"Self-Regulating Electricity Markets?" by Nodir Adilov, Thomas Light, Richard Schuler, William Schulze, David Toomey & Ray Zimmerman

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Advanced Workshop in Regulation and Competition Rutgers Center for Research in Regulated Industries 17th Annual Western Conference San Diego, CA June 24, 2004

Abstract

An experimental structure is demonstrated that represents end-use customers in electricity markets who can substitute part of their usage between day and night. Each customer's demand relationship is represented by a two-step value function for each period, disaggregated from observed market demand relationships, that varies between day and night and during heat-waves. Three alternative demand-side market structures are evaluated: 1) customers pay the same fixed price (FP) in all periods - - the base case, 2) a demand response feature (DRP) is added to the fixed price case in periods of supply shortages, wherein buyers receive a pre-specified credit for reduced purchases, and 3) a real time pricing (RTP) case where prices are forecast for the upcoming day/night pair, then buyers select their quantity purchases sequentially and are charged the actual market-clearing price, period-by-period.

After demonstrating the ability of buyers to make efficient purchases, six experienced sellers with experience in exercising market power were paired with seventeen buyers over twenty two auctions (eleven day-night pairs) that included heat waves and unit outages. The same periods were repeated under each of the three different market treatments, and the RTP structure resulted in the greatest market efficiency, despite the difficult cognitive problem it poses for buyers. Both DRP and RTP reduced the severity of price spikes as compared to the FP structure. A preference poll comparing DRP and RTP was conducted after each treatment. In one experiment, 74% of the participants said they preferred DRP before trying RTP, but 64% chose RTP afterward, a statistically significant reversal of preferences, and in a second experiment, 53% preferred DRP initially, but 68% selected RTP after experiencing both treatments. Finally, the relationship between total system load and line flows was examined under each of the three market treatments and for a simulated fully-regulated regime. The relationship demonstrated under the regulated regime deteriorates under FP, but is re-established under the DRP and RTP market structures.

1. Introduction

Can electricity markets be more self-regulating if we encourage customers to enter the game as active participants? Existing markets for electricity in the United States emphasize wholesale exchanges between generators and marketers and/or utility distribution companies that act as intermediaries between generators and retail customers. Furthermore, that market structure is usually just a variation of the least-cost, optimal power flow dispatch paradigms that were used before deregulation by the utilities' power pool managers. What has changed is that offers are substituted for cost schedules on the supply side, and an ISO/RTO clears the market and determines the dispatch, but the demand is still an aggregation of schedules submitted by the load serving entities that are buyers, with very little, if any, price response associated with that forecast demand quantity.

Effectively, these wholesale electricity markets are single-sided, and both in previous experimental studies in the laboratory (see Bernard, et. al. [3] and Mount et. al. [5]) and in the experiments of the whole conducted on electricity markets in California, the ability of a number of suppliers to drive prices well above competitive levels is demonstrated in these markets that are repeated frequently. Consequently, routine regulatory interventions like price caps and automatic market protection (AMP) mechanisms have been introduced in most jurisdictions. But rarely are the customers active participants in these markets, and where retail competition exists, it usually is between aggregators who offer a constant price for electricity in all periods. Rarely are customers exposed to the real time cost of their buying decisions. Suppose a larger portion of customers were confronted with the fact that in some periods their cost of acquiring electricity spikes anywhere from five to twenty times their average price? Would they be willing to alter their consumption patterns, and if so, might that mute some of the suppliers' potential market power? To what extent could some of the regulatory band-aids that have been applied over the past several years to these markets be reduced or removed? Would a power exchange with active demand-side participation be harder or easier to operate than existing systems?

To address these questions, simple mechanisms are studied that permit customers to participate actively in electricity markets. Then through experimental analyses, where the customers' valuation of electricity in different periods is calibrated according to historically observed usage patterns, the buyers' collective ability to perform efficiently and to offset attempts at exercising market power by generators is analyzed. The relationship between flows on individual lines and total system load is also explored for these different market structures.

While theoretical constructs suggest that the overall efficiency of any market should be improved with both active buyers and sellers, that theory does not prescribe how many participants are required to achieve that efficiency, nor how the number of required sellers varies as the number of active buyers increases. In fact all benchmarks about the number of participants required to achieve efficient markets are empirically-based, like the Herfindahl-Hirschman Index (HHI) frequently used by the Ant-trust Division of the US Department of Justice to gauge potential market power in an industry. Furthermore, these guidelines are all developed from experience with industries whose characteristics are unlike the unique aspects of electricity markets: there's only one way to transport electricity on a system subject to congestion, transport feasibility is governed primarily by physical, not commercial, laws, and production and usage must be matched in real time (inventories are not economically feasible) or else the entire system will collapse (a blackout).

Questions to be asked include: how do buyers actually perform under different demandside structures, given a choice, which structure might they select voluntarily, and to what extent do different demand-side structures succeed in muting the exercise of market power by suppliers and lead to self-regulating markets? These are practical questions that are amenable to experimental analysis; game-theoretic analyses of repeated markets where sellers have multiple units that they can choose to supply are too complex to develop definitive conclusions. And while promising large-scale tests of alternative demand-side structures have been conducted by utilities around the country with their customers, the results are frequently difficult to generalize because of the diversity of customers and the elapsed time of the tests during which other things can change. As an alternative, a representative demand-side structure is developed and used in controlled laboratory experiments with human participants.

2. Laboratory Experimental Analyses

Rassenti, Smith and Wilson [7] have conducted illustrative two-sided market experiments to represent what might occur in electricity markets were customers let into the game. Their structure clears price and quantity, bids and offers that are made simultaneously into a real-time energy market with four buyers, five sellers and one computer-simulated buyer. The participants face three different demand periods in a day (peak, shoulder and off-peak periods). As in all laboratory experiments with buyers, the actual valuation of purchases must be pre-assigned (induced valuations) and the participants must be paid in proportion to the difference between the assigned valuation for the electricity purchased and the price paid. In Rassenti, Smith and Wilson's experiment, buyers were assigned a multi-step demand relationship calibrated so that the maximum possible combined reduction in the quantity demanded was 16 percent. Their results [7] illustrate the potential for an active demand-side to completely eliminate the exercise of market power by suppliers.

Using an alternative methodology of "preprogrammed" autonomous agents in a numerical simulation of a two-sided market, Talukdar, et. al. [8] have shown how buyers' and sellers' propensity to "learn" how to maximize their gains can offset each other. This agent-based simulation methodology has the advantage of being able to replicate hundreds of market periods with a large number of participants much more rapidly and at

a lower cost than in laboratory experiments with human subjects (human subjects must receive appreciable compensation in proportion to their performance in order for the experimental outcomes to be valid). The suspicion about numerical simulations, particularly when not preceded by controlled experiments with humans that reveal their cognitive processes, is that the outcomes are biased by the agent's "learning" mechanisms that are pre-programmed, and that these simulations can never accurately reflect the cognitive insights and/or limitations that are inherently human.

In this analysis, therefore, a demand-side platform was constructed and tested that is representative of the decisions that electricity customers would have to make in real-time markets, and whose valuations are calibrated to reflect previous statistical analyses of aggregate buyer behavior. In particular, since much of the response by customers to demand response programs and real time pricing has been to shift a portion of their usage to adjacent time periods, it was essential to incorporate this inter-temporal decisionmaking into the demand side platform. As a consequence, the demand-side representations that are tested here can be used to address many other important issues in the future, including tests on markets for reserves, forward markets, etc. However, the current experiments are designed to demonstrate that a representative mechanism is available for future analyses of three alternative (and/or in various combinations) forms of demand-side participation in electricity markets: a pre-announced Demand Response Program (DRP), a Real Time Pricing (RTP) program, and as a base case for comparison, the pre-specified identical Fixed Price (FP) charged in every usage period by most utilities today. This analysis also tests the relative efficiencies of the three alternative demand-side treatments, as well as the participants' subjective preferences in a sequence of polls.

3. Experimental Structure

3.1 Demand-Side Representation

To keep the demand-side decisions simple for the participants, each buyer is assigned a simple two-step discrete demand function with separate valuations for day and for night usage, as shown in Figure 1. In fact, these individual demand relationships are decomposed from an aggregate demand function, shown in Fig. 2 that has a retail price elasticity of demand at the mean price of –.3, Faruqui and George [4]. Furthermore, the overall demand function ranging from very low prices to the reservation price was given the inverted S-shape suggested by Schulze's work (reported by Woo, et. al. [9]) on consumer value loss for interruptible service. Note that each customer's day valuation is somewhat higher than their night valuation. Furthermore, there is an additional "substitutable" block of energy that customers can choose to buy either during the day or the subsequent night period (unused substitutable energy cannot, however, be carried over to the next day/night pair of periods). Typically, substitutable electricity purchases are valued less than the regular purchases in each of these periods, and substitutable night energy is valued less than if it is used during the day. These substitutable blocks were also decomposed from the aggregate demand curve that has an elasticity of substitution between day and night usage of .3, Faruqui and George [4].

Thus, the buyer is confronted with an inter-temporal optimization problem. In addition, these induced valuations are increased substantially in pre-specified periods called "Heat-Waves" to reflect the added value of electricity in extreme climatic conditions. The buyer's problem then is to maximize the spread between their assigned valuation for each quantity of electricity they buy, and the price they have to pay for it.

3.2 Alternative Demand-Side Market Structures Considered

The experiments are designed to test the efficiency of two alternative forms of active demand-side participation in electricity markets. As a base case for comparison, the first set of experiments reflect typical utility pricing where buyers pay a pre-determined fixed price (FP) in all periods and merely determine how much electricity they wish to purchase in each period. In the second treatment, buyers are alerted prior to consumption periods when supply shortages are anticipated. In those periods, customers are given the opportunity of reducing their consumption below their normal benchmark purchases in similar periods, and by doing so they can earn a pre-specified credit per kWh for each unit of electricity less than their benchmark that they choose to buy. This treatment is analogous to the NYISO's Emergency Demand Response (DRP) program. All electricity actually purchased under this DRP scheme is priced at the same fixed price used in the base case, but total customer payments are reduced by any DRP credits earned. The third treatment is a simple real time pricing (RTP) scheme; wherein, price forecasts are announced for the next day and night periods, and based upon these forecasts, buyers decide how much electricity to purchase. However, buyers must pay the actual marketclearing price in each period for their actual purchases, and that price may differ from the forecasted price.

In early pilot experiments, a second type of RTP market structure was also tested. In that alternative, participants were asked to submit prices that reflected their maximumwillingness-to-pay together with each block of quantity bids. In these pilots, each participant's performance was compared for the same three day-night pairs that included heat-waves and supply outages under the FP, DRP, RTP and RTP with limit price bids market structures, in that order. Using experienced undergraduate students, they nevertheless did not perform as well in the RTP runs where they specified a limit price as they did in the previous RTP runs where their only bids were for the quantity they wished to buy (their earnings were 94.9% of optimal, as compared with 98.7% under the simpler, quantity-only bid structure). Because the full experimental runs were to be much longer than the pilots, and participant fatigue was a concern, only the simple form of the RTP market structure was used in the full experiments. This is a mechanism that approximates a scheme used in France where customers are notified a day ahead by color code whether electricity prices are anticipated to be high, moderate or low, and then based upon that information the customers make their quantity purchases, but are charged the actual clearing price (Aubin, et. al.[2]).

In an analytic model of electricity supply and demand that was developed to understand which components might be solved in theory by markets, and which would have to rely

upon regulatory oversight, both DRP and RTP structures emerged as mechanisms for partial de-centralization of electricity supply (See Mount, Schulze & Schuler [6]). Three conclusions resulted. First, since all the customers in a neighborhood served from the same electrical network receive the same level of reliability, regardless of differences in their individual preferences for reliability, the determination of that optimal level of reliability is a public function and must be set by a regulatory authority. Individual private expressions of their valuation of reliability cannot be relied-upon, if a price is attached, because of free-rider problems. Second, while some customers may be willing to interrupt or reduce their level of demand in response to a pre-announced request with a specified credit per kWh of reduction, the optimal response will not be forthcoming from customers unless the credit they receive is equivalent to the forgone reserve and capacity payments that would have been incurred were that reliability provided by additional generation, plus they must save the real-time energy price for electricity not used. In short, efficient demand side participation requires both demand response programs (DRP) and real time pricing (RTP). Third, unless the loss in consumer value from an unanticipated interruption is identical to their loss in value for a planned demand reduction through DRP, the customers' willingness to participate in DRP programs cannot be used to infer the value of reliability. Reliability is a public good and its level can only be set and enforced by a regulatory body; however, once set, that standard can be met efficiently through market mechanisms made available to both suppliers and customers. These analytic results support the experimental evaluation of both DRP and RTP demand-side structures.

3.3 Experimental Tests of Demand-Side Structures (Single-Sided)

Before undertaking experiments on full two-sided markets, the three different demandside platforms were tested with two separate groups of students against a predetermined supply-side that was varied randomly (see Adilov et. al. [1]). The buyers in these experiments were 21 Cornell professional graduate students who were divided into two groups. They received cash compensation in proportion to their individual earnings that was computed as illustrated in Figure 1. However, since each buyer was assigned different valuations for their purchases, an exchange rate was applied to each participant's earnings so that each subject had a fair chance of earning the same amount of money in the experiment. The actual average earnings per player ranged between \$38 and \$48 in these experiments with an active demand-side only.

The supply configuration was based upon previous supply-side experimental results at Cornell University (Mount, et. al. [5]) and of actual offer structures observed in wholesale electricity markets. Thus the typical "hockey-stick" shaped offer function shown in Figure 3 was applied in all cases, except that randomly-selected outages of particular generators caused this offer curve to slide back and forth horizontally in some market periods. Furthermore, to insure that the market always clears, regardless of buyer behavior, sufficient external supplies are always available to meet demand at the highest offer price. In all treatments, the market is conducted for wholesale supplies of electricity, only, but the price paid by customers has a \$.04/ kWh wires charge added to it (demand valuations were calibrated at retail prices). In the FP treatment, the retail price was set at \$.11/ kWh (\$.07/kWh for the average wholesale price for electricity), based on the stochastic offer structure that was predetermined, and the assumption of optimal bidding strategies by the buyers where their valuations are also known by the designers of the experiments.

For the Demand Response Program, the price for retail purchases of electricity remained at \$.11/kWh, but whenever a randomly pre-determined supply shortage occurred, a DRP period was announced and buyers received a credit of \$.25/ kWh for the difference between their benchmark consumption (what they would have bought in that period had they behaved optimally without the DRP credit under fixed-price purchases) and the amounts they actually bought. The DRP credit was computed to include both the estimated savings in the actual wholesale price of electricity for their reduced consumption, plus the pro-rated savings experienced by the market. Note, that under both the FP and DRP treatments, the actual payment by buyers may not equal the cost of purchases from suppliers, unless the actual participants behave optimally since that was the basis for setting the fixed prices and DRP credit. Under the RTP treatment, the buyers were given an accurate estimate of what the wholesale price plus the \$.04/ kWh wires charge would be, were they to make optimal quantity bids. The buyers were also told, given experience with earlier experiments, that they could expect the actual clearing prices to vary by 20%, but they were also told that they would pay the actual marketclearing price. Thus for the RTP treatment, the prices paid and costs of purchases should be identical.

Participants demonstrated their ability to understand the three alternative buying structures acceptably in these experiments that were conducted over eleven day-night pairs (22 periods, total). Those periods included heat waves during which the buyers' valuations are increased and with occasional supply shortages. All three demand-side structures were tested, and RTP resulted in the greatest overall market efficiency, measured as the sum of consumers' and producers' surplus. Under the RTP treatment, participants attained 99.6 % of the socially optimal level, despite the more difficult cognitive problem RTP poses for buyers. By comparison FP efficiency was at 98.7% of the optimum; whereas DRP attained only 96.9% of the socially optimal benchmark. The aggregate consumers' surplus portion of the total surplus was only at 95.7% of its optimal level under FP, but that fraction increased to 97.2% under DRP and to 101.8% under RTP. These results emphasize the inherent problem with single-sided markets: in most existing electricity markets that have primarily active suppliers, the suppliers gain an advantage; in these experiments with only active buyers, the consumers benefit and acquire surplus from the sellers, which explains how consumers' surplus can exceed 100%.

Statistical tests on the differences in consumers' surplus between the demand-side treatments are reported in Table 1. Using a paired t-test with a separate test conducted over the distribution of the subjects' differences in surplus over each of the eleven day-night period pairs, the first three columns in Table 1 show that in most cases, the surplus

deviated significantly from the socially optimum level. However, RTP resulted in greater than optimal consumers' surplus in 7 of 11 pairs; whereas, both FP and DRP resulted in significantly less than optimal consumers' surplus in 7 of 11 pairs (the consumers' surplus for FP was also significantly greater in 2 of 11 pairs, and DRP was significantly greater in 4 of 11 pairs). Comparing the consumers' surplus between FP and RTP, RTP is significantly better in 7 of the 11 periods and worse in 2 of 11 periods. In comparison with DRP, RTP yields significantly greater surplus in 7 of 11 periods and less in 4 of 11. A preference poll comparing DRP and RTP was conducted after each trial, and while 64% of the participants said they preferred DRP before RTP experiments, 76% selected the RTP structure afterwards, a statistically significant reversal of preferences. Thus the opinion poll seemed to reflect the observed differences in consumers' surplus.

Figure 4 illustrates the effects that these alternative demand-side and market-clearing schemes had on wholesale prices for the first group of buyers. The price pattern was similar for the second group over each of the same 22 market periods. In periods where there are significantly higher wholesale prices (optimally so, according to the theoretical calculations shown in the figure), they are highest under the fixed price treatment and lowest under DRP. However, customers may not be exposed to price spikes in these single-sided markets that are as high as they might be were active participants also representing suppliers. That analysis is left to the next section on two-sided markets where participants acting as generators might speculate and/or withhold capacity from the market. In this section with predetermined, cost-based offers, these price spikes simply reflect the varying marginal cost of meeting demand in different periods. Figure 4 shows that under RTP, the market-clearing wholesale prices are closest to the theoretical optimum (note that RTP prices are generally lower than for FP and DRP in low load periods). These experimental results are therefore consistent with customers' intuition: DRP is shown to be an effective way of curbing price spikes. The problem is it does it in an inefficient way, and once the participants in these experiments experienced RTP and reaped its benefits, they voluntarily switched their preferences and selected RTP as their preferred buying mechanism going forward.

Since there are many regulatory restrictions on the sellers' behavior in most electricity markets operating in the U.S. today (e.g. price caps, prohibitions on withholding of supplies except for maintenance and necessary repairs, and the automatic substitution of historic "reference" offers if a suppliers higher current offer is computed to have a significant impact on raising the current market price AMPs)), the actual market-clearing prices may be close to the suppliers' costs in most instances. Thus the analysis in this section may be representative of the consequences of introducing widespread customer participation in electricity market if all of the existing supply-side regulations are retained. What this analysis does not reflect is the effect that those regulations may have on the suppliers' incentives to reduce their costs or to invest in additional, more efficient generation capacity.

4. Two-Sided Experimental Structure

In the following experiments, these supply-side regulatory restrictions are in effect eliminated, and six active sellers are substituted for the pre-determined, random, costbased offers. Furthermore, there are no restrictions on the suppliers' offering behavior; they may offer as much or as little capacity as they want at whatever price they want in all periods. The only behavioral restriction is a prohibition on talking to each other and/or discussing their offers (no ant-trust violations).

4.1 Supply side representation

Each of the six active suppliers is assigned three different generating units with different constant incremental production costs (20 MW @ \$22/MW, 15 MW @ \$50/MW and 20 MW @ \$ 61/MW). In addition there is a fixed cost associated with each supplier's total capacity that must be paid regardless of the supplier's level of activity (\$20 per market period per generating unit, or \$60 per supplier). The supplier is free to offer as much or little capacity into the market, up to the total capacity limit on their generation, as they wish, and they can specify a different price for each of the three different blocks of power that they can offer into the market. Offers may be made at prices lower or higher than the incremental production cost. The discretionary cost each supplier can choose is associated with whether or not and how much capacity they offer into the market. Each MW offered bears an opportunity cost of \$5.00, regardless of having been selected to generate. This opportunity cost represents the commitment of resources and/or cost of foregone maintenance that is associated with planning to have those units available, as reflected in making an offer. The seller's problem is illustrated in Figure 5, and since the market in each period clears at the highest offer needed to meet the market demand, all suppliers with offered prices at or below that level are paid the identical last (highest) accepted offer. Each seller earns a profit in each period equal to the market price times the quantity they sell, minus the incremental cost of generating the electricity they sell, minus the \$5.00 opportunity cost times all of the energy they offer into the market, minus their fixed costs.

4.2 Market Structure and Calibrations

In these two-sided markets, 19 buyers and 7 sellers were included. However the seventh seller was represented by a computer-simulated agent with a single 30MW block of low-cost \$25/ MW generation (representing a base-load unit) that was always offered at cost, so the \$5/MW opportunity cost of making offers is already included in the \$25/MW. This unit was the only generator subject to random outages, and its behavior was simulated numerically so that none of the six active participants would feel that their earnings were biased by a random phenomenon. Each of the buyers was assigned a different set of valuations for the energy they could purchase. Those valuations were the same as for the single-sided experiments, and for approximately 80 percent of the buyers, those values were set very high but realistically, based upon previous empirical work (see Woo et. al. [9]). Therefore, the optimal quantity purchases would not change for similar market

conditions for this majority of buyers unless the market-clearing prices reached levels many multiples higher than those anticipated. Given the popular sentiment that "most" buyers are not interested in altering their electricity consumption, this assignment of values acknowledges that assertion. It also provided experimental flexibility when some anticipated subjects did not appear for assigned sessions; they were replaced by numerically simulated agents that were assigned valuations that were well above those anticipated to be at the decision-making margin. Thus, human subjects always played the role of the twenty percent of buyers with valuations that appeared at the margin in one or more periods. In fact the number of human buyers ranged from 13 to 17 out of a total of 19 in each of these two-sided experiments.

The same three demand-side treatments were tested as in the single-sided experiments, FP as the base-line, DRP and RTP. Each treatment was run over the identical eleven daynight pairs (22 periods, total) with the same sequence of combinations of normal periods, heat-waves and unit-outages, as listed in Table 1. Here, however DRP was triggered by any predicted retail price that exceeded \$.106/kWh (\$66/MW wholesale price) so that speculative behavior on the part of suppliers might also initiate this program. The average market demand in these experiments was designed to be approximately 200 MW (lower at night, higher during the day and in heat waves), and 330 MW of active supply was available, plus the 30 MW provided by the numerically-simulated base-load unit, when not subject to a random outage. The wholesale market was cleared at and all accepted suppliers were paid the uniform price of the highest (last) accepted offer. Demand was always met, despite withholding, because of the availability of purchases from external sources, which all participants were told about. What subjects weren't told ahead of time was when those sources would be used and at what price (thus external purchases were to represent economic market purchases from outside the system), but all participants were informed of the market-clearing wholesale price after each period. In fact whenever demand could not be met from internal supplies, or whenever the estimated wholesale price exceeded \$150/MW, those external purchases were invoked from the generator outside of the system whose cost was \$72/MW. Whenever that import generator wascalled upon, they set the wholesale market price at the lower of 1) \$150/MW, or 2) the last accepted internal offer plus an increment ranging between \$5 to \$15 that was selected randomly in each instance. The objective was to avoid having suppliers withhold capacity specifically in order to have the import generator set the wholesale price (in effect transforming a hidden price cap into a price floor).

4.3 Market Sequence

Each market period began with the auctioneer (ISO/RTO) providing fair load forecasts (quantities) for the upcoming two (day-night pair) periods. All buyers and sellers were told before each day-night pair whether the upcoming period had normal or heat-wave conditions, and whether or not a unit outage had occurred. Next the suppliers would submit their price-quantity offers for both of the day-night periods. Then, either price forecasts or firm prices and/or anticipated market conditions were given to the buyers. Under FP, the retail price was always set at \$.085/kWh, regardless of wholesale market conditions. Under the DRP treatment, the same fixed price of \$.085/kWh was charged for

all purchases, but when DRP was announced to be in effect, a \$.079/kWh credit for purchases below each buyer's announced benchmark consumption level was provided. These fixed prices and DRP credits differed from the amounts in the previous singlesided market experiments because of the fewer increments of cost assigned to suppliers in these two-sided experiments, but the range of demand valuations remained the same. Under the RTP treatment, a fair forecast of market clearing prices for the next day-night pair was announced, based upon market conditions and the suppliers' offers. The buyers then made their quantity purchases, suppliers were committed and the market clearing wholesale prices were declared. In the case of RTP, buyers were told the actual price they were assessed for their purchases in each of the previous day-night periods, which however didn't vary more than twenty percent from the forecast prices for those periods. Finally, each seller was told their earnings, and each buyer was apprised of the net value of their purchases, including DRP credits where applicable. The process was then repeated for the next day-night pair until all eleven pairs were completed.

Load forecasts were always based upon buyers' performing optimally at the fixed or forecast prices. The \$.085/kWh retail price was based upon an estimate of cost-based offers by suppliers and optimal purchases by buyers. The DRP credit reflected the saving in supply, at production cost, to the reacting customer plus a pro-rata share of the cost-based savings to the market. The price forecasts for the RTP treatment used the suppliers' actual offers and presumed the buyers would behave optimally.

Since retail prices and/or DRP credits were pre-specified and fixed under the FP and DRP treatments, there is no guarantee that the revenues collected from the buyers, minus the \$.04/kWh wires charge, would match the wholesale market obligations to the sellers. Therefore, after each of the first two treatments (FP and DRP), the change in retail price that would have been required to balance the ISO/RTO's budget was announced. In the case of RTP, no rate adjustment is required since buyers pay the actual market-clearing prices for their purchases.

4.4 Preference Polls

A poll was conducted after each of the three treatments in which the participants were asked which of two treatments they preferred: DRP or RTP. The poll was conducted and results tabulated before the subjects had any experience with either treatment, again after they completed the DRP treatment, and then again after they completed both DRP and RTP. The required adjustments in retail prices were also announced after the FP, and again after the DRP treatments, but before the respective preference polls were conducted. What differed about the final poll is that the participants were told that based upon a majority vote, they would play four additional day-night pairs using the treatment (DRP or RTP) they selected. Furthermore, in this final round they were told that their exchange rates (always < 1.0 to keep the cost of the experiments within the researchers' budget, but different for every participant so each had an equivalent chance to make the same money despite different costs and valuations) would be adjusted so that they might anticipate earning as much money for this final four period round as they had in the earlier sessions that covered eleven day-night pairs.

4.5 Selection of Subjects, Training and Compensation

Since a primary issue addressed in these experiments is the extent to which the introduction of active demand-side participation in these markets might reduce the exercise of market power by suppliers, it was essential to have subjects acting as generators who knew how to speculate and lift prices. In prior experiments advanced undergraduates and graduate students had demonstrated after sufficient experience that even six suppliers who were prohibited from exchanging information outside of the context of the market, and where only market-clearing information was provided, could nevertheless raise prices substantially above competitive levels.

Initial pilots were conducted with faculty and industrial sponsors who were experts in the electric industry, but the decision-making time was so long for these professionals (approximately 15 minutes per period) that a total requirement of two days was projected for running all three treatments over 22 periods. While each trial could have been restricted in duration to a shorter time, it was evident that in doing so, many of the subjects would have continued to learn how to perform more effectively as the experiments proceeded. This was particularly important for the suppliers' behavior, since as an example, in one pilot run with an abbreviated number of day-night pairs, it was evident that most of the generators were beginning to try to speculate only during the third, RTP, treatment. Therefore, unless the subjects were available for prolonged training, it appeared that comparative behavior between the three demand-side treatments would have been subject to severe order effects which could have been controlled for only by conducting many more experiments in permutated sequences, a costly proposition. Since the purpose of these experiments was not primarily to test scientifically for cognitive lags, the choice was made to use students with prior experience as subjects, and to give additional prior training to those who would represent sellers to be sure they understood how to lift prices before the experiments began!

Even after separate training sessions for prospective sellers, several trial runs were made on each market treatment before that treatment was begun, and all questions by buyers and sellers were answered and communicated to all subjects before the actual experiments began (all questions that arose during the experiments were also answered privately). Thus the entire experiment lasted several hours on each of three separate evenings: one session for training, one to run FP and DRP treatments and one for RTP plus the final four high payment rounds using the treatment selected by the subjects.

All participants were paid in proportion to their total earnings. In the first experiment conducted late in 2003, 17 active buyers and 6 sellers participated, and they earned an average of \$49.27 in their training session and \$66.15 in the two experimental sessions (\$91.47 was the highest; \$10.53 the lowest). Only one buyer did not complete all trials, but since their valuation of purchases was extremely high, a computer agent was substituted in the absent rounds. All 13 active buyers and 6 sellers who began the April 2004 identical experiment completed it. In all cases, spare extra subjects who were trained as sellers were paid to appear at each experiment, but they were never called upon

to participate. In the second experiment the average payment during the training round was lower, \$22.32, but the average payment during the two experimental sessions was \$62.09, nearly identical to the earlier payments, although the spread was smaller (\$74.09 was the highest; \$34.55 was the lowest). Because each buyer had different assigned valuations for their purchases, and to ensure that all participants, whether buyer or seller had an equal opportunity to leave the experiments with the same amount of money, different exchange rates were assigned to the nominal earnings of each participant.

5. Experimental Results for Two-Sided Markets

5.1 Overall Efficiency and Wholesale Prices

Consumers' surplus, producers' surplus and total market efficiencies are summarized in Table 2 for the DRP and RTP treatments as a percentage of the wholesale revenues under the FP treatment. These efficiency measures are provided separately for each experiment and in combination. As a benchmark, the socially optimal levels of available efficiency are also presented, and the combined data indicate that a 6.75 % overall gain in efficiency, compared to a FP system without regulatory controls on suppliers, is possible. However, that gain in overall net efficiency is comprised of a large gain in the buyers' surplus and an enormous decline in producers' surplus because of the ability subjects who were acting as suppliers exhibited in raising prices and earning large profits under the FP regime. Overall market efficiency improvements were also obtained, compared to the FP regime, with RTP, but with a smaller transfer of surplus from sellers to buyers. By comparison, DRP leads to an overall loss in market efficiency compared to FP, but consumers gain substantially at the cost of an even greater loss to producers. Although the percentage differences varied between the two separate experimental groups, the qualitative results were similar in both cases.

Wholesale price patterns are illustrated separately for each of the experimental trials in Figures 6 and 7 for each market period. Prices are displayed for the FP, DRP and RTP treatments, as well as the socially optimal marginal-cost-based price. Both figures demonstrate the ability of suppliers to generate price spikes under a FP retail regime. In general, wholesale prices are the highest under FP, followed by RTP, DRP and the socially optimal prices in descending order. There are exceptions however. Suppliers were able to generate a severe price spike under DRP in the November 2003 experiments and under RTP in the April 2004 experiments. In fact, this speculative behavior by suppliers may not have been in their self-interest under the RTP regime, and the DRP price spike came at night during normal weather conditions! What these graphs may reflect is lagged learning; in these experiments subjects representing suppliers who had learned how to speculate and to lift prices against a FP retail market were slow to learn about circumstances when it no longer paid under RTP. Nevertheless, despite the persistent speculative behavior by suppliers, overall efficiency improvements are reaped through the RTP treatment as shown in Table 2. What Figures 6 and 7 confirm is that all prices should be and are higher in heat-waves and during supply shortages, with the exceptions noted, including the socially optimal price, and DRP and RTP tend to follow

that optimal pattern. However in the April trials, the DRP prices were lower than the socially optimal level in several instances, indicating a real welfare loss in those periods.

5.2 Statistical Tests on Differences in Surplus and Quantities

Because this experiment was repeated only twice, statistical inferences may not be drawn about overall changes in welfare, but if each subject is viewed as an observation, then the distributions of surplus and of quantities transacted can be estimated for each treatment, and a t-test can be conducted on the differences in surplus and differences in quantities among all buyers and all sellers between treatments. These pair-wise comparisons for both buyers and sellers are summarized in Table 3 for surplus differences and in Table 4 for quantity differences. Furthermore, the individual buyer's consumers surplus needs to be adjusted under the FP and DRP regimes to reflect the effect of the rate changes that would have had to be implemented in order to balance the ISO/RTO's budget. In the case of FP that increase would have been \$.0155/ kWh and \$.0152/ kWh, respectively, in the two sets of experiments. Under DRP, that increase would have been an even larger \$.0205/ kWh in the November 2003 experiment, but a much smaller \$.0081/ kWh for the April 2004 group. The buyers' surplus was adjusted both on the basis of this per kWh charge and by an equal lump-sum allocation, as reported in the statistical summary in Table 3. Comparing buyers' surplus adjusted for the rate increase on a per kWh basis between FP and either DRP or RTP, customers are better off with active participation at the .95 level under RTP and nearly so under DRP. In this case the differences between DRP and RTP are not statistically significant. The conclusions regarding buyers are less concrete when consumers' surplus is adjusted on an equal per customer basis except DRP is still significantly better than FP at the .95 level. Sellers are significantly better off under FP as compared pair-wise with either DRP or RTP, but they should prefer RTP over DRP based upon the welfare effects on them.

Table 4 illustrates the substantive behavioral differences in quantities consumed by buyers under the three treatments. Buyers consume less electricity in all periods under DRP, as compared to FP; whereas, under RTP customers buy more electricity at night and less during the day than under FP. Furthermore, the last column emphasizes the overall conservation effect of DRP since it results in a statistically significant reduction in purchases both during the day and at night, as compared to RTP. Unfortunately, this is inefficient as highlighted by the quantity comparisons between DRP and RTP with the socially optimal level of consumption: under DRP too little electricity is purchased in all periods; whereas, consumption under RTP was not significantly different than the optimal levels, except during normal day periods when too little was purchased. Similar results are shown for the suppliers' quantities, since supply always equals demand, but the statistical tests are somewhat less significant for sellers because of their smaller number.

5.3 Participant Preferences

The results from the polls comparing preferences between DRP and RTP are summarized in Table 5. In both groups, there is a reversal of stated preferences between DRP and RTP

between the initial poll taken before either treatment was tried, and after experience was gained with both. The first group switched from 74% preferring DRP initially to 64% preferring RTP afterward, a statistically significant reversal. The second group's reversal was less appreciable and significant, moving from only 53% thinking they preferred DRP ahead of time, but a similar fraction to group one, 68%, stating they preferred RTP after having tried both. Furthermore, this last fraction reflected learned self-interest since the results of the poll were used to select the treatment that was used in the last four rounds with high-stakes payoff potential for the participants. In particular, note that all suppliers in both groups selected RTP as their preferred market-clearing mechanism after having tried both, a reflection of their self-interest as illustrated in Table 3.

5.4 Line Flow Predictability

As Robert Thomas has shown, under the former regulated regime with cost-based dispatch there is a systematic proportional relationship between power flow on any line in the system and overall system load; however, under market-based dispatch with single-sided markets and a pre-set demand, based upon FP retail pricing, virtually no correlation exists between system load and line-flows because of speculative behavior by suppliers. In a preliminary analysis of line flow implications from these experiments, the positive correlation appears to re-emerge under DRP and RTP. Figure 8 illustrates the PowerWeb 30 bus electrical transmission network that underlies these experiments. The location of all generators is shown, including the import generator that cleared the market when insufficient internal supplies were offered, and the buyers are distributed across the remaining busses.

The variation in power flows on each of the 39 transmission links in this network are plotted in Figure 9 for each of the three demand-side treatments examined in these experiments. Both the socially optimal line flows and an estimate of those flows that would have been observed under the former regulated regime (cost-based dispatch to meet the demand represented by the FP system - - the demand structure widely employed under the prior regulated regime), as the benchmark, are also added. Line 15 has the greatest variability under all regimes, since that is the location where the import generator feeds into the network when there are shortages, and that line is also linked to the generator that experiences random outages. In general, greater variability is associated with the market-based FP treatment, but those swings seem to be lower on most lines for DRP and RTP, approaching the levels of the former regulated regime.

Two of the lines were selected (line 15 with the greatest variability and the more typical line 30), and a statistical test was performed on the correlation between system load and line flows on those links, for all five cases illustrated in Figure 9. These regression results are summarized in Table 6. Because of the location of the import generator, there is actually a negative correlation between system load and the flow on line 15 (as system load increases, the probability of calling on imports increases which serves the load in the right-hand side of the system and reduces flow on that particular line), but that negative relationship exists under all five regimes. What is different is the magnitude and the degree of statistical significance of that relationship. The relationships are nearly identical

under the socially-optimal, previously regulated and RTP regimes; the association is weakest under the FP market case, but improves somewhat under DRP.

In the case of a more typical transmission link like line 30 where there is a positive relationship between system load and line flow in all five cases, once again the socially optimal and former regulated regimes yield almost identical results. Here, the relationship becomes much weaker under the FP market regime, becomes almost identical in magnitude, but not in statistical significance under DRP, and becomes even stronger under RTP, although still not as significant statistically. Thus operators of electrical systems may also find value in the widespread implementation of demand side participation if it strengthens the predictability of flows on any particular line.

6. Conclusions

These experimental results demonstrate the successful construction of a realistic demandside platform that can be used to test a variety of hypotheses about buyer and supplier behavior in two-sided electricity markets. These markets are not trivial, and substantial training was required to get subjects representing six sellers to lift prices well above competitive levels under the fixed, constant retail price regime that is used in most locations around the country. All markets were conducted without price caps, prohibitions on withholding supplies or automatic mitigation mechanisms employed by the ISO/RTO. Nevertheless, when pitted against these trained sellers, less sophisticated buyers with fairly simple demand-side mechanisms, representing pre-set demand response programs or real time pricing regimes, were able to mute much of the suppliers' exercise of market power without any regulatory interventions. Not only did real time pricing lead to the highest overall efficiency of these three market regimes, a majority of participants opted to use real time pricing going forward, including sellers, after having gained experience with that system (a reversal in preferences for DRP from beforehand). In fact the results of earlier experiments with active demand participation but cost-based supplies may be analogous to introducing widespread buyer participation into existing electricity markets that have many restrictions on the suppliers' offering behavior. In this case, RTP and DRP both improved consumers' surplus as compared to a FP market regime, but too much so in the case of DRP, and RTP again resulted in the greatest overall efficiency.

Finally, the predictability of electricity flows on several transmission lines was explored as a function of overall system load for these three two-sided market regimes and under a simulation of the former cost-based regulatory regime. That relationship deteriorates substantially under the FP market regime, is partly re-established under DRP, and under RTP once again resembles the predictability that was previously available to system operators under regulated power pool exchanges.

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Period Pairs	F	P-OPT	D	RP-OPT	R	IP-OPT	J	P-DRP	I	P-RIP	D	RP-RIP	Conditions
	sign	significance											
1&2	-	100.00%	-	100.00%	+	27.20%	-	63.49%	-	100.00%	-	100.00%	Ν
3&4	+	70.13%	+	96.65%	+	90.58%	-	96.30%	-	45.95%	+	96.65%	S, DRP
586	-	100.00%	-	99.30%	+	100.00%	+	67.09%	-	100.00%	-	99.30%	Н
788	-	100.00%	-	100.00%	+	100.00%	0	0.00%	-	100.00%	-	100.00%	Ν
9&10	-	100.00%	-	100.00%	+	100.00%	-	68.99%	-	100.00%	-	100.00%	Ν
11&12	-	100.00%	-	100.00%	+	100.00%	+	67.07%	-	100.00%	-	100.00%	Ν
13&14	+	100.00%	+	99.70%	+	100.00%	-	91.86%	+	99.99%	+	99.70%	H+S, DRP
15&16	+	100.00%	+	99.69%	+	54.05%	-	91.55%	+	99.99%	+	99.69%	H+S, DRP
17&18	-	100.00%	-	100.00%	+	100.00%	-	67.07%	-	100.00%	-	100.00%	Ν
198-20	+	86.89%	+	96.63%	+	99.95%	-	95.63%	+	29.62%	+	96.63%	S, DRP
21&22	-	100.00%	-	100.00%	+	100.00%	-	67.07%	-	100.00%	-	100.00%	Н

 Table 1. Single-Sided Market: Paired t-tests on Weighted Individual Consumer Surplus

 Differences, Active Demand-Side/Preset Cost-Based Supply.

(N=Normal, H=Heat Wave, S=Generator Outage, DRP=In Effect)

Table 2. Two-Sided Market Experiment Results: Differences in Consumer Surplus (CS) Adjusted for Budget Deficit, Producer Surplus (PS) and Total Surplus (TS) from Fixed Price Regime Levels as % of Wholesale Market Revenue

	Adjusted CS Difference	PS Difference	TS Difference
	from Fixed Price	from Fixed Price	from Fixed Price
	Experiment 1 (November, 2003)		
Demand Reduction Program (DRP)	8.97%	-12.71%	-3.73%
Real Time Pricing (RTP)	7.22%	-4.57%	2.65%
Socially Optimal (SO)	31.12%	-21.88%	9.25%
	Experiment 2 (April, 2004)		
Demand Reduction Program (DRP)	18.67%	-22.27%	-3.60%
Real Time Pricing (RTP)	10.79%	-9.38%	1.41%
Socially Optimal (SO)	27.55%	-23.25%	4.30%
	Combined Experiments		
Demand Reduction Program (DRP)	13.86%	-17.52%	-3.66%
Real Time Pricing (RTP)	9.02%	-6.99%	2.02%
Socially Optimal (SO)	29.32%	-22.57%	6.75%

Table 3. Two-Sided Markets: Statistical Analysis Using Paired t-tests of Differences in Surplus Between Treatments (Pooled Data)

A) Surplus A	Adjusted on per	Customer Basis				
· ·	SO - FP	SO - DRP	SO - RTP	FP - DRP	FP - RTP	DRP - RTP
P-Value	0.000	0.025	0.000	0.047	0.128	0.084
Sign	(+)	(+)	(+)	(-)	(-)	(+)
B) Surplus A	Adjusted on Qua	ntitity Purchase	ed Basis			
	SO - FP	SO - DRP	SO - RTP	FP - DRP	FP - RTP	DRP - RTP
P-Value	0.000	0.188	0.000	0.054	0.048	0.216
Sign	(+)	(+)	(+)	(-)	(•)	(+)
Seller Surp	lus (12 Sellers)					
	SO - FP	SO - DRP	SO - RTP	FP - DRP	FP - RTP	DRP - RTP
P-Value	0.000	0.086	0.049	0.000	0.003	0.003
Sign	(-)	(+)	(-)	(+)	(+)	(-)

 Table 4. Two-Sided Markets: Statistical Analysis Using Paired t-tests of Differences in Quantities Between Treatments (Pooled Data)

Buyer Quantities	(29 Buyers)						
	SO - FP	SO - DRP	SO - RTP	FP - DRP	FP - RTP	DRP - RTP	
N ormal Day	0.234	0.001 (+)	0.017	0.000	0.013	0.009	
N ormal N ight	0.036	0.001 (+)	0.641	0.017	0.039	0.000	
Heat Wave Day	0.008	0.000	0.267	0.000	0.004	0.000	
Heat Wave Night	(-) 0.165	(+) 0.046	(+) 0.665	(+) 0.180	(+) 0.160	(-) 0.043	
Combined Day	(+) 0.029	(+) 0.000	(+) 0.051	(+) 0.000	(-) 0.005	(-) 0.001	
Combined Night	(-) 0.016 (+)	(+) 0.002 (+)	(+) 0.535 (+)	(+) 0.048 (+)	(+) 0.033 (-)	(-) 0.001 (-)	
Seller Quantities (
	SO - FP	SO - DRP	SO - RTP	FP - DRP	FP - RTP	DRP - RTP	
N ormal Day	0.989	0.087	0.575	0.007	0.356	0.180	
N ormal N ight	0.799	0.462	0.984	0.401	0.555	0.237	
Heat Wave Day	0.281	0.025	0.782	0.001 (+)	0.021	0.100	
Heat Wave Night	0.726	0.352	0.992	0.635	0.663	0.525	
Combined Day	0.519	0.023 (+)	0.669	0.002	0.051	0.111	
Combined Night	0.768	0.384	0.987	0.436	0.350	0.221	
P-Values Assoc							

	Experiment 1 Experiment 2						
	Raw Vote (%)						
	DRP RTP DRP RTI						
1. <u>After FP</u>							
Buyers	17 (100)	0 (0)	7 (54)	6 (46)			
Sellers	0 (0)	6 (100)	3 (50)	3 (50)			
Combined	17 (74)	6 (26)	10 (53)	9 (47)			
2. After DRP							
Buyers	5 (29)	12 (71)	6 (46)	7 (54)			
Sellers	1 (17)	5 (83)	0 (0)	6 (100)			
Combined	6 (26)	17 (74)	6 (32)	13 (68)			
3. After RTP							
Buyers	8 (50)	8 (50)	6 (46)	7 (54)			
Sellers	0 (0)	6 (100)	0 (0)	6 (100)			
Combined	8 (36)	14 (64)	6 (32)	13 (68)			
Note: P-Value for Differences in Preferences between							
Stages 1 and 3 by Binomial Proportions Test							
Experiment 1: P = 0.0113							
$Experiment \ 2: \ P = 0.1890$							

Table 5. Two-Sided Markets: Participant Expression of Preferences (DRP vs. RTP) After Each Trial

Table 6. Two-Sided Markets: Statistical Relation Between Line Flowsand System Load

$ \begin{array}{ c c c c c c c } & (Reg. Regime) \\ Fixed Price with \\ Regulated \\ Optimum Sellers \\ Regression Results for Tie Line 15 \\ \hline Fixed Price \\ Program \\ Pricing \\ Program \\ Program \\ Pricing \\ Program \\ Prog$				Results with Active Participants				
Social Optimum Regulated Sellers Reduction Fixed Price Reduction Program Real Time Pricing Intercept 40.1779 39.1761 17.9780 29.9462 33.0568 Std Err 3.0375 2.1514 3.1385 3.8662 3.5013 Slope Coefficient (0.1982) (0.1901) (0.1025) (0.1789) (0.1909) Std Err 0.0167 0.0116 0.0168 0.0236 0.0197 R-Squared 0.7701 0.8657 0.4695 0.5777 0.6906 F-Statistic 140.6651 270.7614 37.1714 57.4517 93.7394 P-value 0.0000 0.0000 0.0000 0.0000 0.0000 Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0			(Reg. Regime)					
Optimum Sellers Fixed Price Program Pricing Regression Results for Tie Line 15 Intercept 40.1779 39.1761 17.9780 29.9462 33.0568 Std Err 3.0375 2.1514 3.1385 3.8662 3.5013 Slope Coefficient (0.1982) (0.1901) (0.1025) (0.1789) (0.1909) Std Err 0.0167 0.0116 0.0168 0.0236 0.0197 R-Squared 0.7701 0.8657 0.4695 0.5777 0.6906 F-Statistic 140.6651 270.7614 37.1714 57.4517 93.7394 P-value 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 Regression Results for Tie Line 30 Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024			Fixed Price with		Demand			
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Std Err 0.0167 0.0116 0.0168 0.0236 0.0197 R-Squared 0.7701 0.8657 0.4695 0.5777 0.6906 F-Statistic 140.6651 270.7614 37.1714 57.4517 93.7394 P-value 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 Regression Results for Tie Line 30 Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 Slope Coefficient 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0001 0.0000 0.0001 0.0000 Net The follo								
R-Squared 0.7701 0.8657 0.4695 0.5777 0.6906 F-Statistic 140.6651 270.7614 37.1714 57.4517 93.7394 P-value 0.0000 0.0000 0.0000 0.0000 0.0000 Regression Results for Tie Line 30 Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0001 0.00001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Power Store Stor	Slope Coefficient	(0.1982)	(0.1901)	(0.1025)	(0.1789)	(0.1909)		
F-Statistic 140.6651 270.7614 37.1714 57.4517 93.7394 P-value 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 Regression Results for Tie Line 30 Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Line Line Line Line Line Line Line Line <thline< th=""> <thline< th=""> <thline< td="" th<=""><td>Std Err</td><td>0.0167</td><td>0.0116</td><td>0.0168</td><td>0.0236</td><td>0.0197</td></thline<></thline<></thline<>	Std Err	0.0167	0.0116	0.0168	0.0236	0.0197		
F-Statistic 140.6651 270.7614 37.1714 57.4517 93.7394 P-value 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 Regression Results for Tie Line 30 Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Line Line Line Line Line Line Line Line <thline< th=""> <thline< th=""> <thline< td="" th<=""><td></td><td></td><td></td><td></td><td></td><td></td></thline<></thline<></thline<>								
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P-value 0.0000 0.0000 0.0000 0.0000 0.0000 Regression Results for Tie Line 30 Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Line Line Line								
Regression Results for Tie Line 30 Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line State Line	F-Statistic	140.6651	270.7614	37.1714	57.4517	93.7394		
Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Statistic 10.0001 0.0000	P-value	0.0000	0.0000	0.0000	0.0000	0.0000		
Intercept (17.5262) (18.5527) (9.1573) (13.9666) (17.5818) Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Statistic 10.0001 0.0000								
Std Err 1.5631 1.7259 2.4566 3.0202 3.1587 Slope Coefficient 0.0751 0.0753 0.0437 0.0802 0.1024 Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Line Line Line Line Line Line Line								
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Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Line Line Line Line Line Line Line	Std Err	1.5631	1.7259	2.4566	3.0202	3.1587		
Std Err 0.0086 0.0093 0.0132 0.0184 0.0178 R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load Line Line Line Line Line Line Line Line								
R-Squared 0.6449 0.6111 0.2079 0.3104 0.4409 F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load	Slope Coefficient	0.0751	0.0753	0.0437	0.0802	0.1024		
F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0000 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load	Std Err	0.0086	0.0093	0.0132	0.0184	0.0178		
F-Statistic 76.2617 66.0048 11.0260 18.9069 33.1193 P-value 0.0000 0.0000 0.0019 0.0000 0.0000 Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load								
P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS.	R-Squared	0.6449	0.6111	0.2079	0.3104	0.4409		
P-value 0.0000 0.0000 0.0019 0.0001 0.0000 Note: The following linear regression equation was estimated with OLS.								
Note: The following linear regression equation was estimated with OLS. Line Power Flow = Bo + B1 x System Load	F-Statistic	76.2617	66.0048	11.0260	18.9069	33.1193		
Line Power Flow = Bo + B1 x System Load	P-value	0.0000	0.0000	0.0019	0.0001	0.0000		
	Note: The following	g linear regressi	on equation was e	estimated with	OLS.			
N = 44 for all regressions	Line Power Flow =	Bo + B1 x Syst	em Load					
	N = 44 for all regres	sions						

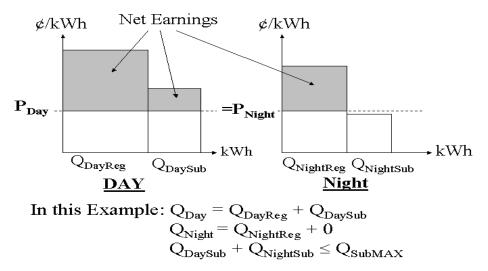
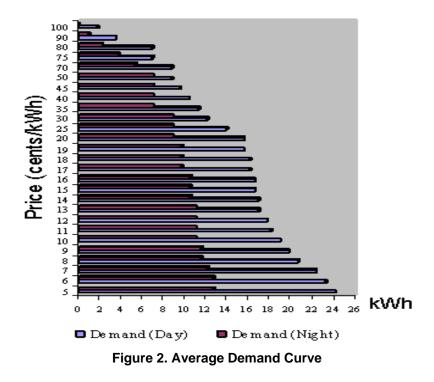


Figure 1. Buyer's Problem under a Fixed Price System



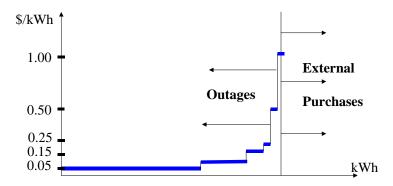


Figure 3. Typical "Hockey-Stick" Offer Curve

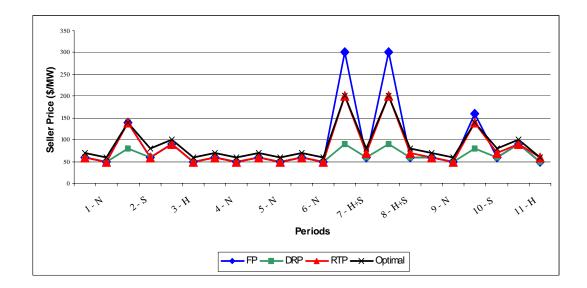


Figure 4. Experimental Prices: Active Demand/Preset Cost-Based Supply with Random Shift (Group 1)

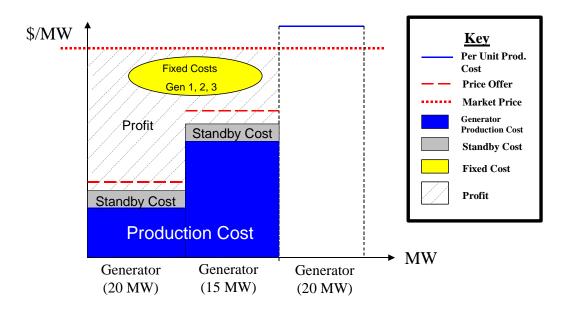


Figure 5. Illustration of Seller's Problem.

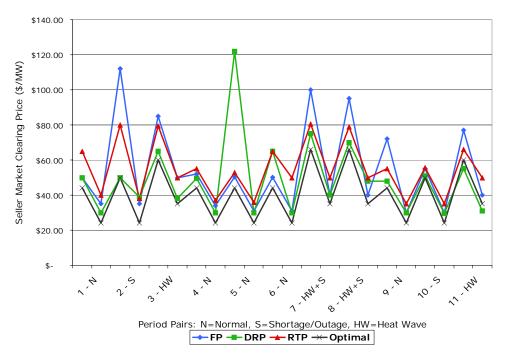


Figure 6. Prices by Treatment in Two-Sided Market Experiment (November, 2003)

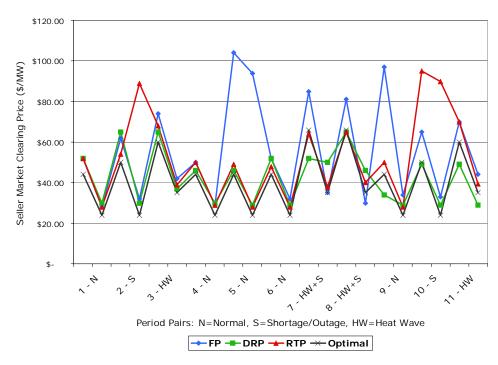


Figure 7. Prices by Treatment in Two-Sided Market Experiment (April, 2004)

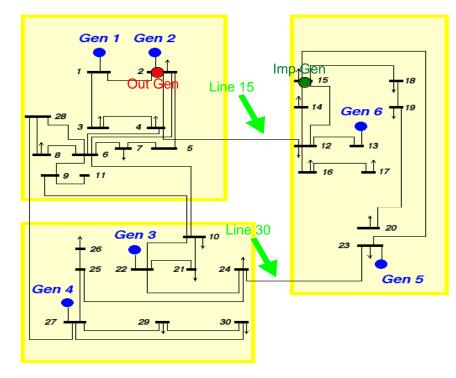


Figure 8. Power Web Simulated Electricity Network with Monitored Lines

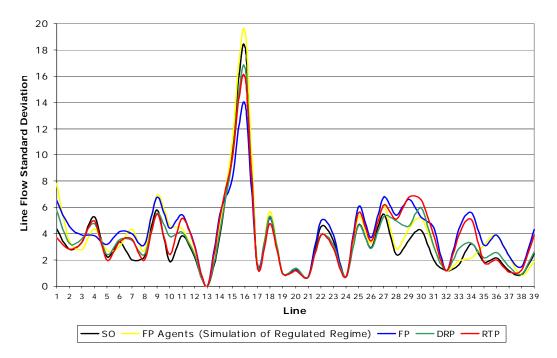


Figure 9. Two-Sided Markets: Line Flow Standard Deviation by Treatment Using Pooled Data From Experiments 1 and 2