

Powering Demand: Solar Photovoltaic Subsidies in California

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Outline

- 1 Motivating the Project
- 2 Modeling Choices
- 3 Data & Preliminary Results

Current Topic

- 1 Motivating the Project
- 2 Modeling Choices
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Solar Industry

- Resurgence of interest in solar energy
- Federal and state dollars funding policies
- Emerging industry globally

Geographical Focus

- California
 - Most succesful state in the US
 - Responsible for more than 60% of solar installations
 - Data on installations
- 1998 - 2006
 - Three different subsidy regimes
 - Capacity based subsidies

Subsidies from 1998-Present

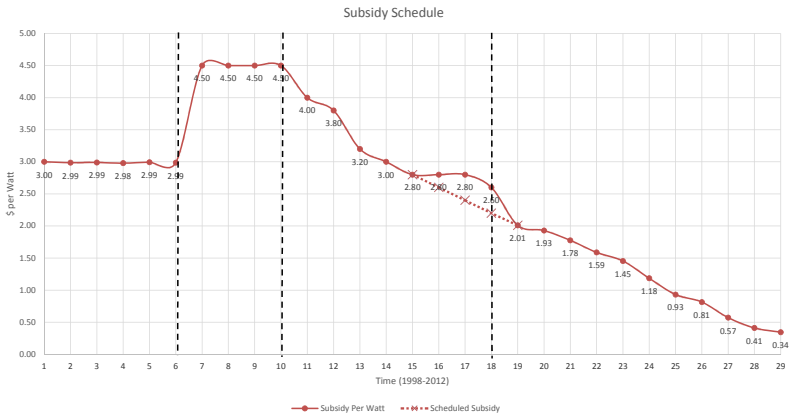


Figure : Subsidies Over Time

Consumer Choice

What product to purchase?

- 1 Capacity (kW)
- 2 Efficiency Rate
- 3 Area of the System (m^2)
- 4 Other characteristics

When to purchase?

- 1 Option value of waiting
- 2 Technology innovation, Prices, and Subsidies

Consumer Tradeoffs in the Market for Solar

Benefits:

- 1 Income Stream
- 2 Warm Glow
- 3 Signaling Green

Costs:

- 1 High up front cost, \$35,000
- 2 Foregoing future technology, subsidies, and prices

Research Question:

How do subsidies affect the consumer's purchasing decision?

Solar Policy

- Policy Counterfactuals (Lobel & Perakis 2011, Burr 2012)
 - ① Changes in subsidy levels affect adoption rates
 - ② Capacity Based vs Production Based Policies

Significance of the Research

- ① Improves characterization of the market
 - Including consumer and product level heterogeneity
 - Accounting for multiple levels of uncertainty
 - Introduction of newly assembled data set
- ② Enriches policy design and testing
 - Improves accuracy of policy predictions
 - Opens up room for new designs

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Households:

- $i = \{1, \dots, N\}$

Choice Set over Capacity:

- $s \in \{0, 1, 2, 3, 4\}$

Discrete Time Infinite Horizon

- $t = \{1, \dots, \infty\}$

State Space

- $\Omega_t = \{\omega_t, \epsilon_t\}$
- $\omega_t = \{P_t^e, P_t^{SP}, Z_t, \tau_t\}$

Consumer's Utility From Purchasing

$$V_i(\omega_t, \epsilon_t, s_{it}) = \underbrace{\theta_1 \delta_i^f(\omega_t, s_{it})}_{\text{NPV}} + \underbrace{\theta_{i2} z_{st}}_{\text{Solar PV Char}} - \underbrace{\alpha_i^{SP} (p_{st}^{SP} - \tau_{st})}_{\text{Price - subsidy}} + \epsilon_{ist}$$

where,

$$\delta_i^f(\omega_t, s_{it}) = E \left[\sum_{\tau=t}^{\infty} \beta^{\tau-t} \underbrace{\alpha^e \bar{p}_i^e q_{is\tau}^{SP}}_{\text{Solar Elec Gen}} \right]$$

Consumer's Utility From Not Purchasing

$$V_i(\omega_t, \epsilon_t, 0) = \epsilon_{i0t} + \beta \overbrace{\int_{\omega_{t+1}} EV_i(\omega_{t+1}, s_{it+1}) p(\omega_{t+1}|\omega_t) d\omega_{t+1}}^{\text{Expected Future Utility Staying in the market}}$$

where,

$$EV_i(\omega_{t+1}, s_{it+1}) = \ln \left\{ \sum_s e^{V_i(\omega_{t+1}, s_{it+1})} \right\}$$

Consumer's Maximization Problem

$$\max \left\{ \overbrace{V_i(\omega_t, \epsilon_t, 0)}^{\text{Value Of Waiting}}, \underbrace{\max_s V_i(\omega_t, \epsilon_t, s_{it})}_{\text{Utility maximizing product choice}} \right\}$$

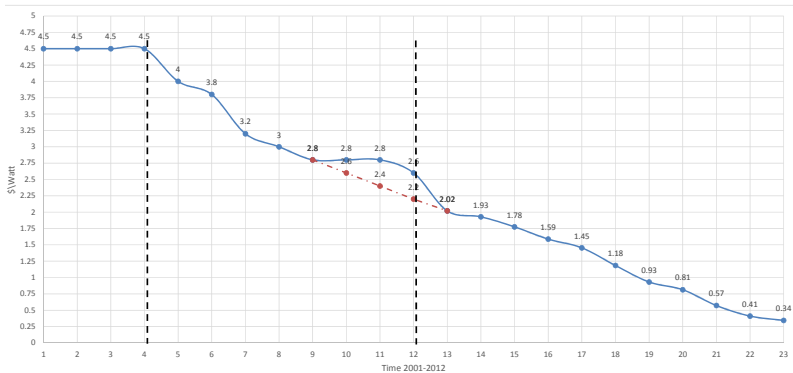
Supply Side

- Lack of data on the supply side of the market
- Assuming states evolve according to an AR(1) process

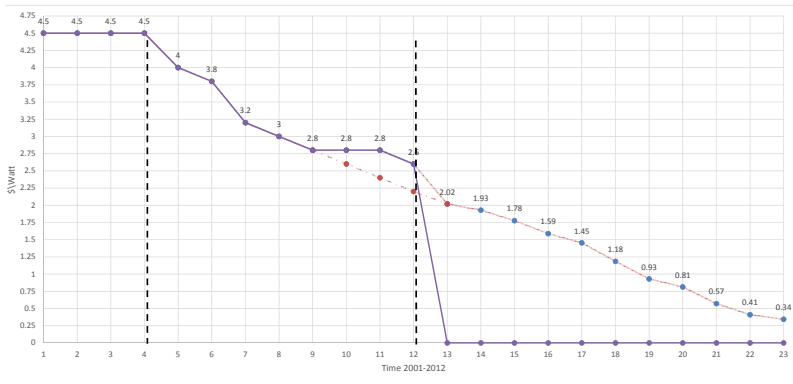
$$Z_{t+1}^{SP} = \lambda_1 + \lambda_2 Z_t^{SP} + v_{t+1}$$

- Assumes consumers have limited information about future states

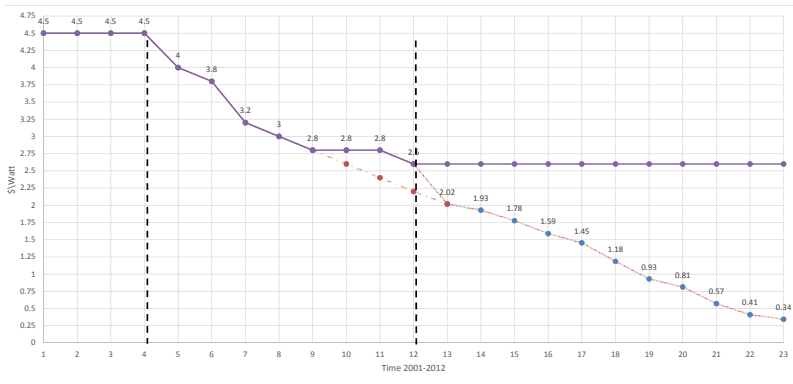
Perfect Foresight Case



Pessimistic Case



Optimistic Case



Assumptions and Restrictions:

- ① Electricity Consumption Data is not available
 - Assumption 1: Electricity is a homogeneous good
 - ① Utility from consumption of electricity is independent of the purchasing decision
 - Assumption 2: Price of electricity is an average price
- ② Restrictions include:
 - Do not model the change in consumption after purchase
 - Do not capture the effect of consumption levels on the decision to purchase

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California Energy Commission Program Data

- Tracks the purchase of all solar panel systems 1998-2006
- Three policy regimes over the time period
- Variables consist of
 - ① Physical Location
 - ② Total Price Paid
 - ③ Total Incentive Paid
 - ④ Capacity
 - ⑤ Brand/Model Product Identity
 - ⑥ Relevant Dates

Zillow.com: Housing Characteristics Data (Purchasers)

- Real estate website
- Gathered housing characteristics for consumers who purchased
- Variables include:
 - ① Housing Value
 - ② Square Footage
 - ③ Number of bedrooms/Bathrooms
 - ④ Number of Stories
 - ⑤ Lot Size
 - ⑥ Year Built

DataQuick: Housing Characteristics Data (Non-Purchasers)

- Purpose: To include non-purchasers in the model
- Purchased data that includes the following measures for each zip code:
 - 1 Count, Mean, Median, Standard Deviation
 - 2 Quintiles
 - 3 Correlation Tables
- For each of the following housing characteristics:
 - 1 Housing Value
 - 2 Square Footage
 - 3 Number of Bedrooms
 - 4 Number of Stories
 - 5 Year Built

Utility Companies: Electricity Pricing Data

- Electricity rates throughout the sample
- The major utility companies in CA

NOAA & NREL: Weather, Solar Data

- Data on weather throughout the sample
- Retrieved from the closest weather station
- Variables included:
 - ① Cooling degree days
 - ② Heating degree days
 - ③ Temperatures
 - ④ Solar Irradiation

Manufacturer Website: Solar Panel Product Characteristics

- Specification sheets for all panels purchased through 2011
- Variables include:
 - ① STC/PTC Capacity
 - ② Dimensions
 - ③ Warranty Information
 - ④ Efficiency Rates

California Energy Commission: Eligibility Dates

- Variables Include
 - ① Approval Date
 - ② Exclusion Date

Estimation Procedure:

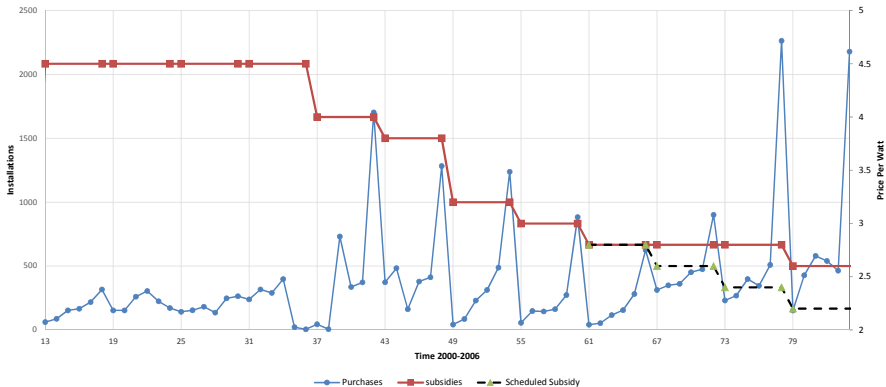
- Using the Nested Fixed Point Algorithm (NFXP) with full maximum likelihood estimation
- Steps for estimation:
 - 1 Estimate the parameters for the AR(1) processes that govern consumers beliefs over future states
 - 2 Given the estimates from step 1, run MLE until convergence:
 - 1 Calculate the expected value using the NFXP algorithm
 - 2 Calculate the log-likelihood function

	Case 1	Case 2
log(net price)	-6.523	-6.548
α_2		0.031
Area (m^2)	-0.131	-0.131
Efficiency	1.584	1.587
NPV	1.875	1.517

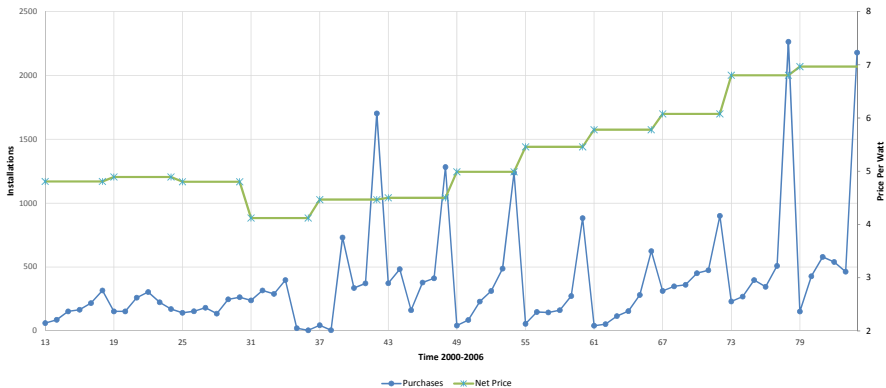
Next Steps

- Run the model that includes the nonpurchasers.
- Run counterfactual simulations to investigate alternative subsidies
 - Changing subsidy rates
 - Changing consumers information set about future subsidies
 - Targetting consumers

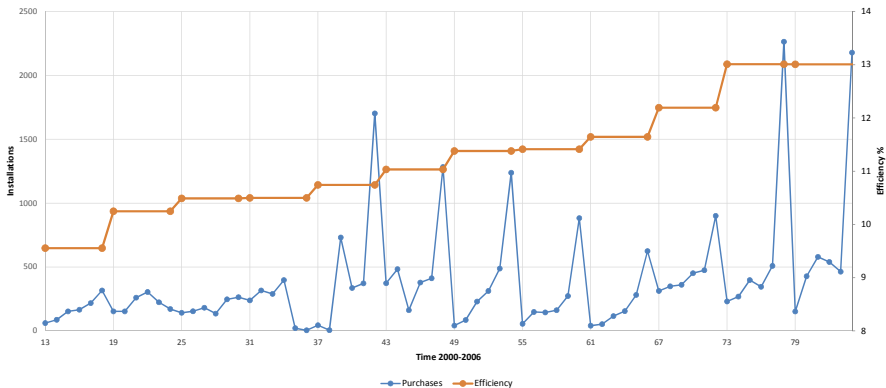
Installations vs Subsidies



Installations vs Net Price



Installations vs Efficiency



Efficiency vs Watts per m2

