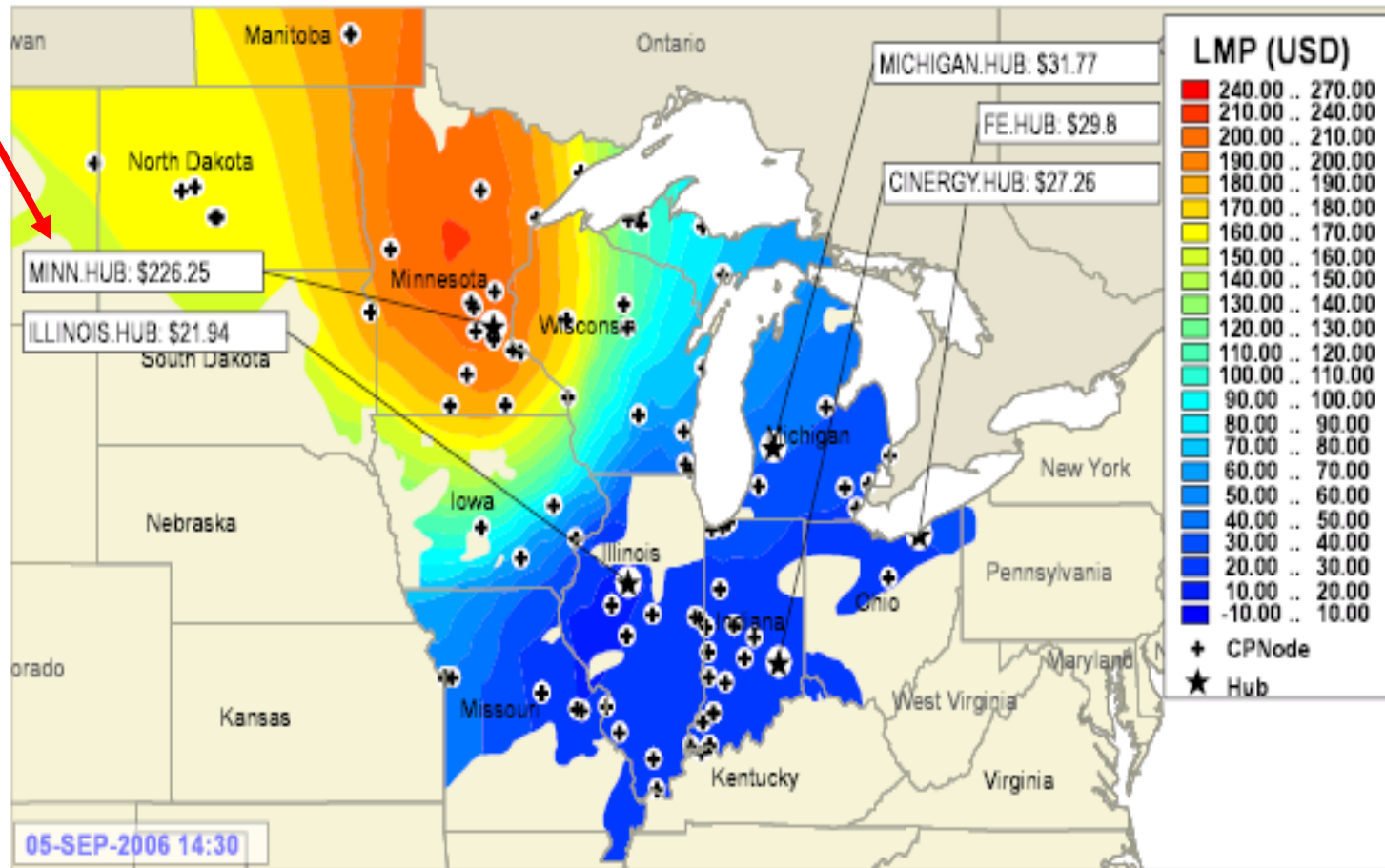
73% drop in price in 5 minutes!



Actual Electricity Prices in Midwest ISO (MISO)

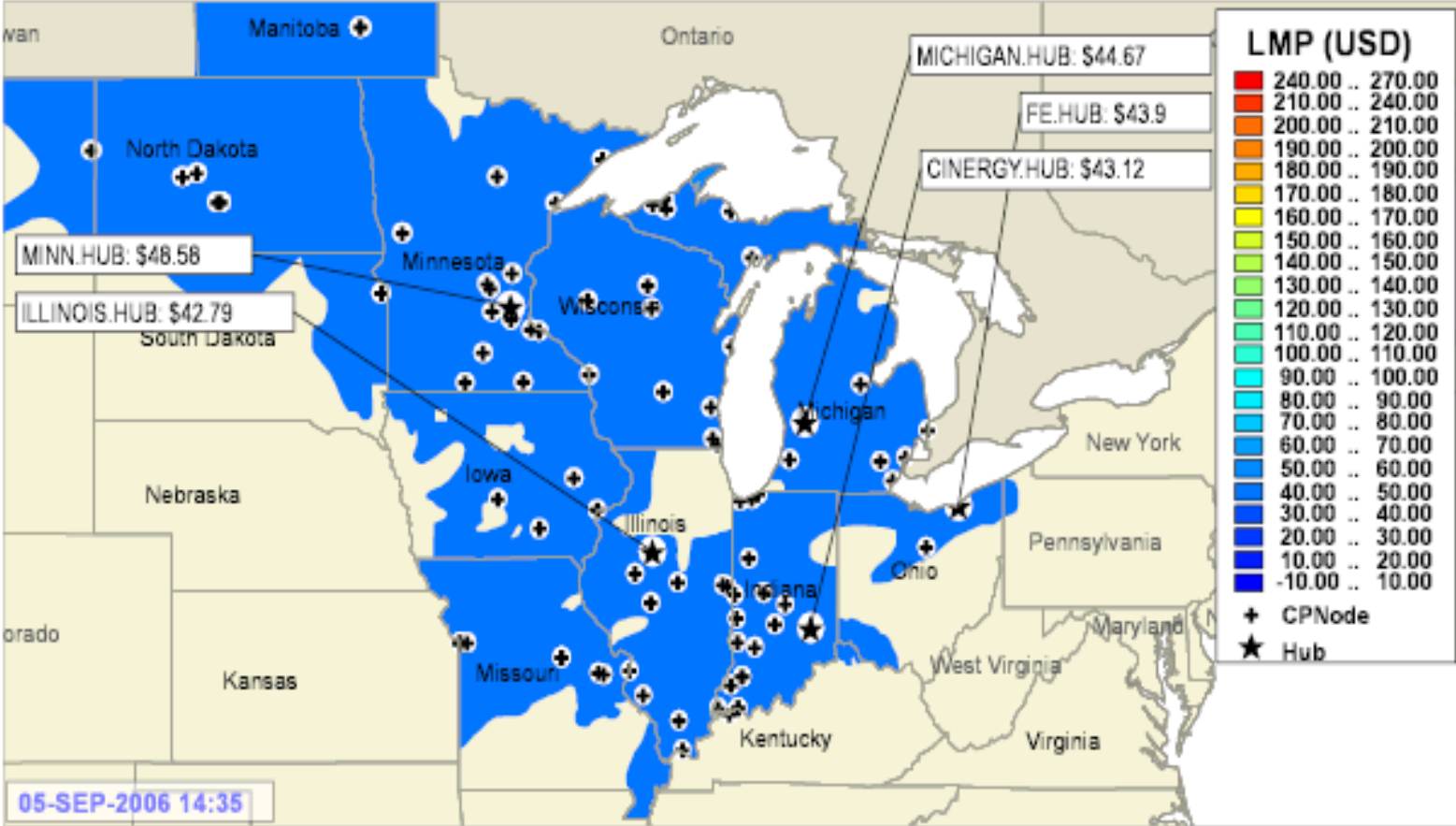
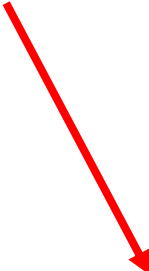
September 5, 2006, 14:30

Note this price, \$226.25



Five Minutes Later...

79% drop in price in 5 minutes!



Can Agent-Based Modeling (ABM) Help?

ABM = Computational modeling of systems as collections of autonomous interacting “agents”

Agent = Encapsulated bundle of data and methods that represents physical, bio., institutional or social entity

- ABM is designed to handle complex systems.
- ABM tools permits researchers to construct testbeds in the form of **computational virtual worlds**
- Starting from user-specified initial conditions, world events are **driven entirely by agent interactions.**

Meaning of “Agent” in ABM ... Continued

Decision-making agents are capable (in various degrees) of

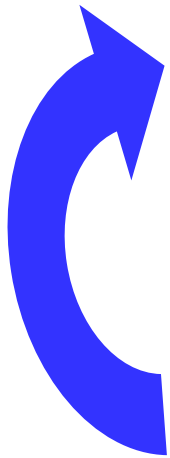
- Behavioral adaptation
- Goal-directed learning
- Social communication (talking with each other!)
- Endogenous formation of interaction networks
- **Autonomy:**
Self-activation and self-determination based on encapsulated (hidden) internal data and methods

Importance of Agent Encapsulation

- ◆ In the real world, all calculations must be done by entities actually residing in the world.
- ◆ Encapsulation forces ABM modelers to respect this constraint:
 - Each procedure must be encapsulated in method of some agent
 - Any procedure encapsulated in a method of a particular agent can only be implemented using the resources of that agent
- ◆ **Encapsulation** → more realistic modeling of real-world systems composed of interacting distributed entities with limited information and computational capabilities.

ABM via Iterative Participatory Modeling

- ◆ Stakeholders and researchers from multiple disciplines engage together in a **repeated looping through four stages of analysis:**



- 1) Field work and data collection
- 2) Role-playing games/human-subject experiments
- 3) Incorporate findings into agent-based test bed
- 4) Generate hypotheses through intensive computational experiments.

ABM and Institutional Design

Key Issues:

- ◆ Will a proposed or actual design promote **efficient, robust (resilient), & fair social outcomes over time?**
- ◆ Will the design give rise to **unintended consequences?**

ABM Culture-Dish Approach:

- ◆ **Develop a computational world** embodying the design, physical constraints, strategic participants, ...
- ◆ Set initial world conditions (agent states).
- ◆ **Let the world evolve** with no further intervention, and observe and evaluate the resulting outcomes.

Example ABM Project: Integrated Wholesale/Retail Power System Operation with Smart-Grid Functionality

Project Directors: Leigh Tesfatsion (Prof. of Econ, Courtesy Prof of Math & ECpE, ISU)
Dionysios Aliprantis (Ass't Prof. of ECpE, ISU)
David Chassin (Staff Scientist, PNNL/DOE)

Research Assoc's: Dr. Junjie Sun (Fin. Econ, OCC, U.S. Treasury, Wash, D.C.)
Dr. Hongyan Li (Consulting Eng., ABB Inc., Raleigh, NC)

Research Assistants:

Huan Zhao (Econ PhD Candidate, ISU)

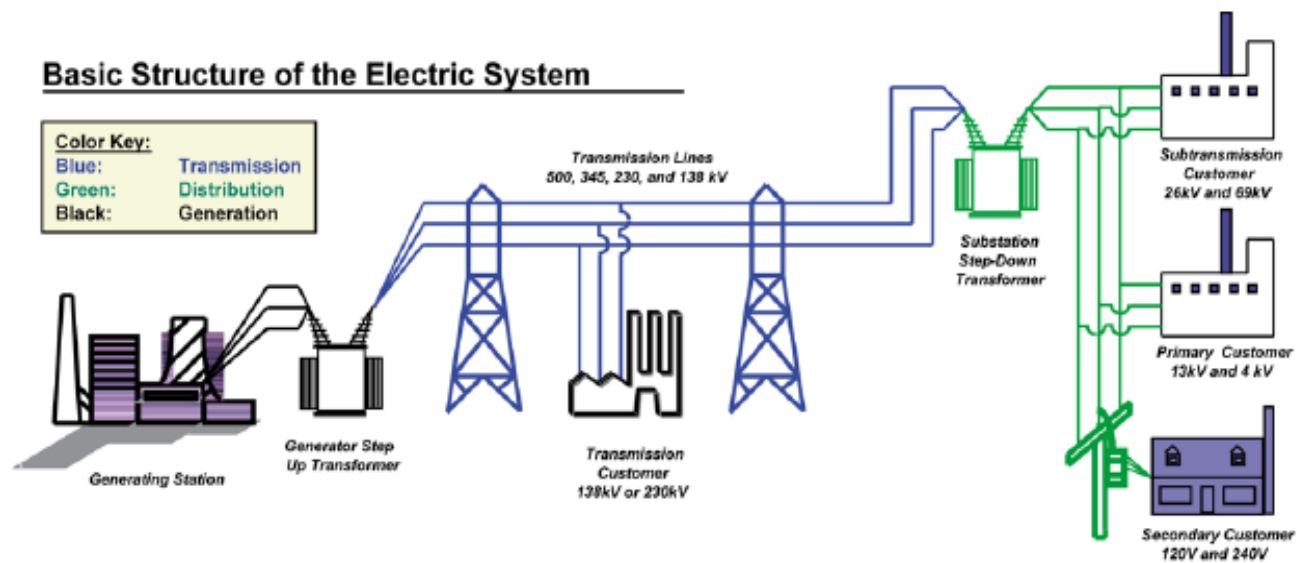
Chengrui Cai (ECpE PhD Candidate, ISU)

Pedram Jahangiri (ECpE grad student, ISU)

Auswin Thomas (ECpE grad student, ISU)

Retail & Wholesale Power System Operations

Source: <https://www.nerc.com/page.php?cid=1> | 15



Generation Transmission

Distribution

Wholesale

Retail

Meaning of “Smart Grid Functionality”?

◆ For our project purposes:

Smart-grid functionality =

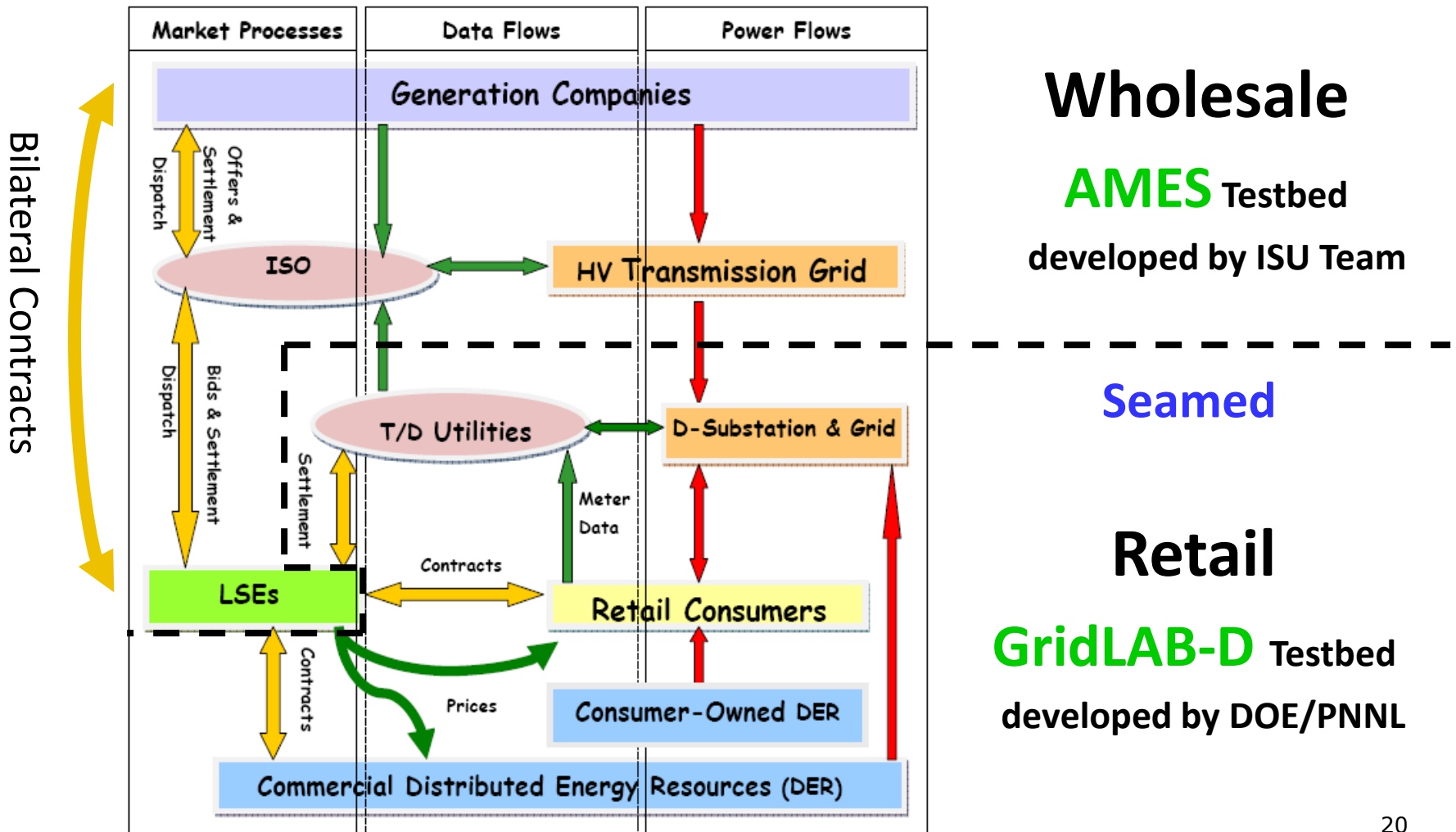
Service-oriented grid enhancements permitting more responsiveness to needs and preferences of retail customers.

Examples: Introduction of advanced metering and other technologies to support

- flexible retail contracting between “load-serving entities” (retail suppliers) and retail consumers
- embedding and use of distributed energy resources

Integrated Retail/Wholesale Testbed Platform

Based on Texas (ERCOT) Retail/Wholesale Structure



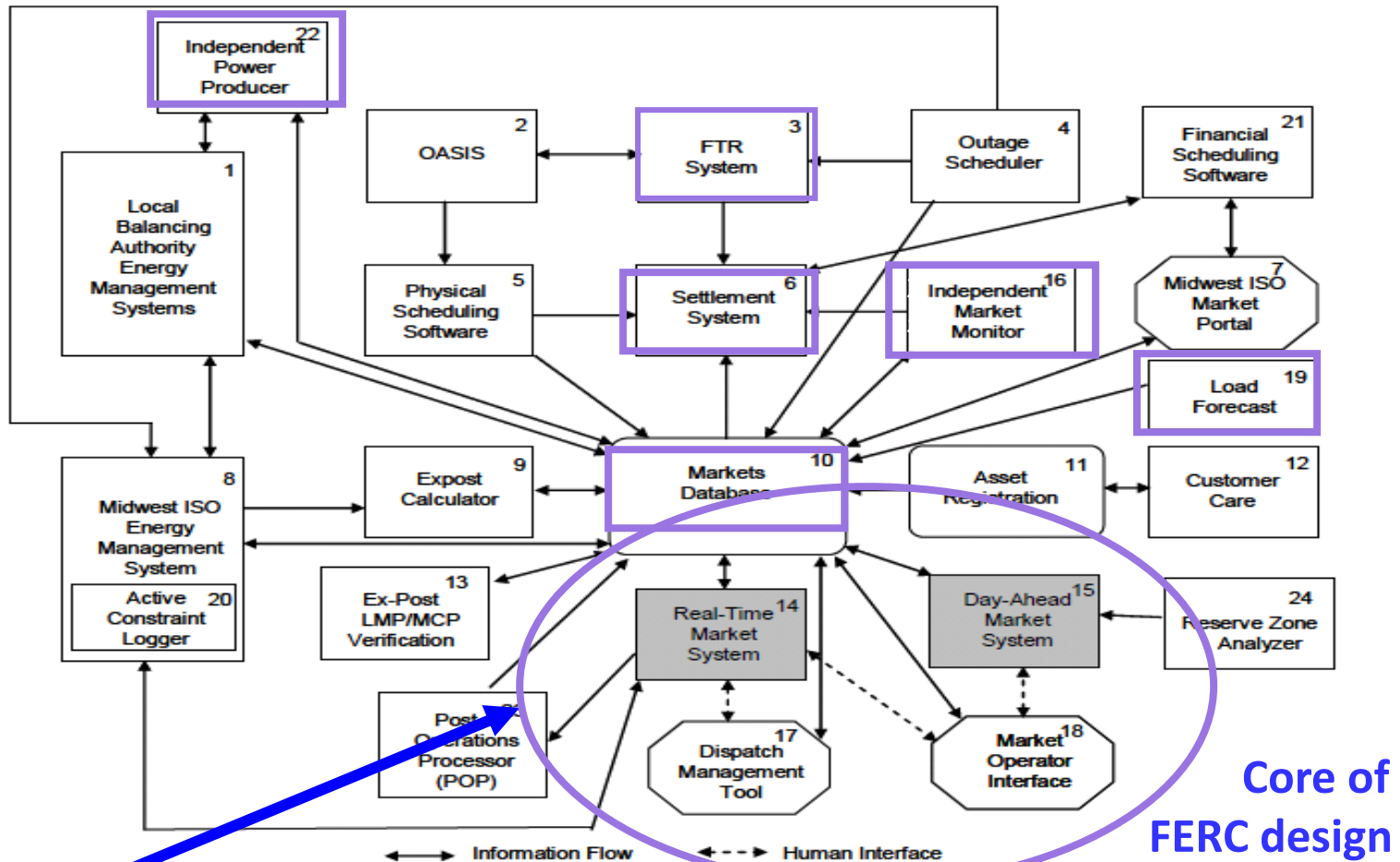
Core of FERC's 2003 Wholesale Power Market Design

- Wholesale power market to be managed by *independent system operator (ISO)* having no ownership stake
 - *Two-settlement system*: Concurrent operation of day-ahead (forward) & real-time (spot) markets
 - Transmission grid congestion managed via *Locational Marginal Prices (LMPs)*, where LMP at bus k = least cost of servicing 1 additional MW of power at bus k
 - Oversight & market power mitigation by outside agency
- ➔ **Has led in practice to complicated systems difficult to analyze by means of standard analytical & statistical tools !**

Example: Complex MISO Market Organization

Business Practices Manual 001-r1 (1/6/09)

Exhibit 2-3: DART Components Overview



Two-Settlement Power Market System under LMP

AMES focus to date

Project Work to Date: Wholesale Level

- Development and open-source release of **AMES = Agent-based Modeling of Electricity Systems**
 - AMES = ABM testbed with core FERC design features
 - Used to test performance under FERC design
 - Used to test performance under modifications of design
- ★ **AMES Homepage (code/manual/publications):**
<http://www.econ.iastate.edu/tesfatsi/AMESMarketHome.htm>

AMES (V2.05) Architecture

(based on business practices manuals for MISO/ISO-NE)

➤ Traders

- GenCos (sellers)
- LSEs (buyers)
- **Learning** capabilities

➤ Independent System Operator (ISO)

- System reliability assessments
- Day-ahead scheduling via **bid/offer based optimal power flow (OPF)**
- Real-time dispatch

➤ Two-settlement system

- Day-ahead market (double auction, financial contracts)
- Real-time market (settlement of differences)

➤ AC transmission grid

- **Generation Companies (GenCos) & Load-Serving Entities (LSEs)** located at user-specified transmission buses
- Grid congestion managed via **Locational Marginal Prices (LMPs)**
- **LMP at bus k** = Least cost of servicing one additional MW of power at bus k.

AMES Modular & Extensible Architecture (Java)

◆ Market protocols & AC transmission grid structure

- **Graphical user interface (GUI) & modularized class structure** permit easy experimentation with alternative parameter settings and alternative institutional/grid constraints

◆ Learning representations for traders

- **Java Reinforcement Learning Module (JReLM)**
- “Toolbox” permitting experimentation with a wide variety of learning methods (Roth-Erev, Temp Diff/Q-learning,...)

◆ Bid/offer-based optimal power flow formulation

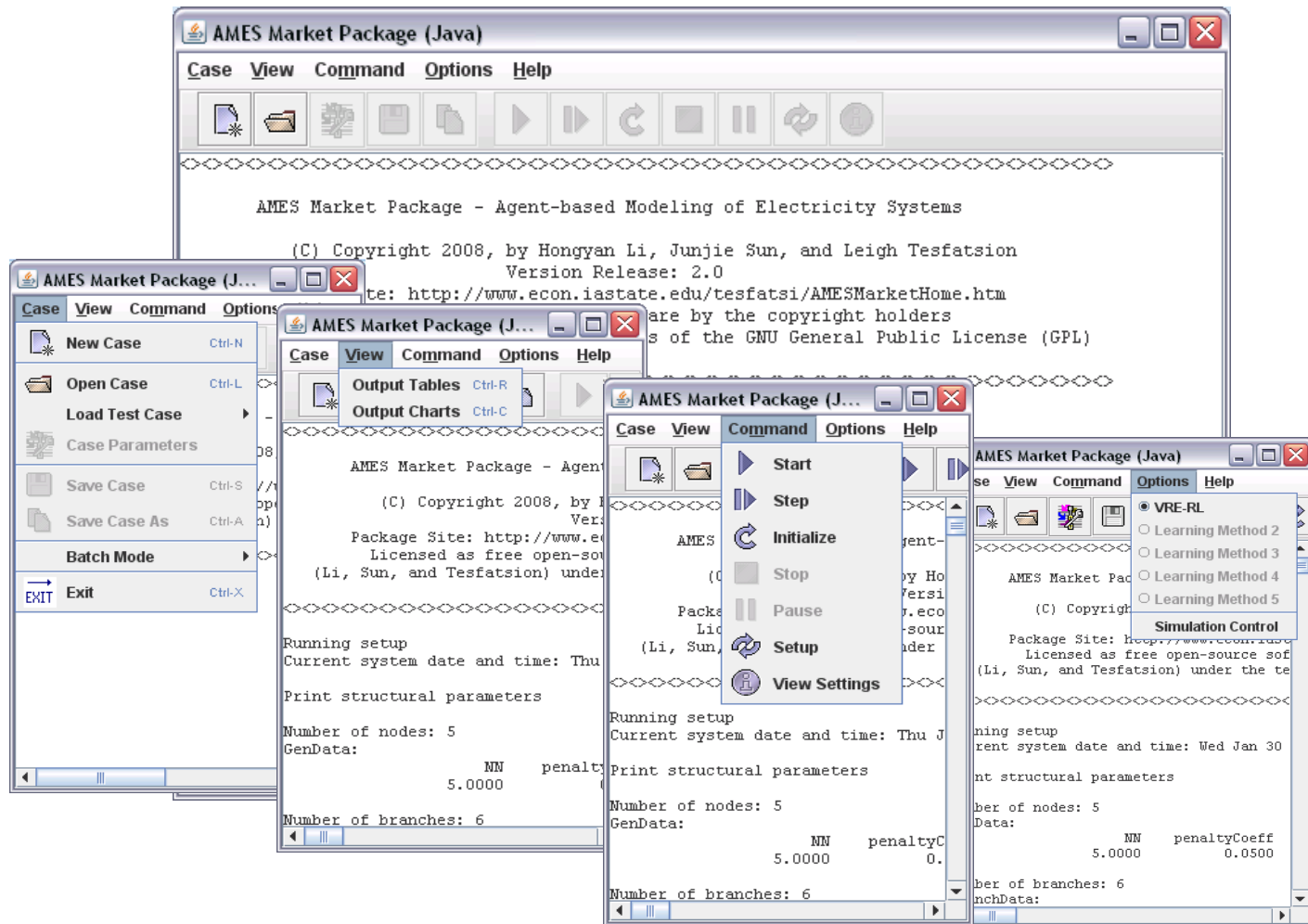
- **Java DC Optimal Power Flow Module (DCOPFJ)**
- Permits experimentation with various DC OPF formulations

◆ Output displays and dynamic test cases

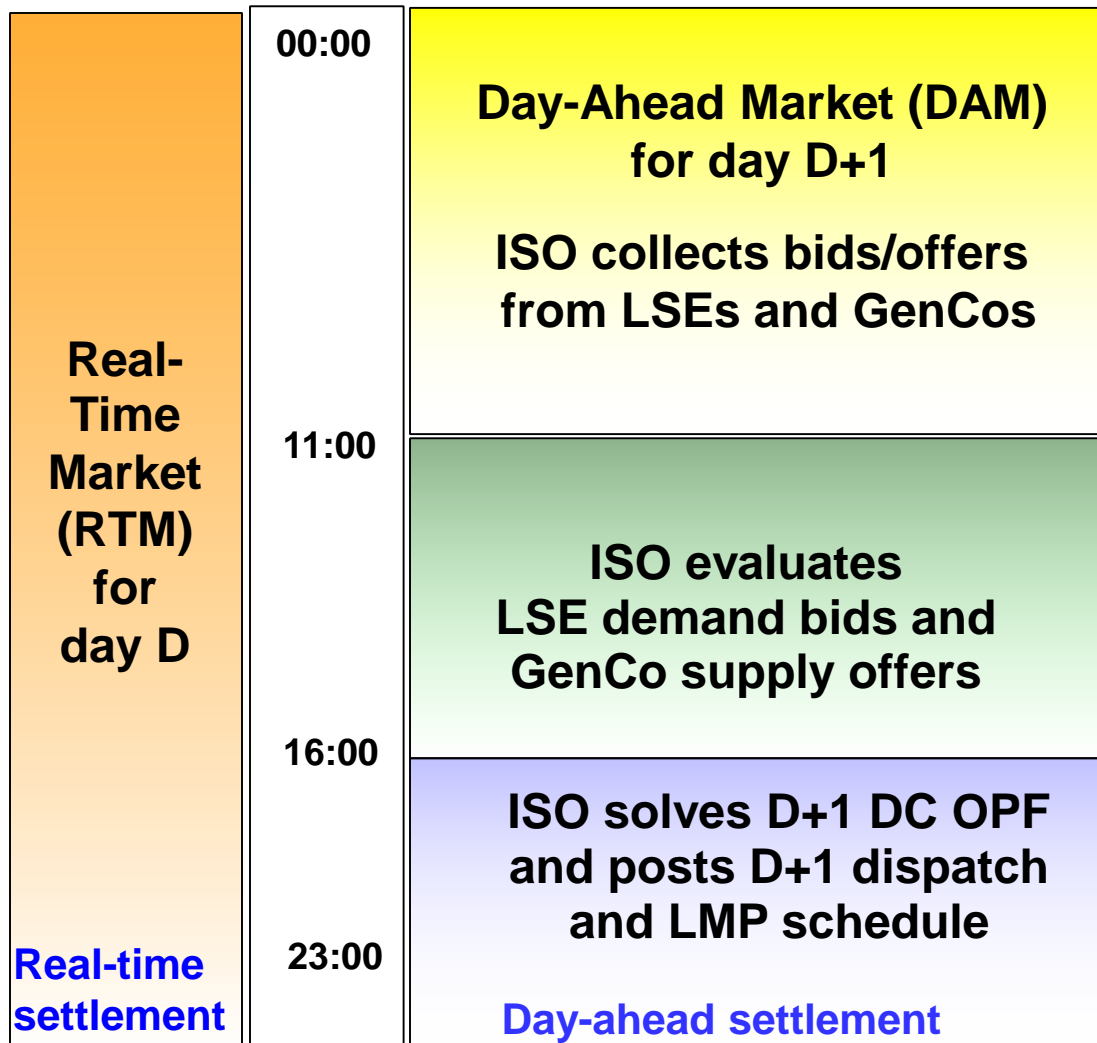
- Customizable chart/table displays & 5-bus/30-bus test cases

AMES Graphical User Interface (GUI)

Tool Bar and Menus for Data Input and Output Displays



Activities of AMES ISO During Each Operating Day D: Timing Adopted from Midwest ISO (MISO)



AMES LSE Hourly Demand-Bid Formulation

- ◆ Hourly demand bid for each LSE j

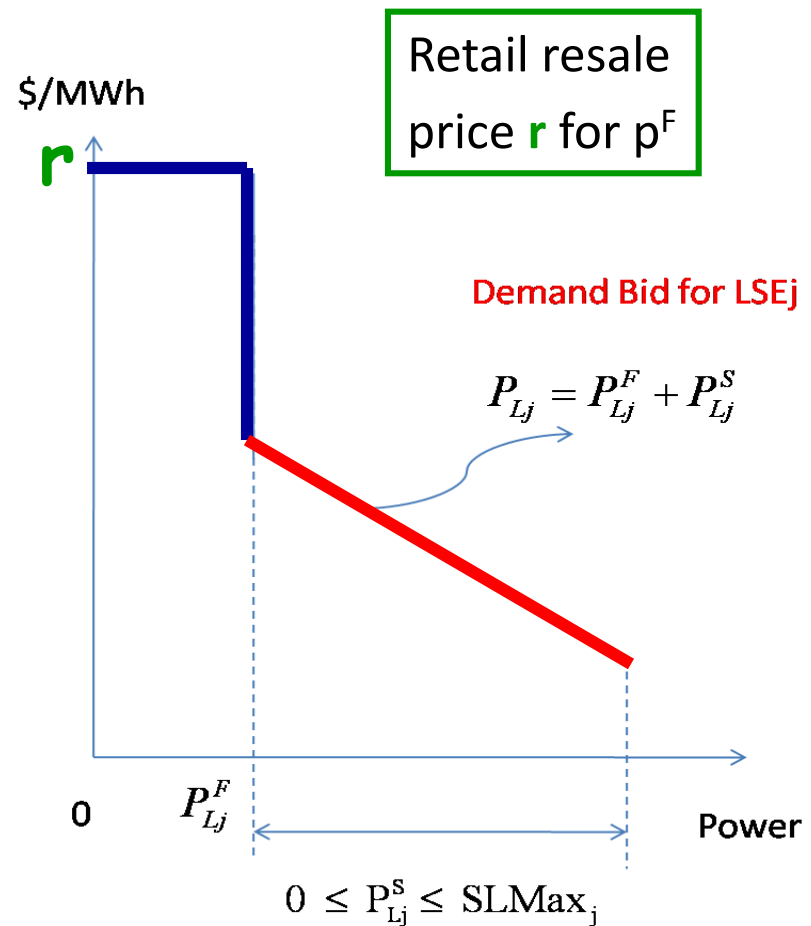
Fixed + Price-Sensitive Demand Bid

- **Fixed** demand bid = p_{Lj}^F (MWs)

- **Price-sensitive** demand bid
= Inverse demand function for real power p_{Lj}^S (MWs) over a purchase capacity interval:

$$F_j(p_{Lj}^S) = c_j - 2d_j p_{Lj}^S$$

$$0 \leq p_{Lj}^S \leq \text{SLMax}_j$$



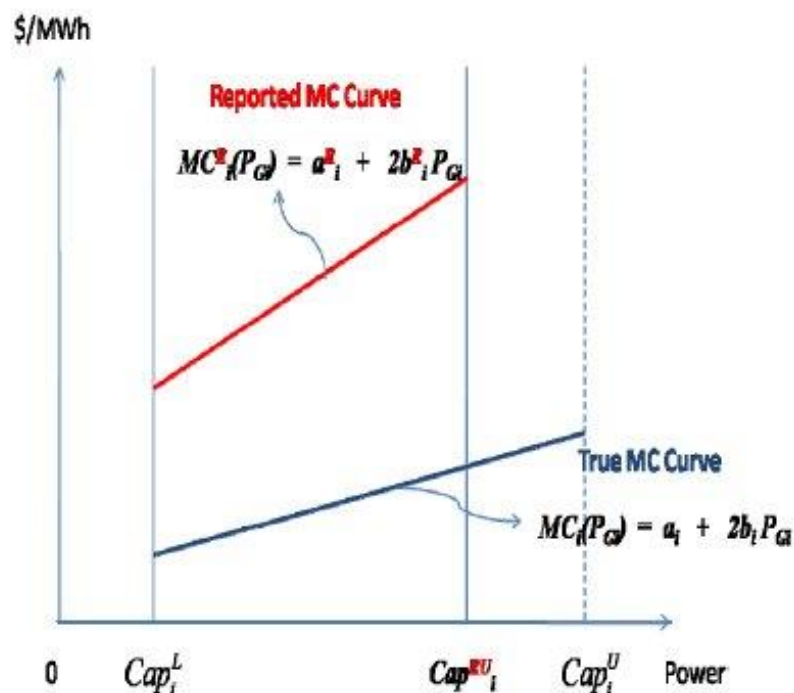
AMES GenCos are learners who report strategic supply offers to ISO for DAM

Hourly supply offer for each GenCo i = **Reported** linear marginal cost function over a **reported** operating capacity interval for real power p_{Gi} (in MWs):

$$MC_i^R(p_{Gi}) = a_i^R + 2b_i^R p_{Gi}$$

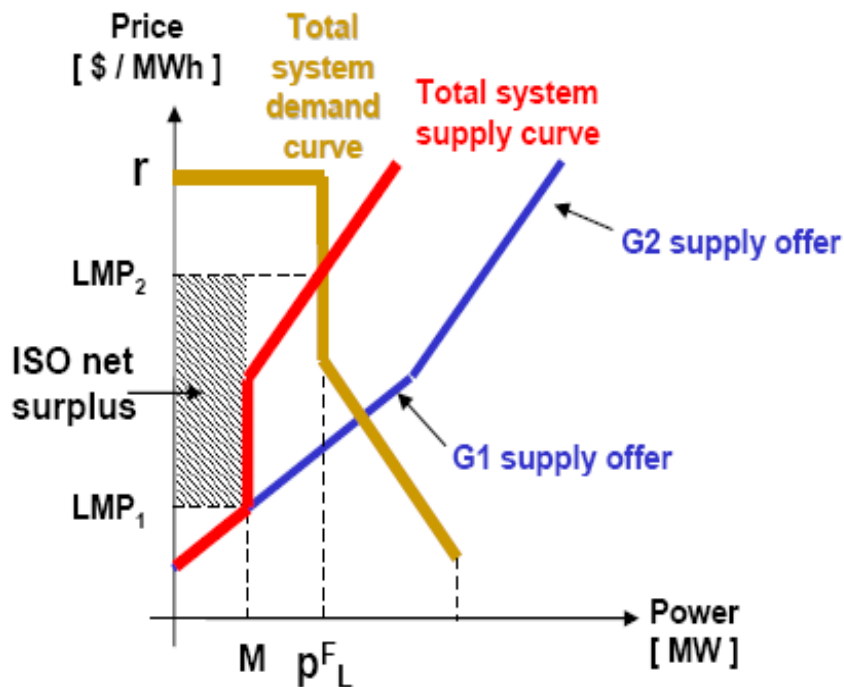
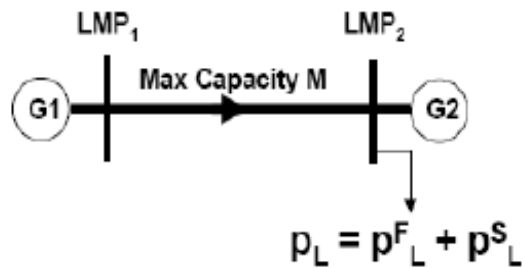
$$Cap_i^L \leq p_{Gi} \leq Cap_i^{RU}$$

GenCos can learn to report **higher-than-true** marginal costs and/or to report **lower-than-true** maximum capacity.



ISO Goal is Max[Total Net Surplus] subject to trans & gen constraints: 2-bus example

(Adapted from H. Salazar, M.S. Thesis, 2008)



Cleared load = p_L^F . LSE at bus 2 pays $LMP_2 > LMP_1$ for each unit of p_L^F . M units of p_L^F are supplied by cheaper G1 at bus 1 who receives only LMP_1 per unit.

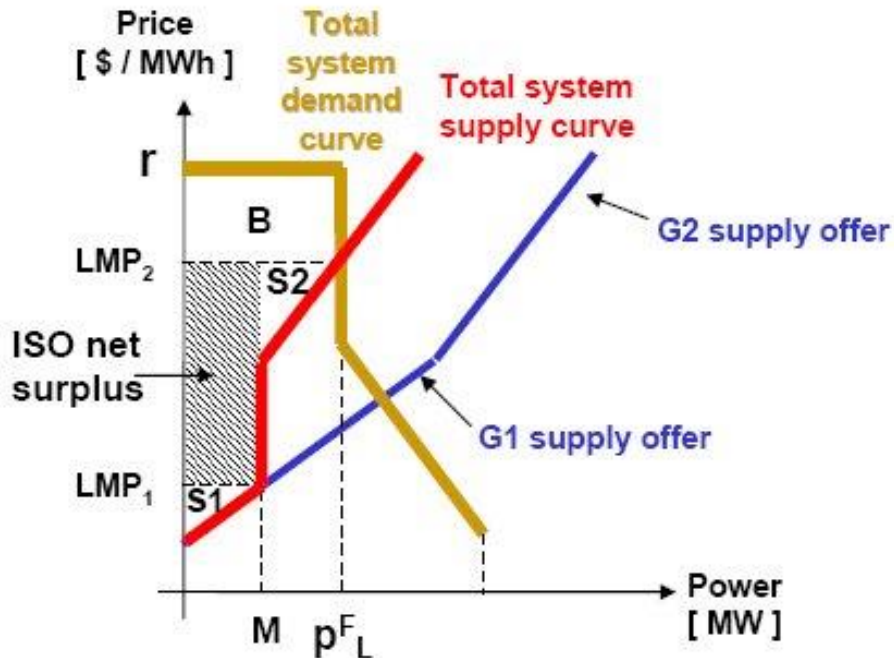
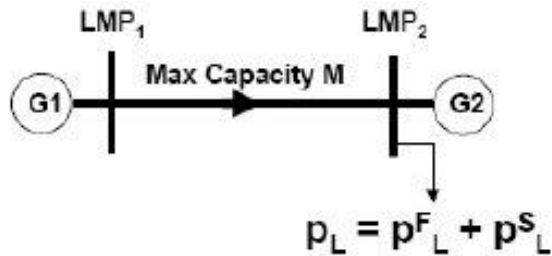
ISO collects difference:

ISO Net Surplus

$$= [\text{LSE Payments} - \text{GenCo Revenues}]$$

$$= M \times [LMP_2 - LMP_1]$$

Calculation of TNS: 2-Bus Example ... Cont'd



ISO Net Surplus:

$$INS = M \times [LMP_2 - LMP_1]$$

GenCo Net Surplus:

Area S1 + Area S2

LSE Net Surplus:

Area B

Total Net Surplus:

$$TNS = [INS + S1 + S2 + B]$$

ISO Objective (DC-OPF):

maximize **TNS** subject to
trans/gen constraints.

AMES ISO Solves Hourly DC-OPF for Day-Ahead Market:

SI unit representation for AMES ISO's DC-OPF problem for hour H of the day-ahead market on day D+1, solved on day D.

DC-OPF formulation is derived from AC-OPF under three assumptions:

(a) Resistance on each branch $km = 0$

(b) Voltage magnitude at each bus $k =$ base voltage V_o

(c) Voltage angle difference $d_{km} = [\delta_k - \delta_m]$ across each branch km is small so that $\cos(d_{km}) \cong 1$ and $\sin(d_{km}) \cong d_{km}$

$$\max \text{TNS}^R \quad (15)$$

with respect to LSE real-power price-sensitive demands, GenCo real-power generation levels, and voltage angles

$$p_{Lj}^S, j = 1, \dots, J; p_{Gi}, i = 1, \dots, I; \delta_k, k = 1, \dots, K \quad (16)$$

subject to

(i) a real-power balance constraint for each bus $k=1, \dots, K$:

$$\sum_{i \in I_k} p_{Gi} - \sum_{j \in J_k} p_{Lj}^S - \sum_{km} P_{km} = \sum_{j \in J_k} p_{Lj}^E \quad (17)$$

where, letting x_{km} (ohms) denote reactance for branch km , and V_o denote the base voltage (in line-to-line kV),

$$P_{km} = [V_o]^2 \cdot [1/x_{km}] \cdot [\delta_k - \delta_m]$$

(ii) a limit on real-power flow for each branch km :

$$|P_{km}| \leq P_{km}^U \quad (18)$$

(iii) a real-power operating capacity interval for each GenCo $i = 1, \dots, I$:

$$\text{Cap}_i^L \leq p_{Gi} \leq \text{Cap}_i^U \quad (19)$$

(iv) a real-power purchase capacity interval for price-sensitive demand for each LSE $j = 1, \dots, J$:

$$0 \leq p_{Lj}^S \leq \text{SLMax}_j \quad (20)$$

(v) and a voltage angle setting at angle reference bus 1:

$$\delta_1 = 0 \quad (21)$$

$\text{TNS}^R =$ Total Net Surplus based on *reported* GenCo marginal cost functions rather than *true* GenCo marginal cost functions.

Lagrange multiplier (or "shadow price") solution for the bus- k balance constraint (17) gives the LMP_k at bus k

Illustrative AMES Experimental Findings for a 5-Bus Test Case

Definition: *Incentive misalignment* → Institutional design fails to align incentives of market participants with efficiency (non-wastage of resources), robustness (resilience), and/or social welfare (socially desirable distribution of total net surplus)

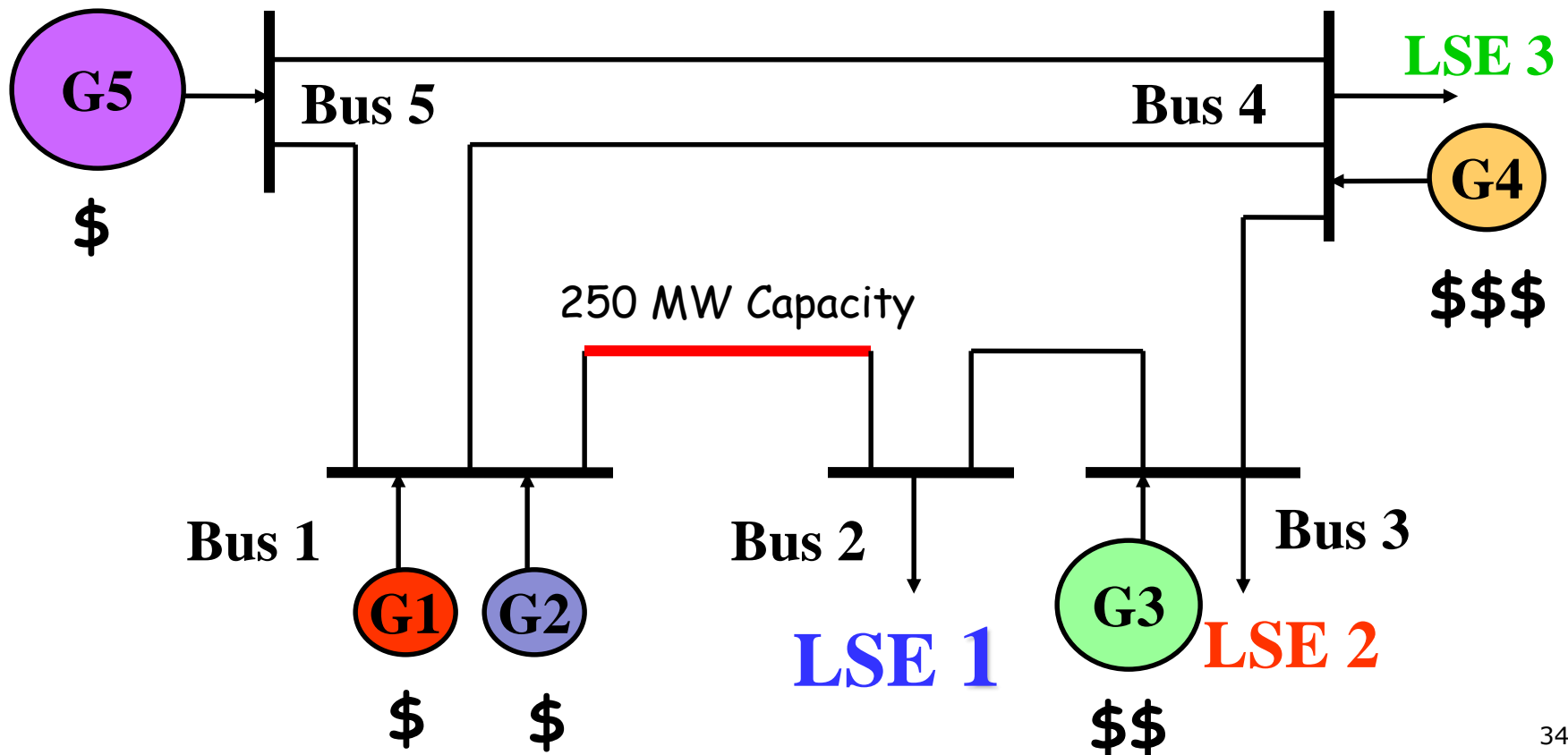
Experiments Reported Below: Incentive misalignment problems under FERC wholesale power market design for a range of experimental treatments:

- **Generator learning** [intensive parameter sweep]
- **Sensitivity of wholesale demand to price** [0 to 100%]

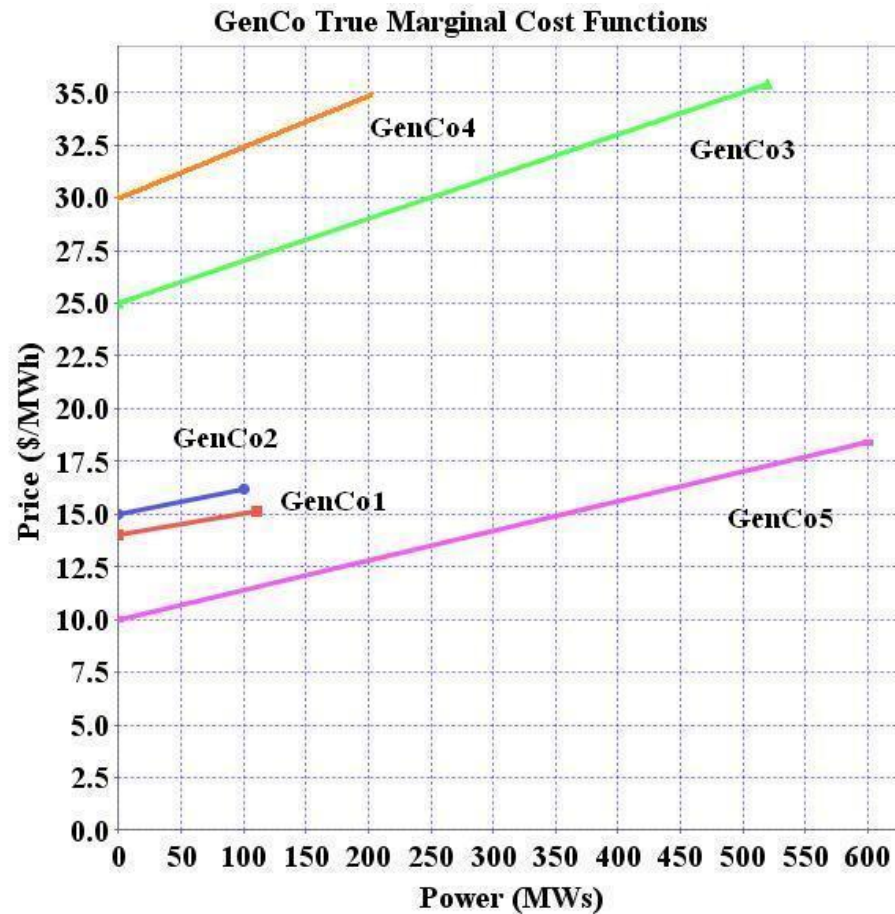
5-Bus Transmission Grid

(Used in many ISO business practice/training manuals)

Five GenCo sellers G1,...,G5 and three LSE buyers LSE 1, LSE 2, LSE 3

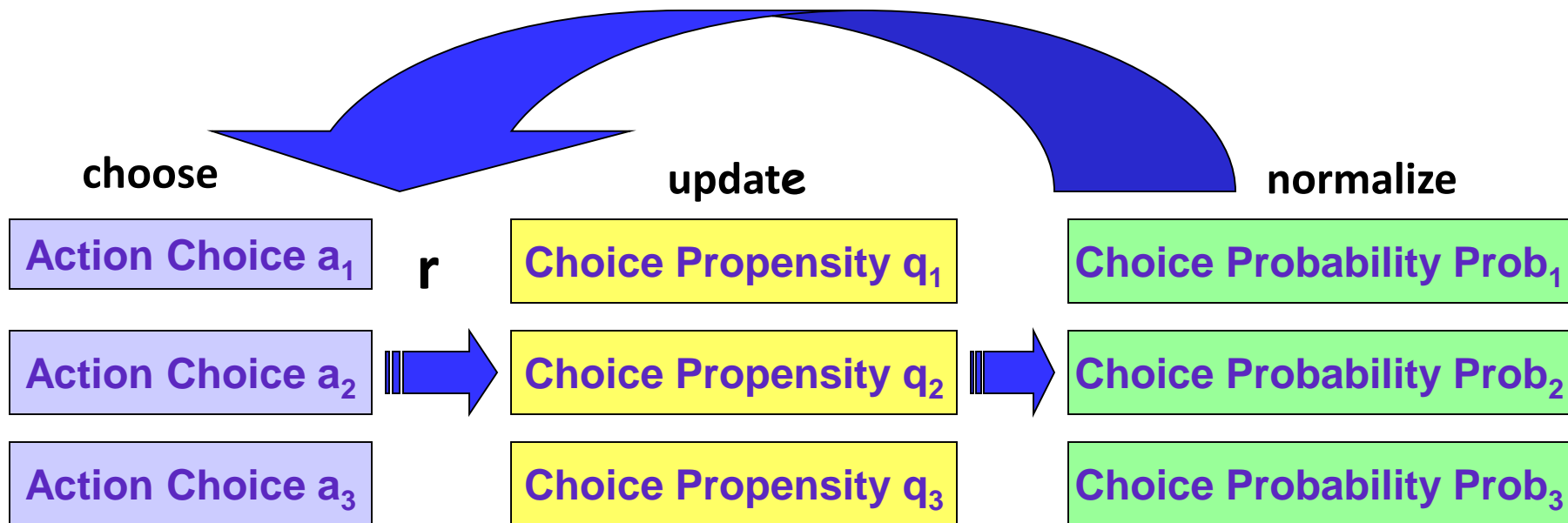


GenCo True Cost & Capacity Attributes



In 5-bus study, AMES GenCos use VRE learning

(version of Roth-Erev stochastic reinforcement learning)



- Each GenCo maintains action choice propensities q , normalized to choice probabilities $Prob$, to choose actions (supply offers). A good (bad) reward r_k resulting from an action a_k results in an increase (decrease) in both q_k and $Prob_k$.

R Measure for Demand-Bid Price Sensitivity

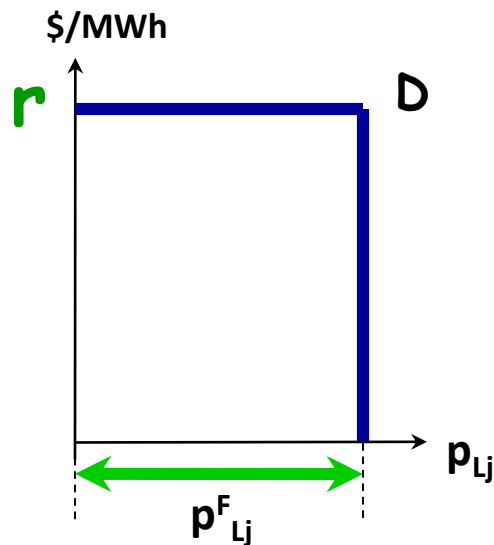
Note: In actual U.S. ISO energy regions, $R \cong 0.01$

For LSE j in Hour H :

p_{Lj}^F = Fixed demand for real power (MWs)

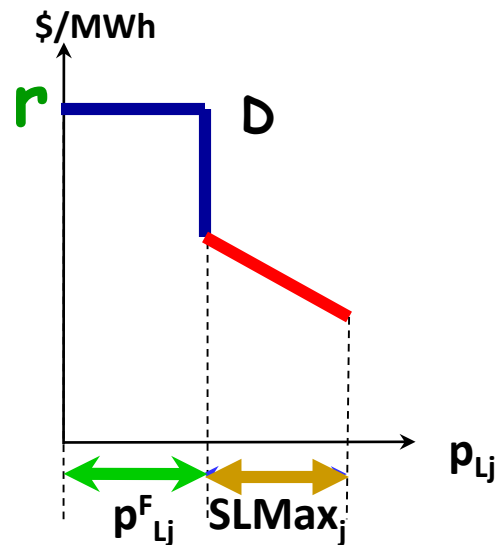
$SLMax_j$ = Maximum potential price-sensitive demand (MWs)

$$R = SLMax_j / [p_{Lj}^F + SLMax_j]$$

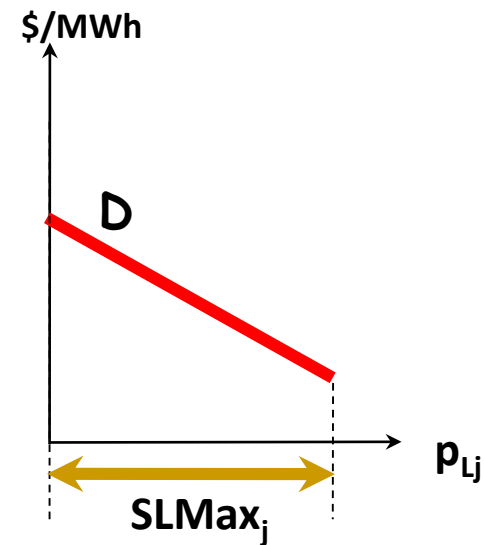


R=0.0

(100% Fixed Demand)



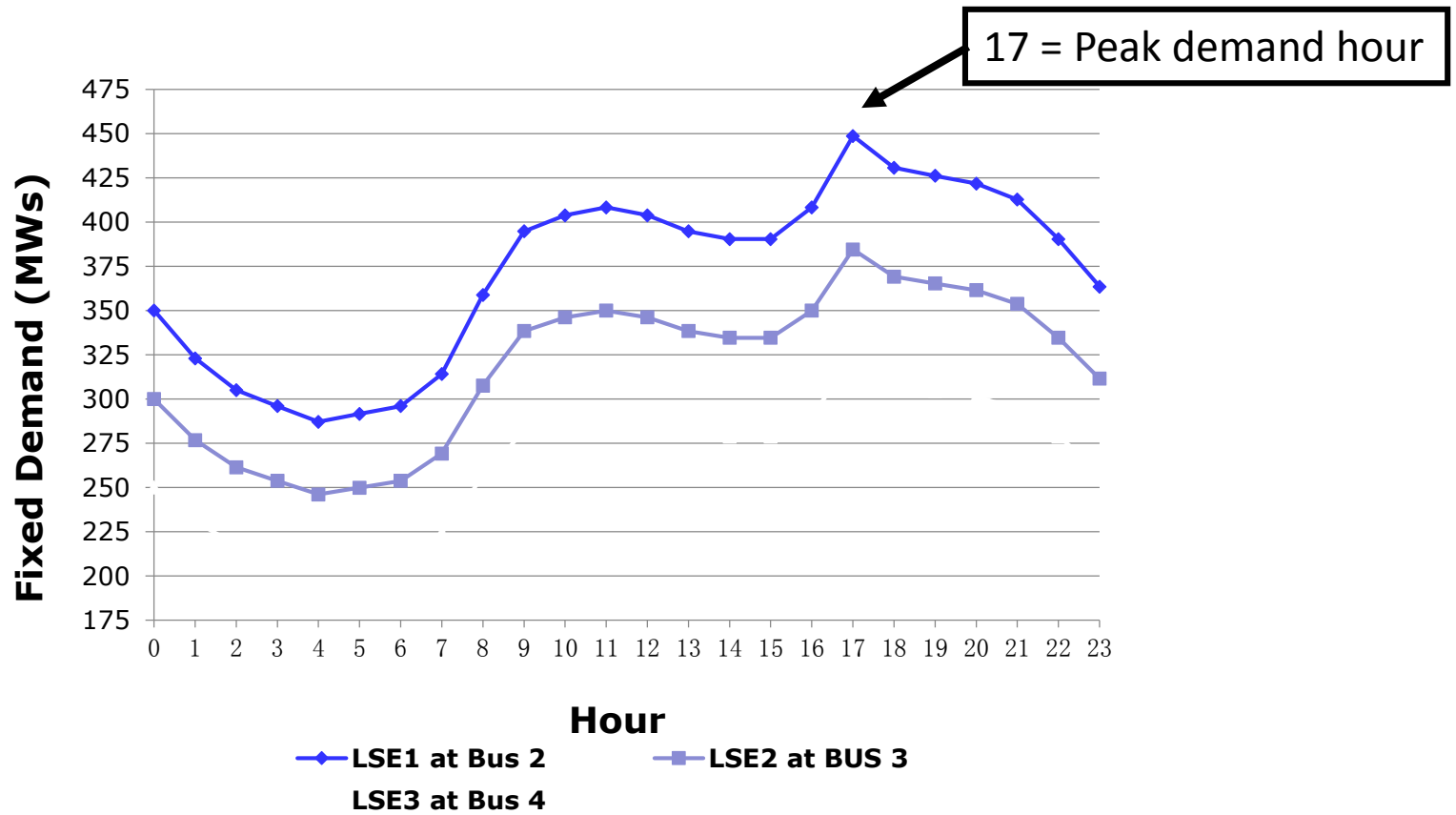
R=0.5



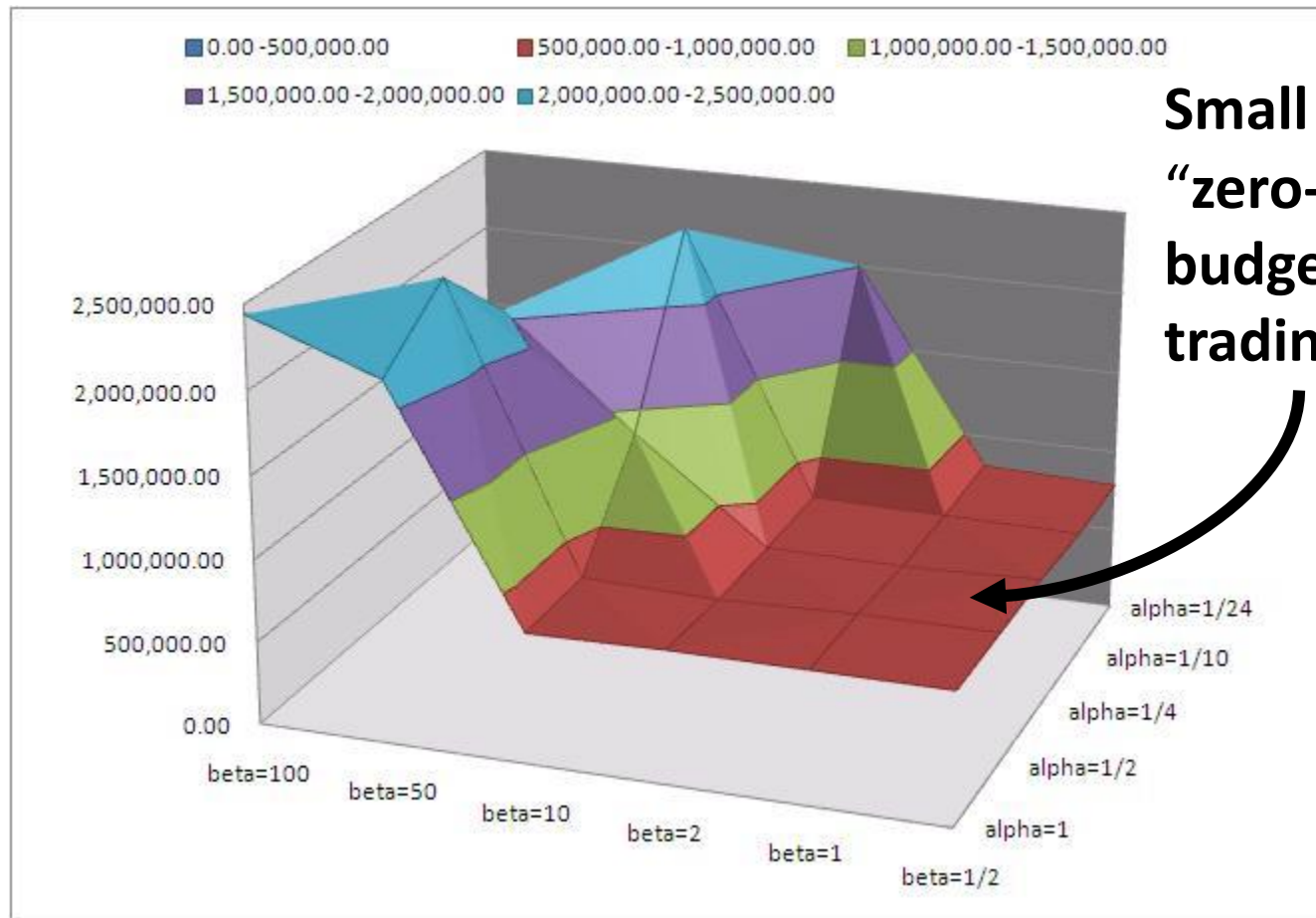
R=1.0

(100% Price-Sensitive Demand)

LSE Hourly Fixed Demands for R=0.0



First Experiments: Avg GenCo net earnings (Day 1000) for R=0 under varied learning parameter settings



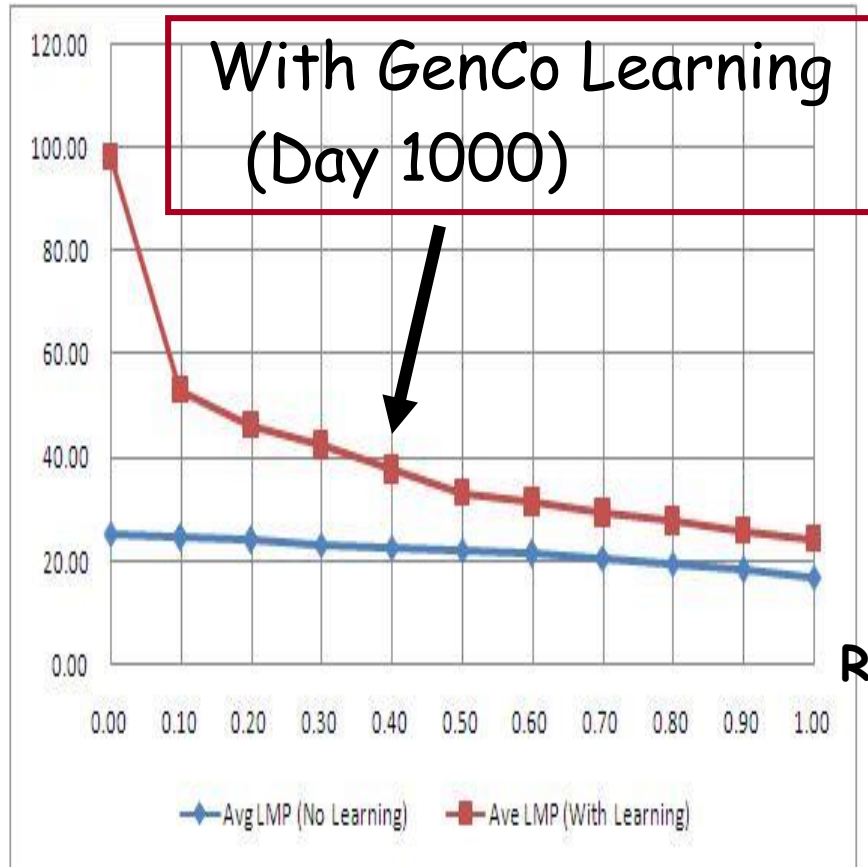
Small beta \cong
“zero-intelligence”
budget-constrained
trading.

**Implication:
Learning
matters !**

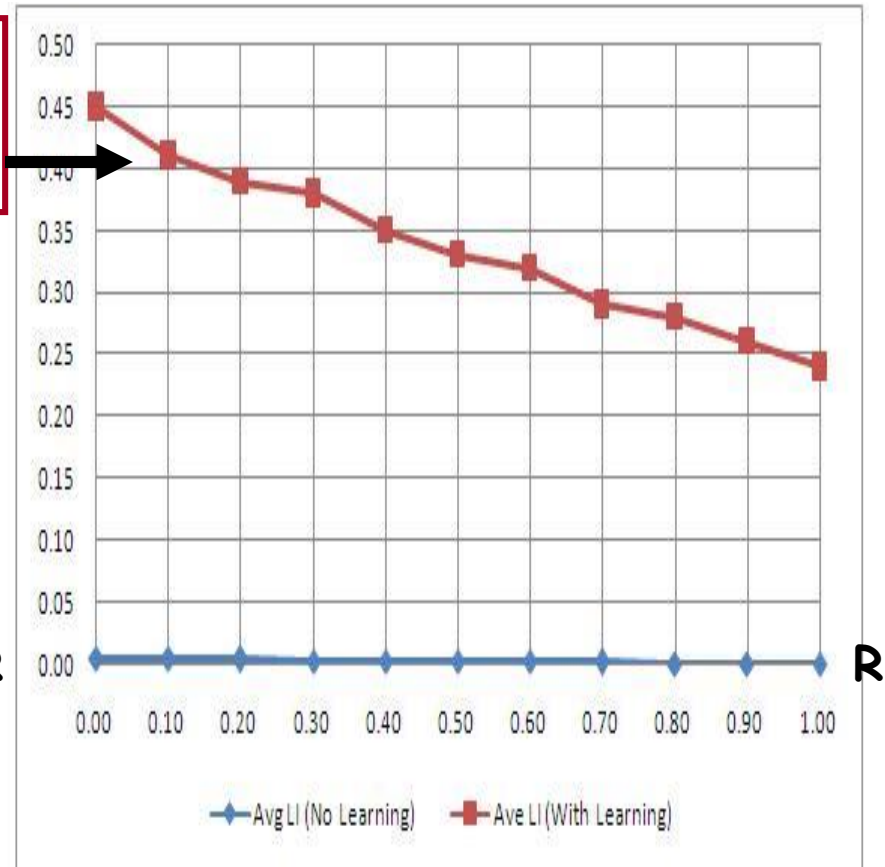
**■ = Sweet spot
region**

Second Experiments: Avg LMP with/without GenCo learning as demand varies from R=0 (100% fixed) to R=1 (100% price sensitive)

Avg LMP (locational marginal price)

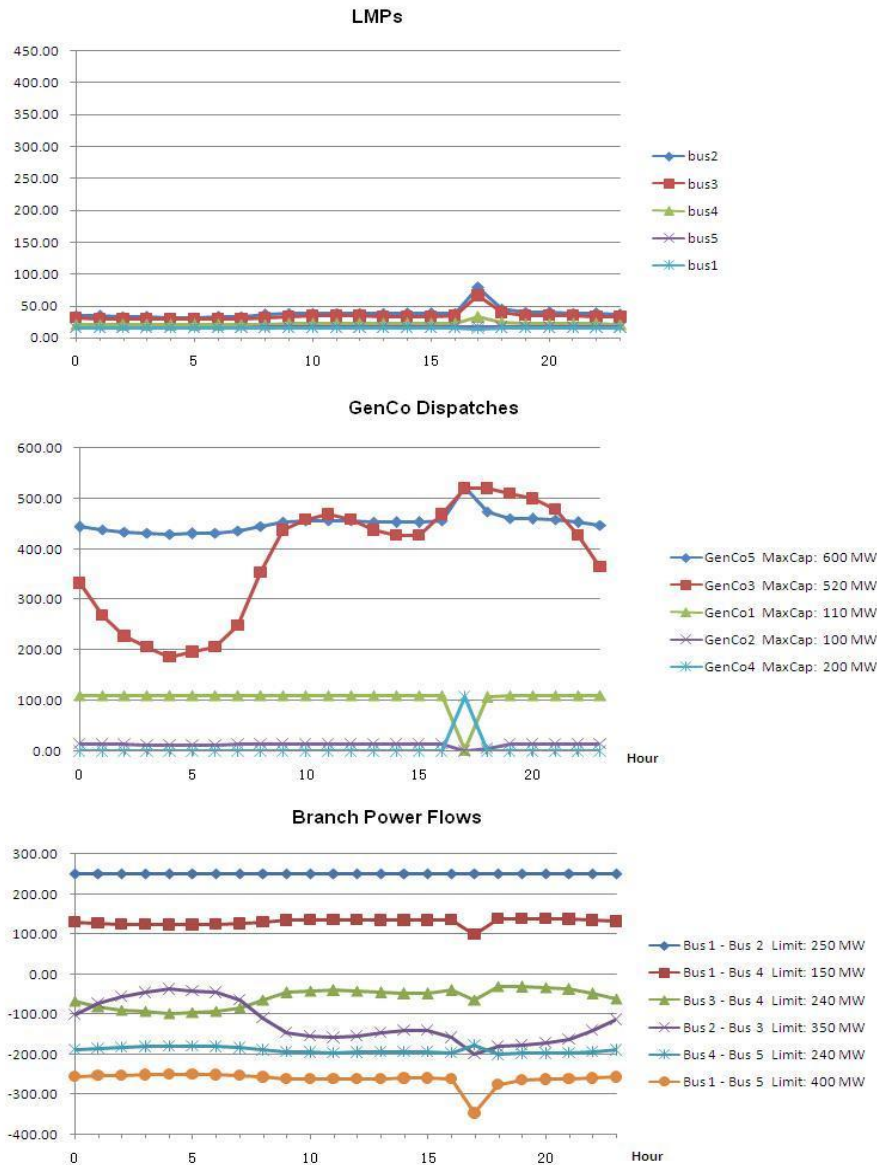


Avg LI (Lerner Market Power Index)

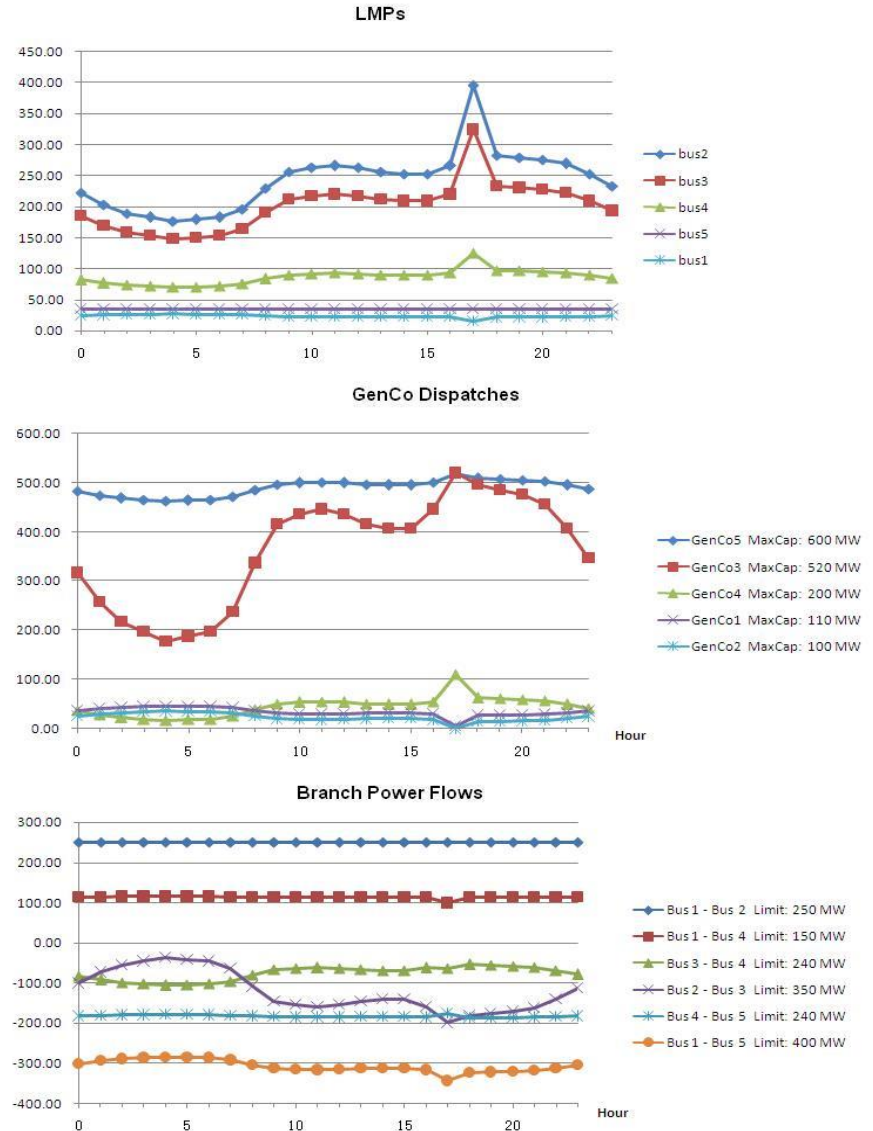


Single-Run Illustration of Findings for R=0.0 (100% Fixed Demand)

W/O Gen Learning (Day 1000)



With Gen Learning (Day 1000)



Implications of Second Experiments

(Li/Sun/Tesfatsion in *Comp Methods in Economic Dynamics*, 2010)

★ BOTTOM LINE:

For all R, prices (LMPs) much higher under GenCo learning due to strategic GenCo supply offers

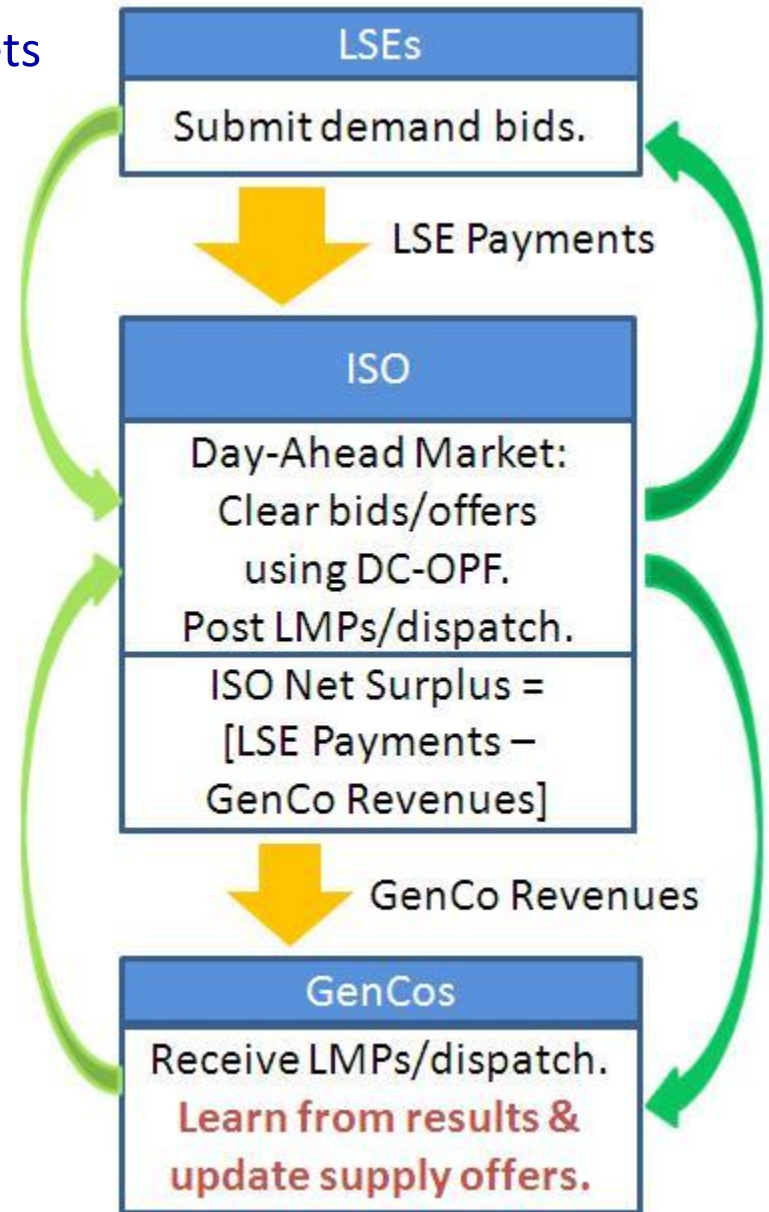
★ NEEDED:

Active price-sensitive LSE demand bidding to offset power of strategic GenCos (well-working double auction)

★ POSSIBLE MEANS:

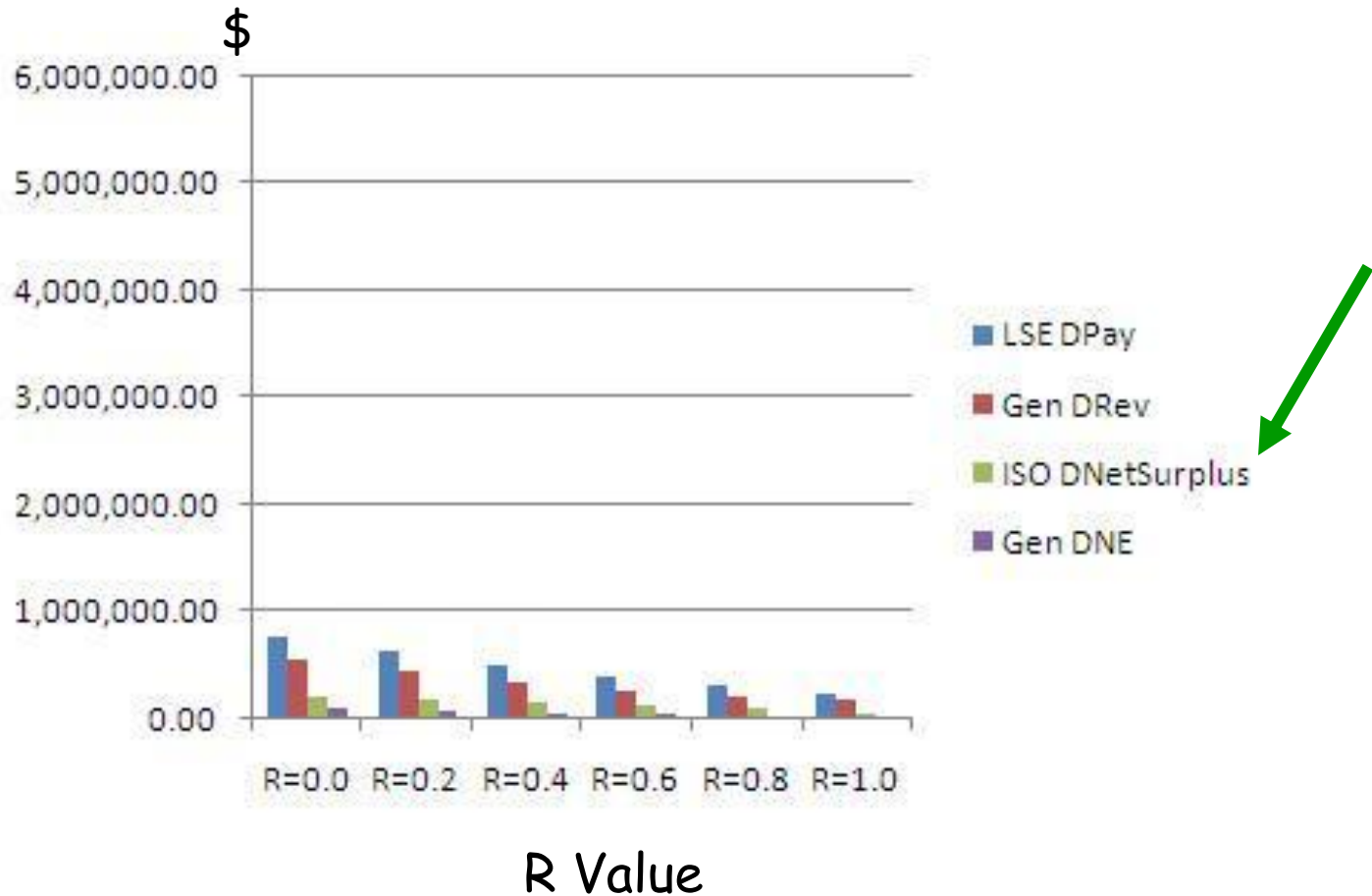
Integrated wholesale/retail restructuring providing array of price-sensitive retail contracts and permitting retail consumers to select their LSE suppliers

Third Experiments: Extraction of net surplus by ISOs in day-ahead energy markets under Locational Marginal Pricing (LMP):

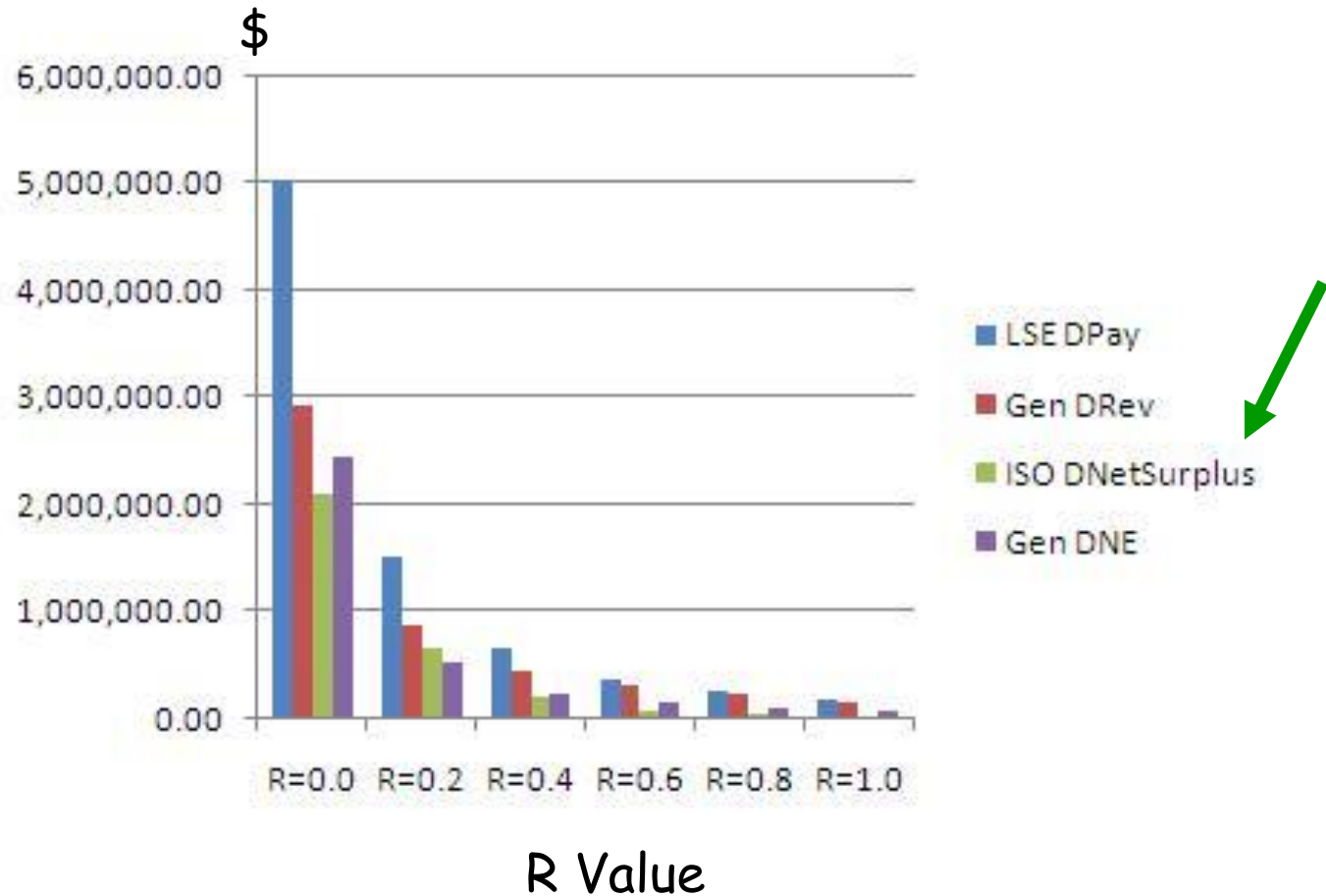


Day-ahead market activities on a typical operating day D:

5-Bus Test Case Results **** Without **** GenCo Learning: ISO net surplus on Day 1000 as LSE demand varies from R=0.0 (100% fixed) to R=1.0 (100% price sensitive)



5-Bus Test Case Results ****With**** GenCo VRE Learning: Mean ISO net surplus on Day 1000 as LSE demand varies from R=0.0 (100% fixed) to R=1.0 (100% price sensitive)



Empirical Comparisons

- **From PJM 2008 report:**

ISO net surplus from day-ahead market: **\$2.66 billion**

- **From MISO 2008 report:**

ISO net surplus from day-ahead market: **\$500 million**

- **From CAISO 2008 report:**

ISO net surplus from day-ahead inter-zonal congestion charges:
\$176 million.

- **From ISO-NE 2008 report:**

Combined ISO net surplus for real-time and day-ahead markets:
\$121 million.

Implications of ISO Net Surplus Findings

(Li/Tesfatsion, *IEEE Transactions on Power Systems*, 2010, to appear)

- ❑ ISO net surplus extractions *not well aligned with market efficiency*
- ❑ Treatments resulting in *greater* GenCo economic capacity withholding (hence higher & more volatile LMPs) also result in *greater* ISO & GenCo net surplus
- ❑ ISO net surplus collections should be allocated for *ex ante* remedy of structural/behavioral problems that encourage GenCo economic capacity withholding.
- ❑ Should not be used *ex post* for LMP payment offsets and LMP risk hedge support (current norm)

Conclusions

- * **Restructured wholesale power markets** are complex large-scale systems encompassing physical constraints, institutional rules of operation, and strategic human participants.
- * **Agent-based testbeds** permit the systematic dynamic study of such systems through intensive computational experiments.
- * For increased empirical validity, testbeds should be **iteratively developed** with ongoing input from actual market participants.
- * To increase usefulness for research/teaching/training and to aid knowledge accumulation, testbeds should be **open source**.

On-Line Resources

Presentation Slides

<https://www2.econ.iastate.edu/tesfatsi/DARPAISATJuly2010.LTTalk.pdf>

AMES Test Bed Homepage (Code/Manual/Publications)

<https://www2.econ.iastate.edu/tesfatsi/AMESMarketHome.htm>

Agent-Based Electricity Market Research

<https://www2.econ.iastate.edu/tesfatsi/aelect.htm>

Agent-Based Computational Economics Homepage

<https://www2.econ.iastate.edu/tesfatsi/ace.htm>

ISU Electric Energy Economics (E3) Group

<https://www2.econ.iastate.edu/tesfatsi/E3GroupISU.htm>