

A Field Experiment on Dynamic Electricity Pricing in Los Alamos

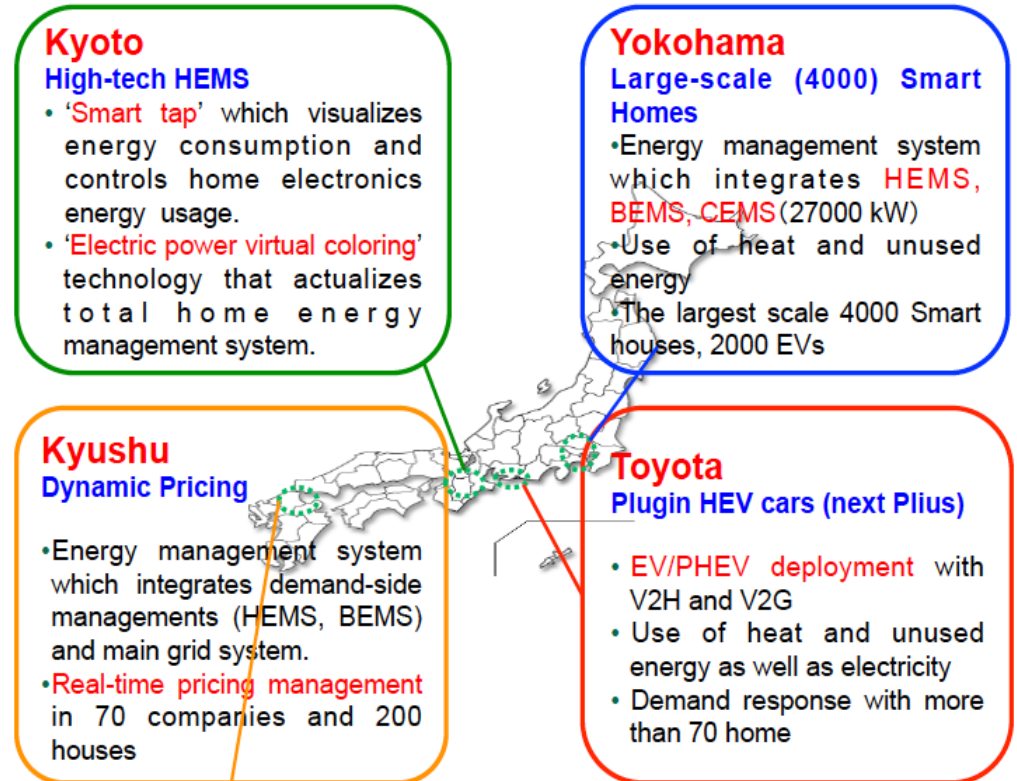
Opt-in Versus Opt-out

Takanori Ida and Wenjie Wang
Graduate School of Economics
Kyoto University
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Research on Smart Grid Economics in Ida Lab

Prof. Ida works as an economic advisor for **Japan's Ministry of Economy, Trade and Industry (METI)** to investigate the economic consequence of smart meter, energy management system and smart community.

He also manages METI's social experiment programs on smart grid economics, which have been conducted in four cities (2011-2014): Yokohama, Toyota, Kyoto and Kyushu.

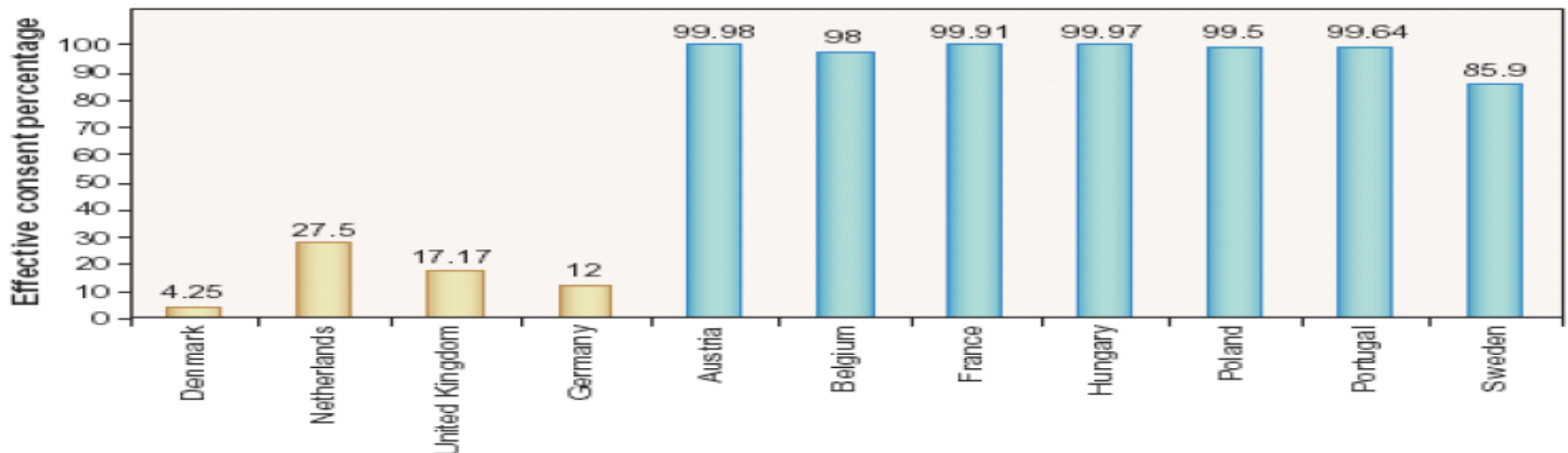


Demand Response Experiment in Los Alamos County

- The main objective of the Mesa Smart Meter Study is to evaluate **Peak-Cut Effects** under distinct combinations of dynamic electricity pricing programs (**Critical Peak Pricing/Peak Time Rebate**) and customer enrollment procedures (**Opt-in/Opt-out**).
- The CPP tariff can be considered as incentive framed as **“loss”**, while the PTR tariff can be considered as incentive framed as **“gain”**. Studies in Behavioral Economics point out that **“loss”** always looms much larger than **“gain”**.
- On-peak period is set to be **4pm – 7pm of weekdays**; Maximum 15 days of event days for summer and also 15 for winter.
- To ensure high reliability of results, we implement the experiment using **Randomized Control Trial (RCT)** with 3 treatment groups and 1 control group.

Decisions by Default: Opt-in Versus Opt-out

- The Los Alamos experiment is **1st in the world** to test the relative effectiveness of distinct combinations of default options (**opt-in versus opt-out**) and framing of economic incentive (**gain versus loss**).
- Decisions by defaults have recently become an important issue in many fields (insurance, finance, marketing, medicine, energy, etc.). Individuals often have a strong tendency to choose the default option (i.e., the option assigned to them).
- Example: **Organ Donation in Europe**, explicit-consent (opt-in) countries vs. presumed-consent (opt-out) countries.

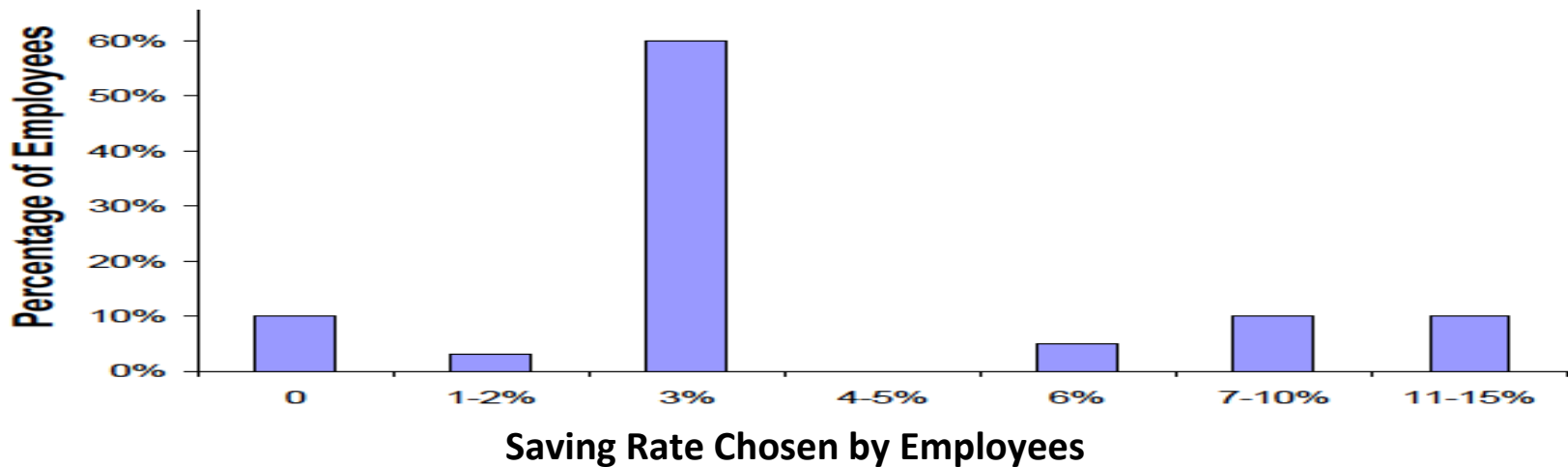


Decisions by Default: Opt-in Versus Opt-out

Famous example in U.S.: **401(k) retirement saving plan**. When employees must **opt-in** the plan, less than **50%** enrolled. When they are automatically enrolled, very few **opt-out**, resulting in nearly **100%** enrollment.

However, using opt-out procedure may generate **customer inertia**: employees stick with the default saving rate (3%) of the opt-out program, leading to dramatic decrease in the number of people who choose to save more than 3%.

The problem of “default effects” can become even more complicated when **incentive can be offered either as gain or as loss**.



Madrian and Shea (2001): "The power of suggestion: inertia in 401(k) participation and savings behavior", QUARTERLY JOURNAL OF ECONOMICS, 116(4), pp.1149-1187

Insights from the Los Alamos Experiment

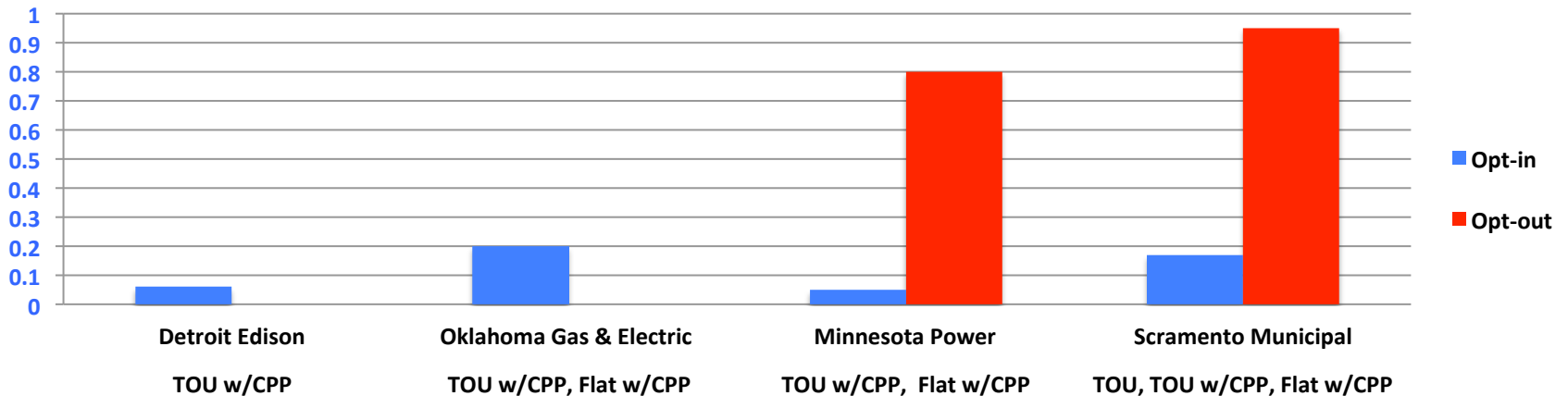
- **Very high customer acceptance rate** can be achieved even **under opt-in enrollment procedure**, which can be largely attributable to LAC customers' **high education levels and technology related experiences**.
- **Peak-cut Effects can be achieved** even in regions with rather **Mild Climates** (avg.max.temperature 77.2°F in summer).
- Average Peak-cut Effects in LAC :
 - **Opt-in CPP > Opt-out CPP > Opt-out PTR**
- i.e., Opt-in x “Loss” > Opt-out x “Loss” > Opt-out x “Gain”
- Customer Welfare & Bill Impact: saving of the **opt-in CPP group dominates** that of the opt-out CPP group.

Opt-in Versus Opt-out in Dynamic Electricity Pricing

Customer enrollment rates can be quite low for opt-in pricing programs, while those for opt-out programs are extremely high.

Example: enrollment rates of several demand-response pilots conducted in U.S.: Detroit Edison (opt-in), Oklahoma Gas & Electric (opt-in), Minnesota Power (opt-in & opt-out), and Sacramento Municipal (opt-in & opt-out).

Customer Enrollment Rates



Existing Concerns for Dynamic Pricing Programs

There are other issues/concerns in the demand-response literature:

- **Opt-in CPP:**
- Those who choose to “opt-in” may disproportionately be **“structural winners” or “free riders”**: customers who have low on-peak consumption, and thus save money under CPP rate even without changing behavior ; e.g., Borenstein (2013).
- **Low aggregate demand reduction impact**

- **Opt-out CPP:**
- However, it is not offered on a voluntary basis, thus can be **politically difficult** for utilities to implement in practice (less acceptable to customers/policy makers); e.g., Wolak (2010).
- **Less likely to be implemented in practice**

- **Opt-out PTR:**
- It can be seen as **Incentive framed as “gains”**, thus probably less effective than CPP (incentive frames as “loss”); e.g., Wolak (2010, 2011b).
- **“Loss Aversion” effect**: e.g., the incentive of PTR customers to save energy on very hot/cold days may be rather low, etc. We will discuss this issue further.

Customer Enrollment Rate of Opt-in CPP group

- Customer Enrollment Rates in the **Smart Pricing Option (SPO) Pilot of Sacramento Municipal Utility District:**

Opt-in CPP: 16% ~ 18% vs. **Opt-out CPP:** 93% ~ 98%

In their experiment, opt-out CPP group has generated much larger aggregate peak reduction. Indeed, opt-in CPP customers need to generate **6 times** the peak-cut effects of their opt-out counterparts to have similar aggregate impact.

- In contrast, Customer Enrollment Rate in the LAC experiment:

Opt-in CPP: 63% vs. **Opt-out CPP:** 97%

* **The conventional wisdom that “because of extremely high customer enrollment rate, peak reduction of opt-out CPP will be much larger than opt-in CPP” has been challenged by the Los Alamos experiment.**

- What factors may have led to low opt-in enrollment rate in existing studies?
- **Status quo bias, Inexperience to the related technology, Procrastination, etc.**

Customer Enrollment Rate of Opt-in CPP group

What factors may led to high Opt-in CPP enrollment rate in LAC?

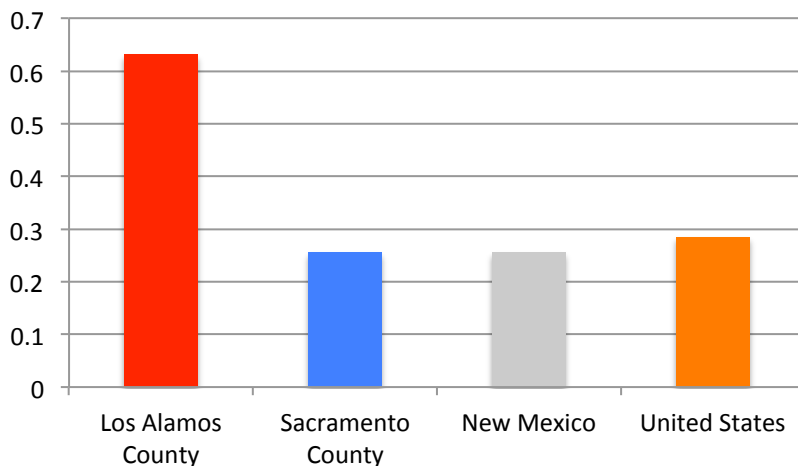
1. Bill protection

2. Education level/Experience: one of the highest education levels in U.S.; many hold Ph.D. degrees and/or work as scientists/engineers at LANL.

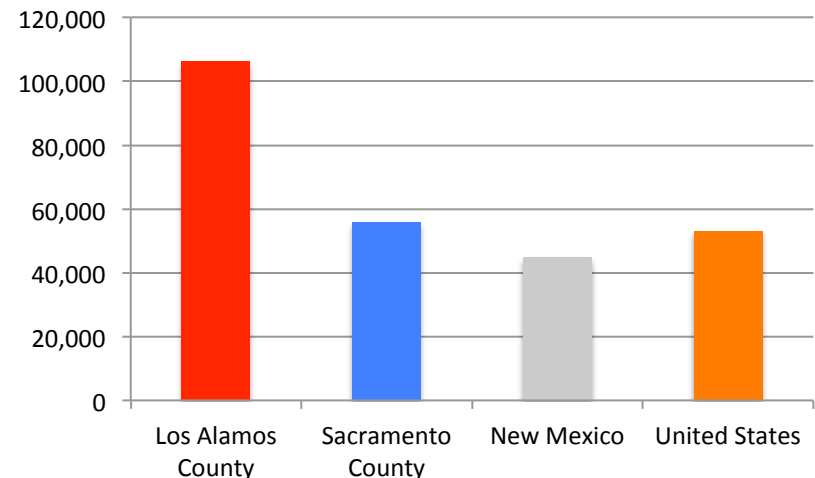
→ More open to/familiar with smart grid technologies/dynamic pricing programs;

→ More rational decision making.

Bachelor's degree or higher,
percent of persons age 25+
(2008-2012)



Median household income
(US\$, 2008-2012)



Peak-cut Impact in Regions with Mild Climates

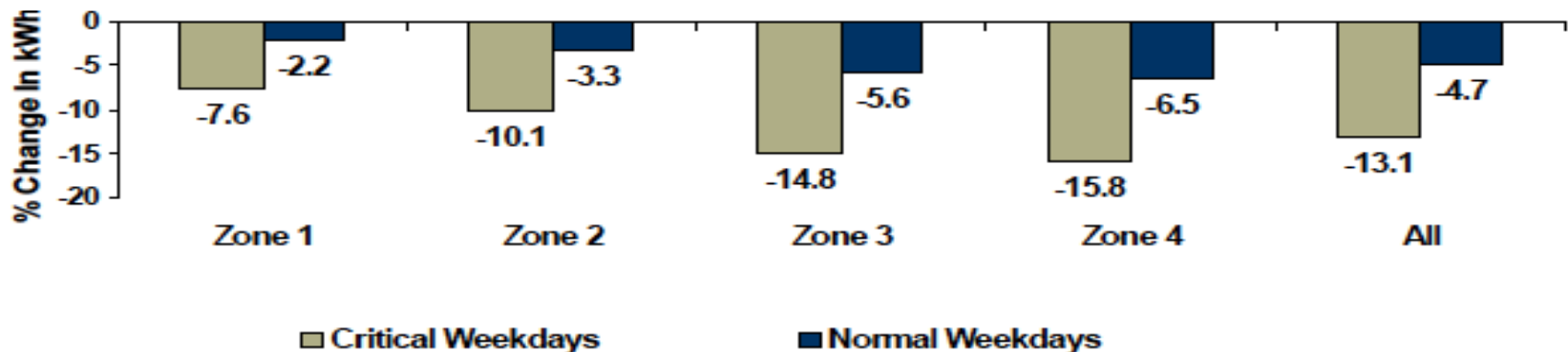
Some Characteristics of the **Los Alamos Experiment**:

1. **Mild climate**: 65.9°F (avg., Jul.-Sep.); 77.2°F (avg.max., Jul.-Sep.).
2. **Low Central Air Conditioner (CAC) saturation**: lower than 10% of households.

Almost all existing studies have been conducted in regions with hot climates. One exception is California's **Statewide Pricing Pilot (SPP)**.

Avg. Temperature (Jul.-Sep.): Zone 1: 73.5°F, Zone 2: 79.4°F, Zone 3: 88.6°F, Zone 4: 98.8°F; Statewide Avg. CAC saturation: 43%.

(Net) CPP Peak-cut Effects of Zone 1-Zone 4:



City Examples for Each Climate Zone:

Zone 1: Santa Cruz; Zone 2: Long Beach; Zone 3: Sacramento; Zone 4: Lancaster.

- "Impact Evaluation of the California Statewide Pricing Pilot, Final Report", Charles River Associates

Who participate into the experiment? & Who choose Opt-in CPP?

- Estimation results suggest that customers with **high average daily consumption** are more likely to participate in our experiment, and among customers assigned to the opt-in CPP group, those with **high on-peak usage** are more likely to choose CPP rate, very different from standard prediction.
- Therefore, in the Los Alamos experiment, not only **“structural winners”**, but also **“structural losers”** (i.e., customers with high on-peak consumption) also choose to opt-into CPP rate. This could be very important for the success of an “opt-in” based dynamic pricing program.
- **This suggest that it is desirable to use customer tools such as Bill Protection, at least in the beginning year of the dynamic pricing program.**

*Note: we use Binary Choice model here:

$$\bullet \quad \Pr(y_i = 1 \mid \mathbf{X}) = F(\mathbf{X}'\mathbf{b})$$

- Let $y=1$ if “I” participates in the experiment and $y=0$ otherwise. “X” denotes household characteristics, e.g., average daily consumption, on-peak/off-peak ratio. F is normal distribution function. Similar specification for the analysis on opt-in CPP customers.

Estimation Results (Summer 2013)

Treatment Groups	Gross Peak-cut (Intent-to-treat)	Net Peak-cut (Treatment-on-treated)
Opt-in CPP	-6.8%*** (0.016)	-10.4%*** (0.025)
Opt-out CPP	-4.5%** (0.020)	-4.7%** (0.020)
Opt-out PTR	-3.9%** (0.020)	-4.0%** (0.021)

Gross Peak-cut: the effects of those who are offered certain treatment; important for the analysis of the **aggregate load impact**.

Net Peak-cut: the effects of those who are not only offered but also actually accepted the treatment; important for the **analysis of customer behaviors**.

Test of Equality of Average Treatment Effects

Probability(Opt-in CPP = Opt-out PTR): 10%* for Gross., 1%*** for Net.

Probability(Opt-out CPP = Opt-out PTR): 77% for Gross., 78% for Net.

Probability(Opt-in CPP = Opt-out CPP): 20% for Gross., 1%*** for Net.

Estimation Results (Winter 2013)

Treatment Groups	Gross Peak-cut (Intent-to-treat)	Net Peak-cut (Treatment-on-treated)
Opt-in CPP	-4.6%*** (0.013)	-6.9%*** (0.019)
Opt-out CPP	-4.1%*** (0.015)	-4.3%*** (0.016)
Opt-out PTR	-3.0%** (0.015)	-3.1%** (0.015)

Test of Equality of Average Treatment Effects

Probability(Opt-in CPP = Opt-out PTR): 31% for Gross., 5%** for Net.

Probability(Opt-out CPP = Opt-out PTR): 52% for Gross., 52% for Net.

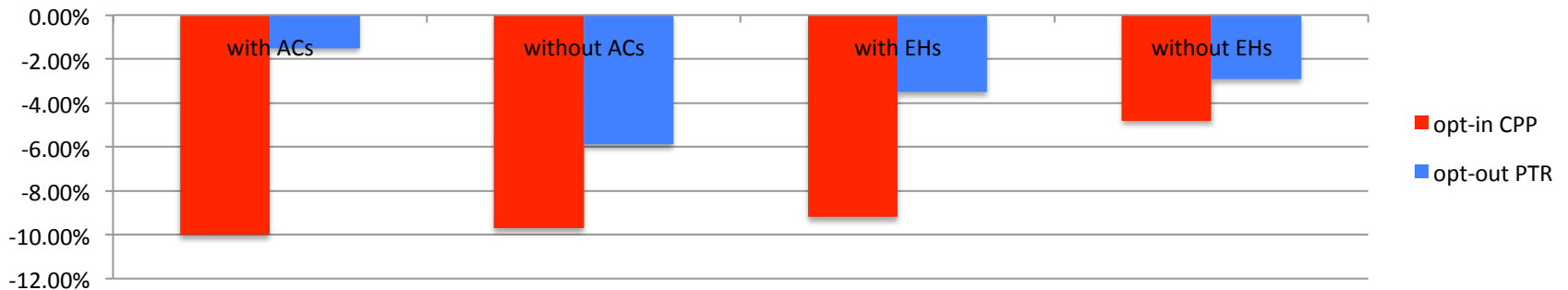
Probability(Opt-in CPP = Opt-out CPP): 77% for Gross., 19% for Net.

*Note 1: Similar to most results in the literature, the Peak-cut effects for winter are relatively lower compared with those for summer.

*Note 2: For the estimation of peak-cut effects (summer & winter), we implemented panel data regression controlling time/household fixed effects.

Subgroup Analysis with Customer Characteristics (Air Conditioners/Electric Heaters)

We separate sample into subgroups of customers with/without ACs for summer (with/without EHs for winter), and estimate (Net) Effects for each subgroup.



Remark: Opt-in CPP performs much better than opt-out PTR among households who have ACs/EHs. These results suggest there is **“loss aversion” effect** in opt-out PTR group.

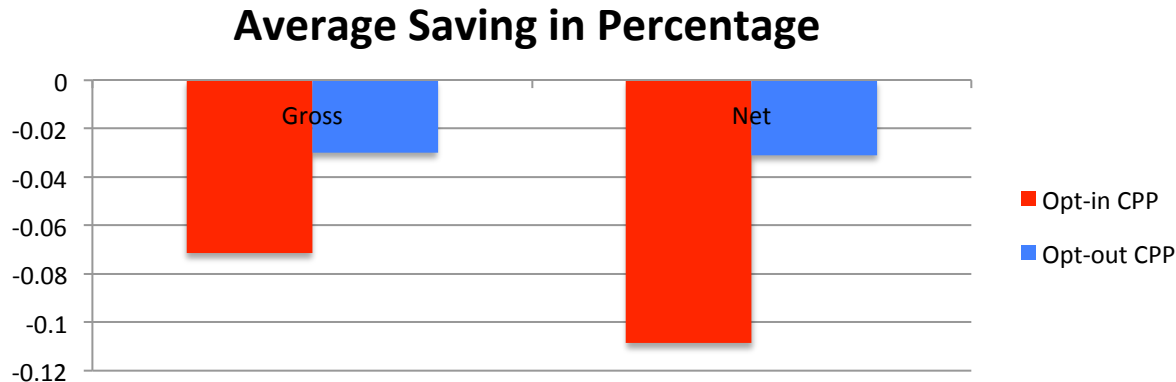
CPP: incentive framed as “Loss”; PTR: incentive framed as “Gain”.

Suppose that on certain CPP/PTR event day, the weather is very hot. CPP customers with ACs still have strong incentive to adjust ACs since they face **“punishment” of high price (\$0.75/kwh)**. But PTR customers do not have such strong incentive; they may choose to abandon the rebate and pay **the standard flat rate (\$0.077/kwh)**.

Bill Impact and Customer Welfare

Bill saving of **opt-in CPP group dominates opt-out CPP group:**

- 1) Compared with the electricity bills of the control group, opt-in CPP customers **save 4% more (Gross Effect)** than opt-out CPP customers, and opt-in CPP customers who actually took the CPP treatment **save 7% more (Net Effect)**.



- 2) We also observe **Polarization Effect**: customers at the tails of the distribution of electricity usage (i.e., left tail: lowest usage customers, right tail: highest usage customers) are saving more.
- * **Possible reasons**: low-usage customers may be **more motivational** and high usage customers may have **more available means** to reduce peak-time usage.

**THANK YOU
FOR
YOUR ATTENTION!!**