

# Simple, Consistent Estimation of the Marginal Willingness to Pay Function: Recovering Rosen's Second Stage Without Instrumental Variables

Kelly C. Bishop

Olin Business School

Christopher Timmins

Duke University and NBER

Camp Resources XV, Wilmington, NC

August 7, 2008

# Motivation


- Property value hedonics allows us to recover the value of a marginal change in  $Z$ :

$$P_{j,k} = \beta_0 + \beta_1 Z_{j,k} + \beta_2 Z_{j,k}^2 + X'_{j,k} \alpha + \epsilon_{j,k}$$

$$\rho_{j^*(i),k} = \left. \frac{\partial P_{j,k}}{\partial Z_{j,k}} \right|_{j=j^*(i)}$$

- Need to recover MWTP function to value non-marginal changes.
- May want to consider preference heterogeneity.
- Potential for omitted variable bias.

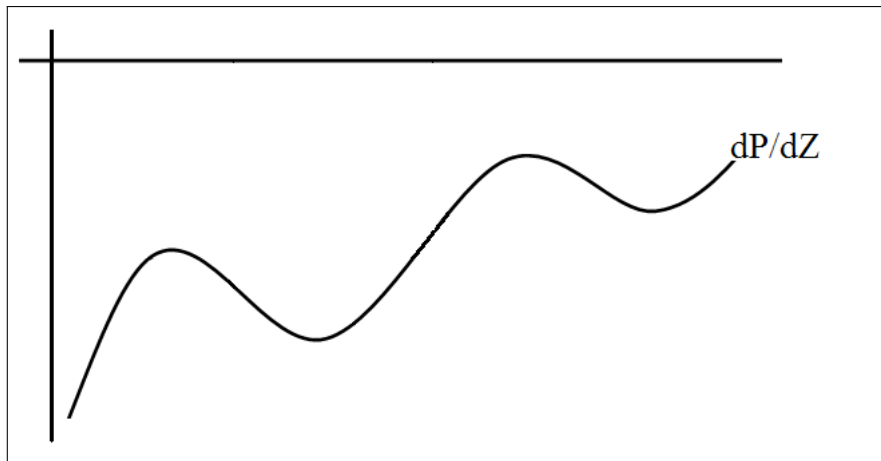
# Motivation - Rosen (1974)

- Second-stage regression of  $\rho_{j^*(i),k}$  on  $Z$  and individual attributes.
- Identification problem [Brown and Rosen (1982), Mendelsohn (1985)].
  - Use multi-market data
  - Estimate a non-parametric first-stage [Heckman, Ekeland, and Nesheim (2004)]
- Endogeneity problem [Epple (1987), Bartik (1987)]. 
  - Instrumental variables [Palmquist (1984), Chattopadhyay (1999), Boyle, Poor, and Taylor (1999)]

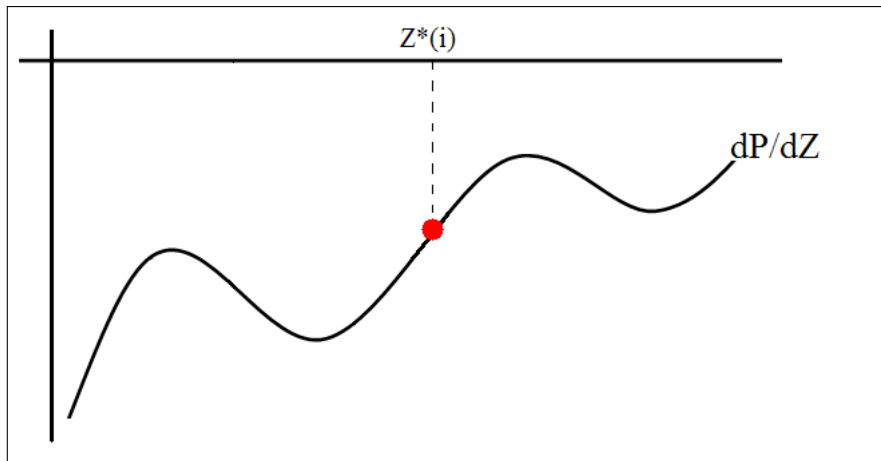
## Motivation - Bajari and Benkard (2005)

- Begins by estimating a non-parametric first-stage gradient.
- Second-stage preference inversion method that avoids endogeneity problems of Rosen (1974).
- Recovers individual-specific utility-function parameters from FOCs.
- Requires strong functional form assumptions on utility.
- MWTP is neither a function of  $Z$  nor a function of expenditure.

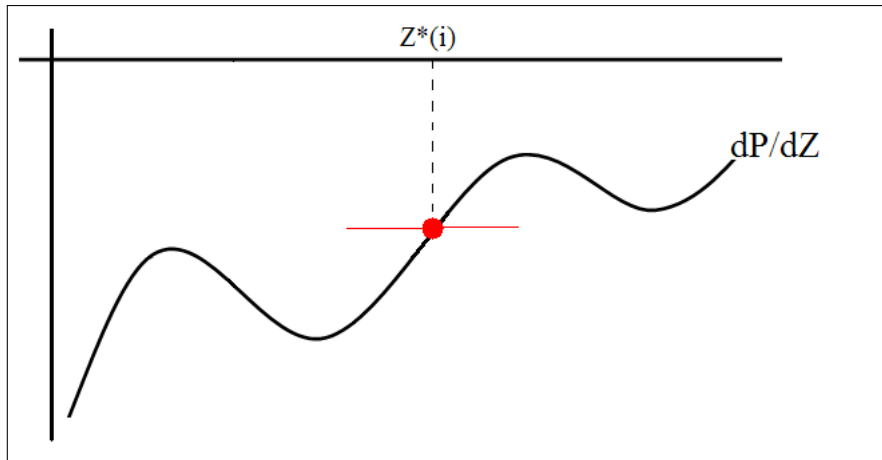
# Motivation - Bajari and Benkard (2005)



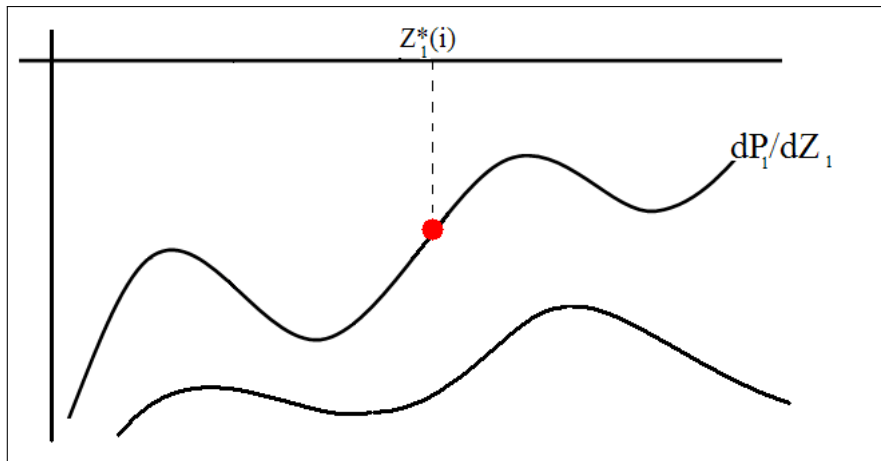
## Motivation - Bajari and Benkard (2005)



# Motivation - Bajari and Benkard (2005)

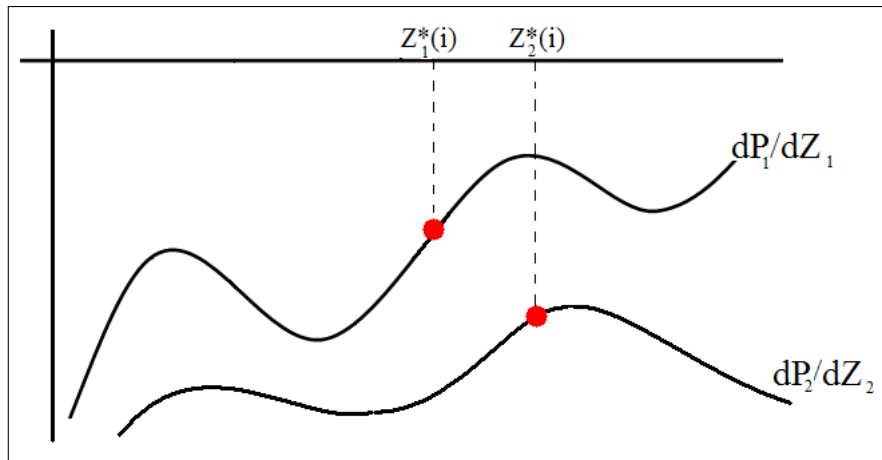


# Motivation - Bajari and Benkard (2005) with Panel Data



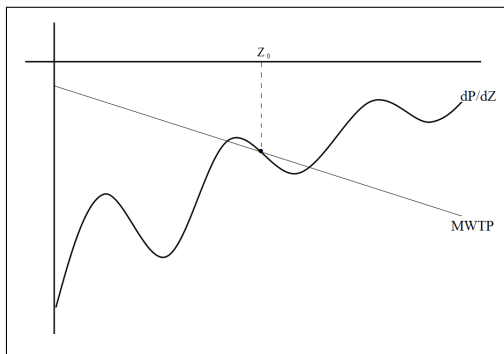


# Motivation - Bajari and Benkard (2005) with Panel Data



# Motivation - Bajari and Benkard (2005) with Panel Data

- Requires strict assumptions on stability of preferences over time
- HUGE data requirements
- Utility minimization (!)



## Property Transactions Panel (Dataquick)

- Bay Area of California 1990 - 2004
- Unique property ID along with housing characteristics
- Buyer name and Seller name

# Data

	Full Sample <i>properties = 594,665</i>		Repeat-Sale Sample <i>properties = 191,210</i>	
variable	mean	median	mean	median
Price (2000 \$)	377,684	330,000	377,158	336,673
Sq. Ft. House	1,696.95	1,500.00	1,589.71	1,432.00
Sq. Ft. Lot	9,611.62	5,500.00	8,384.23	5,044.00
Year Built	1966.48	1970	1968.60	1972
Total Rooms	6.38	6	6.21	6
Num. Bedrooms	2.94	3	2.86	3
Num. Bath	2.02	2	1.99	2
Days $\geq$ 90 PPB Ozone	2.44	2.09	2.54	2.09

# Data

## Individual Characteristics

- HMDA data
- Gender, race, income, down payment
- Transaction date, Census tract, lender name, loan amount

# Data

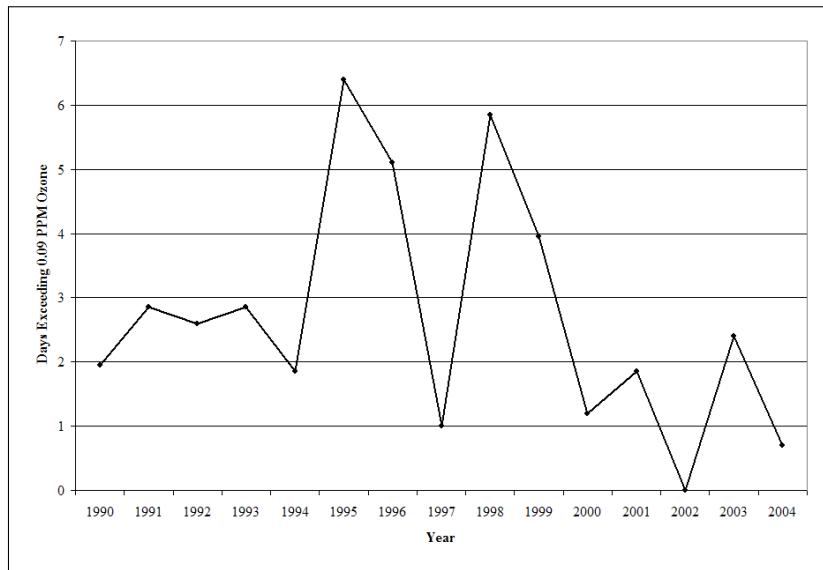
	Full Sample <i>ind = 40,092</i>		2-Purchase <i>ind = 9,805</i>		3-Purchase <i>ind = 1,076</i>	
variable	mean	median	mean	median	mean	median
Asian	0.21	0	0.22	0	0.21	0
Black	0.02	0	0.02	0	0.02	0
Hispanic	0.09	0	0.10	0	0.10	0
White	0.59	1	0.58	1	0.58	1
Other	0.09	0	0.08	0	0.09	0
Income	123,033	98,993	124,805	101,324	131,637	101,826
Price	409.943	344.778	431,786	365,997	437,937	354,653

# Data

## Ozone Data - CA Air Resources Board

- 37 monitors - days exceeding 90 PPB, monitor coverage
- Geographic coordinates
- Calculate a weighted average for each house ( $\frac{1}{dist^2}$ )

# Data





# First Stage - Estimation of the Hedonic Gradient

Estimate a non-parametric hedonic gradient that controls for house fixed-effects:

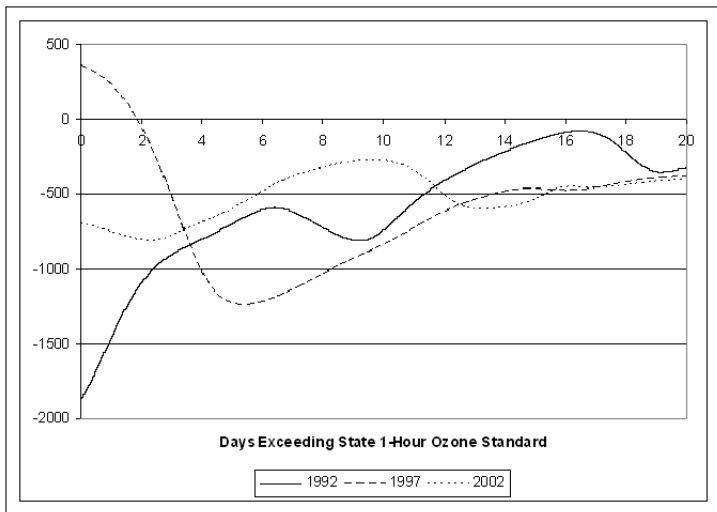
$$P_{j,t} = f(Z_{j,t}) + \xi_j + \nu_{j,t}$$

where  $f(\cdot)$  is an unspecified, flexible function of  $Z_{j,t}$ .

## Estimation Strategy

- Mean-difference over observations of property  $j$  to eliminate  $\xi_j$
- Local linear estimation  $\rightarrow$   
OLS at every observed level of  $Z_{j,t}$  and smooth results

# First Stage - Estimation of the Hedonic Gradient



## Second Stage - Estimation of the MWTP Function

Begin with indirect utility:

$$V_{i,j,t} = \alpha_{0,i} + \alpha_{1,i,t}X_j + \alpha_{2,i}X_j^2 + \alpha_{3,i,t}Z_{j,t} + \alpha_{4,i,t}Z_{j,t}^2 \\ + \alpha_{5,i}Z_{j,t}(I_{i,t} - R_t(X_j, Z_{j,t})) + (I_{i,t} - R_t(X_j, Z_{j,t}))$$

Specify preference heterogeneity as:

$$\alpha_{3,i,t} = \alpha_{3,0} + \alpha_{3,1}A_i + \epsilon_{i,t}$$

$$\alpha_{4,i,t} = \alpha_{4,0} + \alpha_{4,1}A_i + u_{i,t}$$

$$\alpha_{5,i} = \alpha_{5,0} + \alpha_{5,1}A_i$$

## Second Stage - Estimation of the MWTP Function

First Order Condition:

$$\epsilon_{i,t} = \frac{\partial R_t}{\partial Z_{j,t}} (1 + \alpha_{5,i} Z_{j,t}) - (\alpha_{3,0} + \alpha_{3,1} A_i + 2(\alpha_{4,0} + \alpha_{4,1} A_i + u_{i,t}) Z_{j,t} + \alpha_{5,i} (I_{i,t} - R_t(X_j, Z_{j,t})))$$

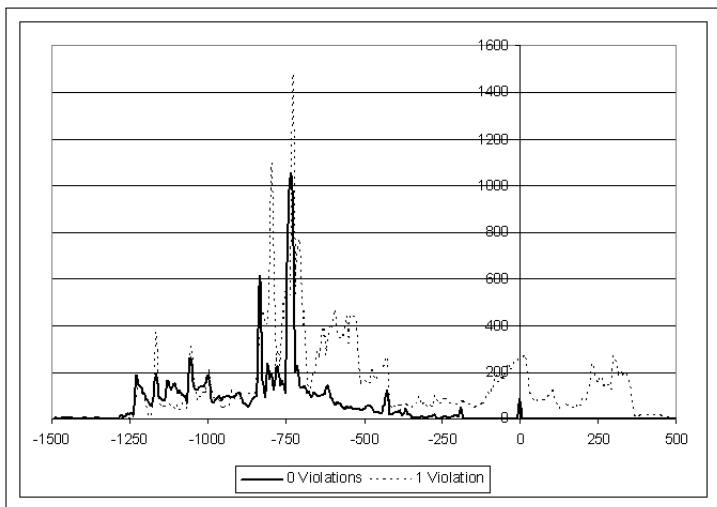
Second Order Condition:

$$u_{i,t} < \alpha_{5,i} \frac{\partial R_t}{\partial Z_{j,t}} - \alpha_{4,0} - \alpha_{4,1} A_i + \frac{1}{2} \frac{\partial^2 R_t}{\partial Z_{j,t}^2} (1 + \alpha_{5,i} Z_{j,t}) = \Gamma_{i,t}$$

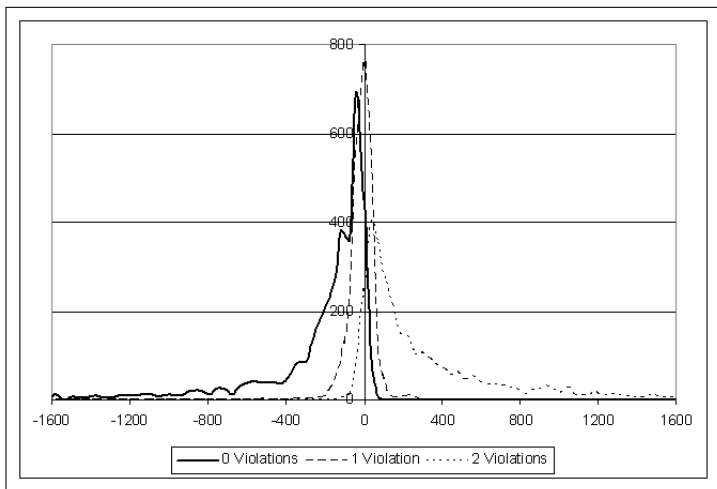
### Estimation Strategy

- Integrate out over feasible range of  $u_{i,t}$  conditional on  $Z_{j^*(i),t}$
- Maximize the likelihood of observing  $Z_{j^*(i),t}$

# Results - Cross Sectional Inversion



# Results - Panel Inversion

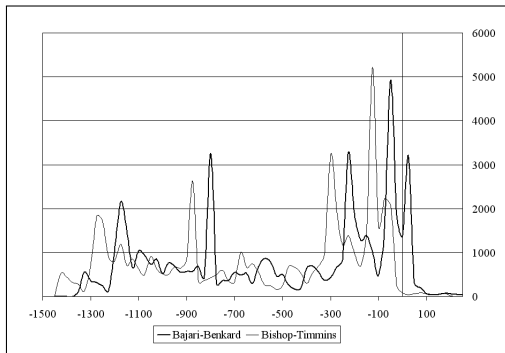


## Results - ML Estimator

parameter		value	s.e.	t-stat.
intercept	$\alpha_3$	200.96	3.45	58.20
ozone	$\alpha_4$	-46.96	0.81	-57.96
expenditure	$\alpha_5$	-0.30E-02	0.59E-04	-51.48
stdev( $\epsilon$ )	$\sigma_\epsilon$	466.13	5.39	86.44
stdev( $u$ )	$\sigma_u$	36.00	0.75	47.73

## Consider a One Day Increase in Ozone...

- Bajari and Benkard yields smaller damage estimate (17% less)
- Average MWTP masks considerable heterogeneity (\$860.94 at median house)

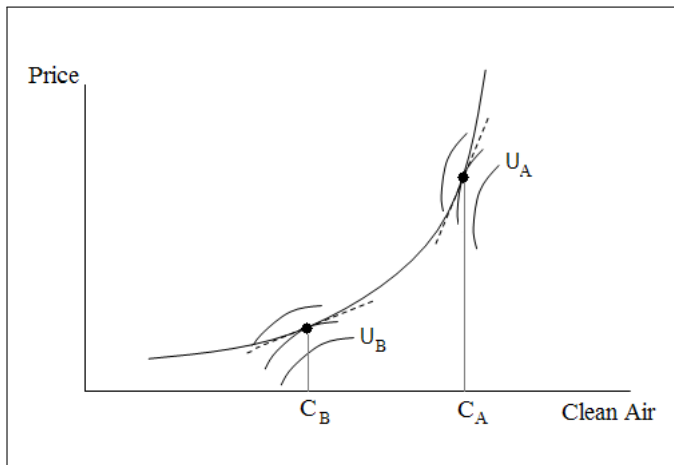




# Summary

- Estimate heterogeneous preferences for clear air.
- Two-stage approach that combines the insights of Rosen and Bajari-Benkard.
- Avoid endogeneity bias of Rosen.
- Doesn't require strict functional form assumptions as Bajari and Benkard.
- Doesn't require strict assumption of time-invariant preferences.
- Simple to estimate.
- Estimate with cross-section of buyers.

# Motivation



# Motivation

