Hedonic pricing (1)

Applied Econometrics for Spatial Economics

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1. Introduction

- 1. Introduction
- 2. The MWTP
- 3. The value function
- 4. Demand functions
- 5. Summary

- Yesterday:
 - 1. Spatial econometrics
 - 2. Discrete choice
 - 3. Identification
- Today:
 - 4. Hedonic pricing
 - 5. Quantitative spatial economics



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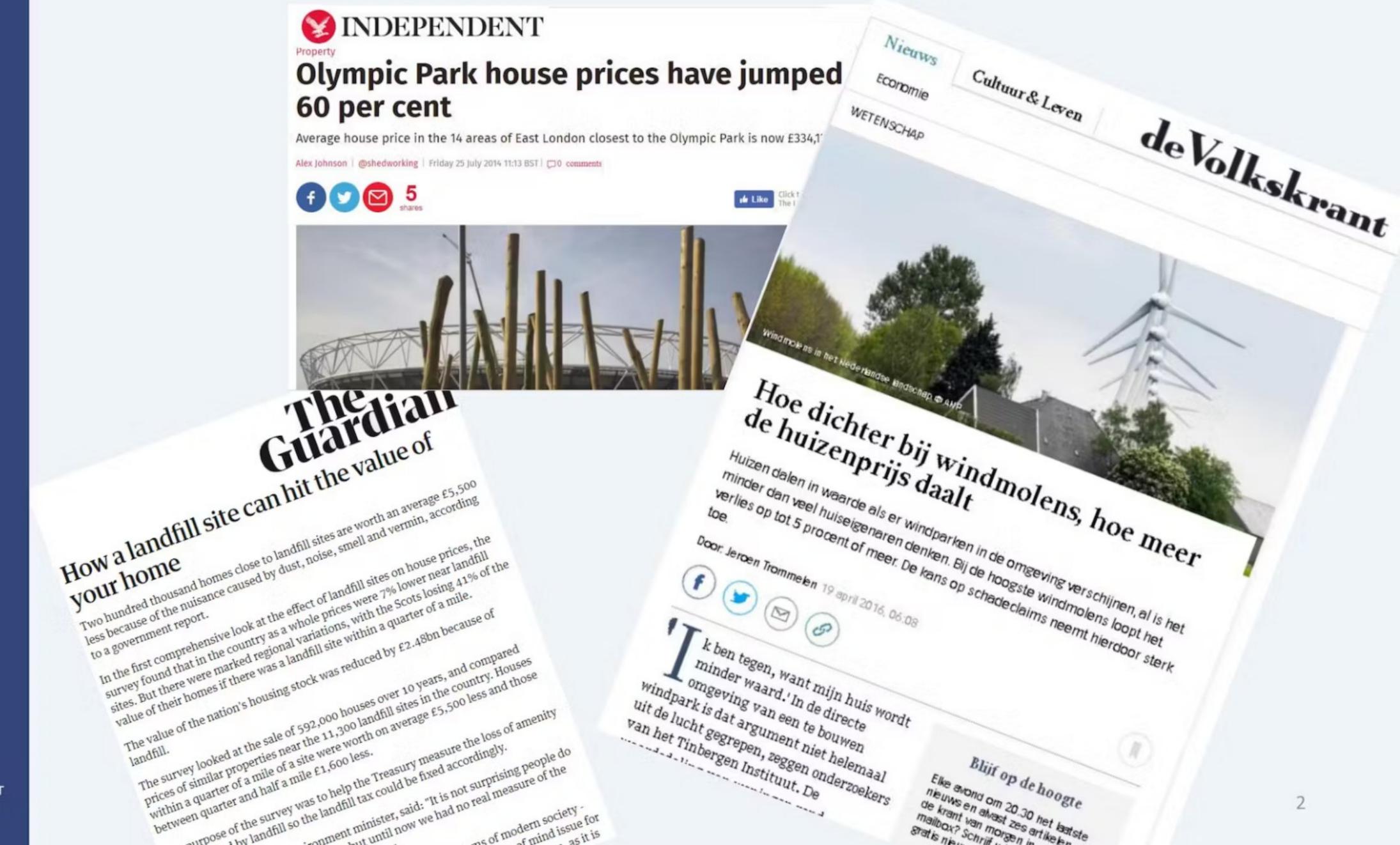
- Yesterday:
 - 1. Spatial econometrics
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 - 3. Identification
- Today:
 - 4. Hedonic pricing
 - Theory and estimation
 - 5. Quantitative spatial economics



Hedonic pricing (1)

1. Introduction

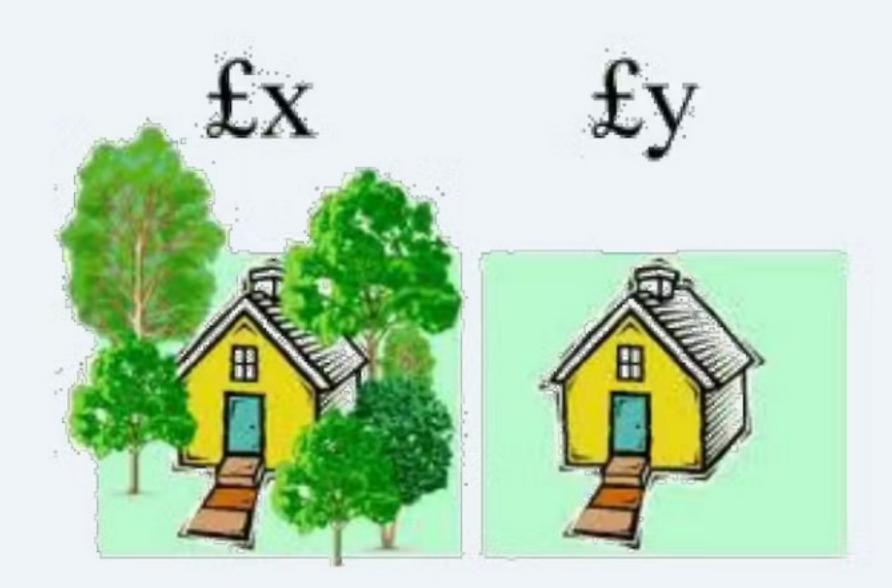
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- We focus on the housing market
- Hedonic price theory is often used to measure the price of public goods





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- Often used in applied research to answers questions like:
 - Is it beneficial to invest in a new park?
 - What are the social costs of a polluting power plant?
 - Are there any external effects of investments in poor neighbourhoods?
 - What is the effect of earthquakes?
 - •



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Goals of this class are:

- 1. Understand what a hedonic price function
- 2. Have basic knowledge about how a hedonic price function is linked to economic theory

Next lecture: more on estimation and price indices

Literature: <u>syllabus</u>



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A hedonic price function is:

 A description of the equilibrium prices of varieties of a heterogeneous good influenced by supply and demand

Is not expected to be stable over time

 Economic theory does not tell much about the shape of the hedonic price function



2. The marginal willingness to pay

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- Consider a heterogenous good
- Heterogeneity is described by a number of attributes k.
- Price is a function of these attributes, so p = p(k).
- If there are enough observations, we can estimate p(k).



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 Hedonic price function developed by Andrew Court (1939) in application to cars

- Extended by Zvi Griliches (1961)
 - Also applied to cars
 - Since then a standard econometric tool

 Rosen (1974) provided a link between the hedonic price theory and standard economic theory



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• How is the hedonic price function related to economic theory?

Let's assume a utility maximising household
 u = u(q, k)
 q denotes a composite good
 k denotes an attribute (e.g. house size)
 subject to a budget constraint y = q + p(k).

Then, it holds that:

$$\frac{\partial u/\partial k}{\partial u/\partial q} = \frac{\partial p}{\partial k}$$



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$$\frac{\partial u/\partial k}{\partial u/\partial q} = \frac{\partial p}{\partial k}$$

- It indicates the amount of money a consumer is willing to pay to get an additional unit of the attribute
 - An additional square meter of house size
 - Engine power for cars
 - while holding utility constant

Problem: households are not homogeneous in their utility functions!



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- To analyse the equilibrium we may use the concept of the 'value function'
 - We may write $u_1 = u_1^* = u(y_1 P_1, k)$.
 - We then invert the utility function with respect to $y_1 P_1$ to obtain $P_1 = y_1 u^{-1}(u_1^*, k)$

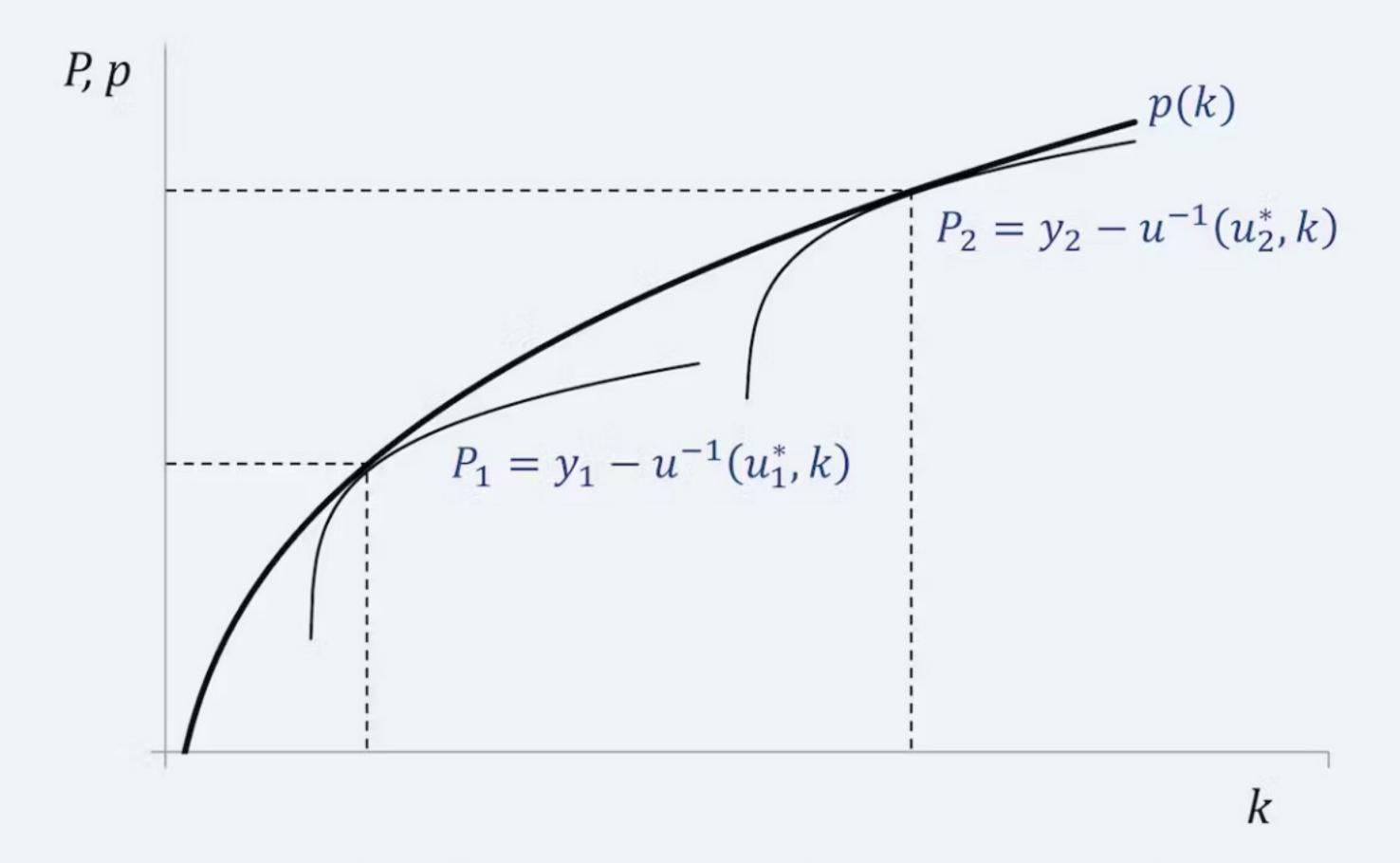
•
$$\frac{\partial P_1}{\partial y_1} = 1 \Rightarrow \Delta y$$
 translates completely into ΔWTP

•
$$\frac{\partial P_1}{\partial u_1^*} = -\frac{1}{\partial u/\partial k}$$
 \rightarrow WTP for good decreases in \bar{u}



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With heterogenous households, we have:



- Note that the *value functions* give the WTP, (P_1, P_2) for different consumers
 - Hedonic pricing function p(k) gives market price



4. Demand functions

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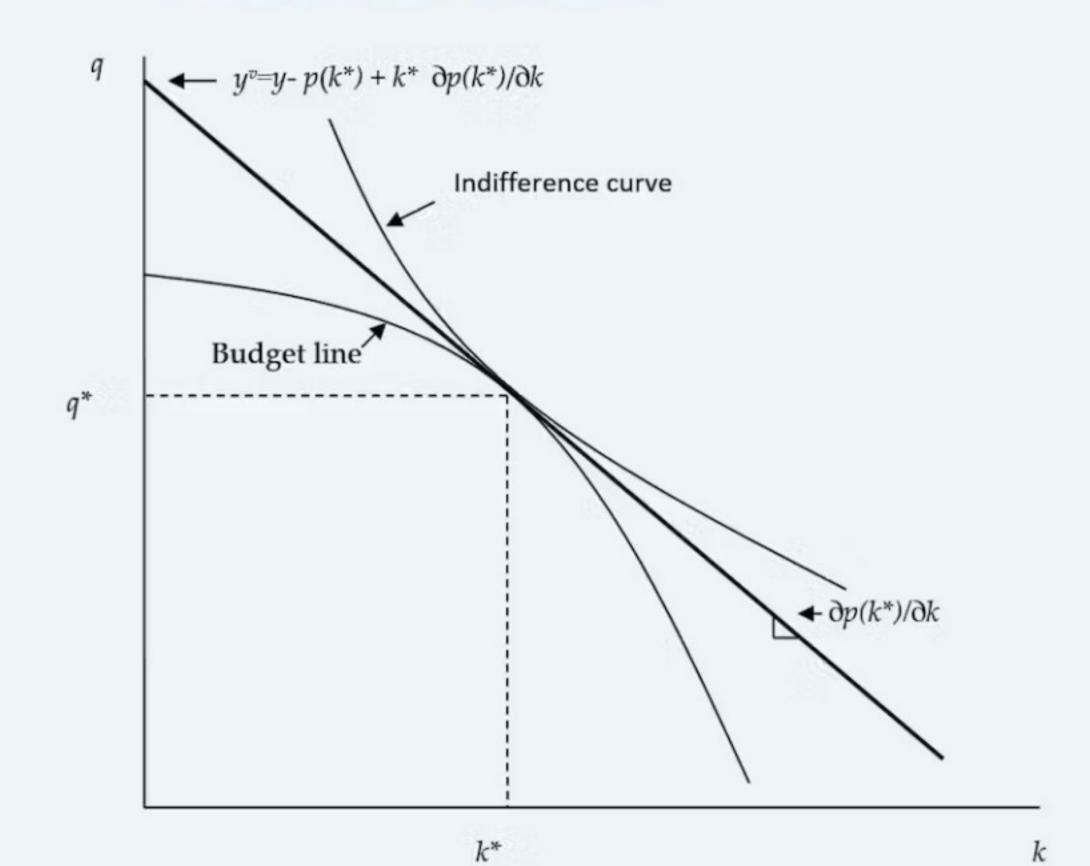
- Recall that $\frac{\partial u/\partial k}{\partial u/\partial q} = \frac{\partial p}{\partial k}$
- So what can we learn from the data if we estimate a hedonic price function?
 - We aim to obtain $\partial p/\partial k = \alpha$

- Can we determine the *demand* for an attribute k for an individual if we obtain α ?
 - e.g. demand for open space
 - Demand function: $k = f(\alpha)$
 - Inverse demand function: $\alpha = f(k)$



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- The budget constraint: y = q + p(k)
- A problem: there is no constant unit price of quality
 - The price (or willingness to pay) depends on the amount consumed





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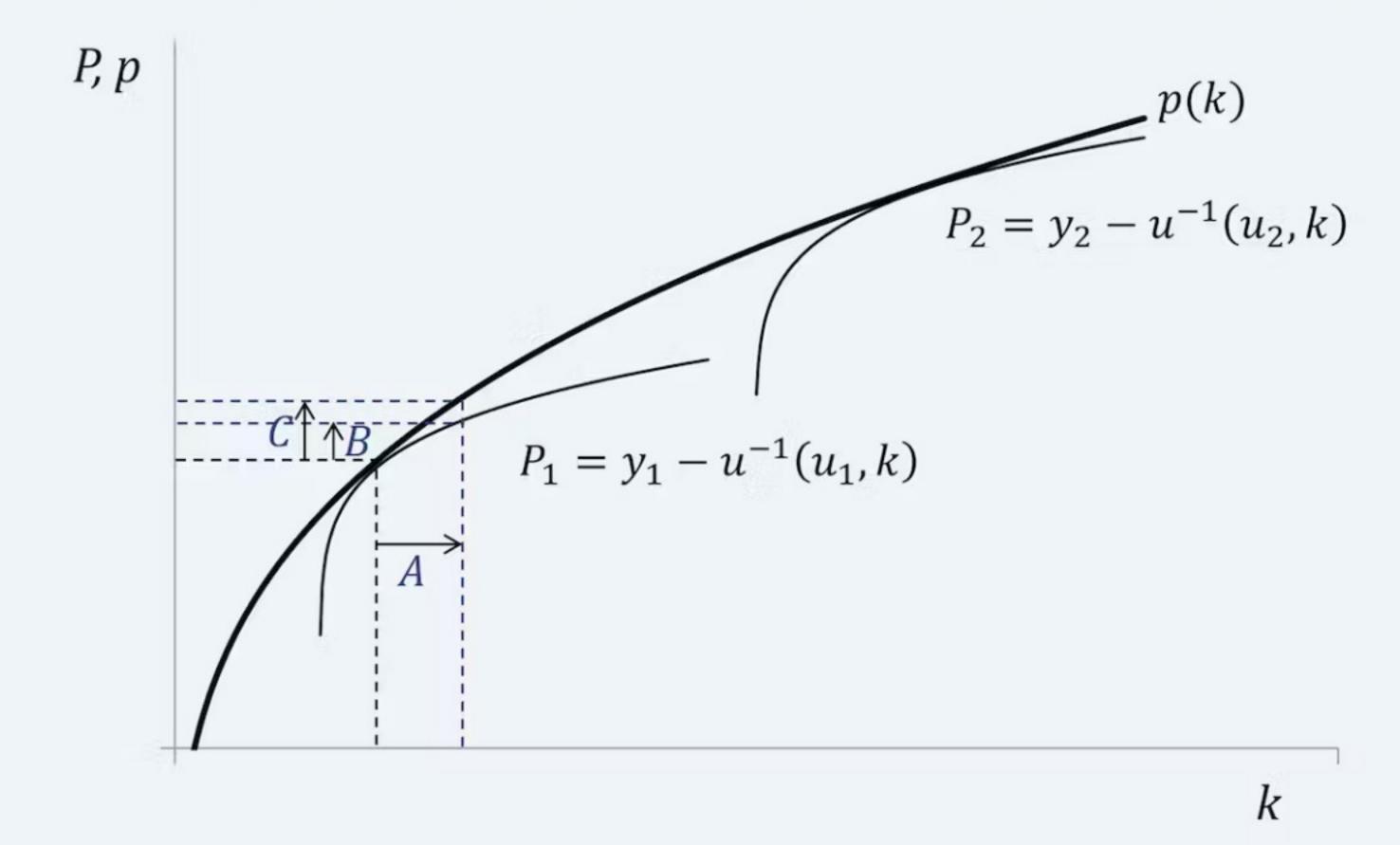
Homogeneity:

- In practice we usually calculate the average of the WTP
- We often do not attempt to estimate structural utility/demand parameters but focus on marginal changes
- Hence, we only identify the point on the value function that touches the equilibrium price function



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Structural (large) vs. marginal (small) changes in k

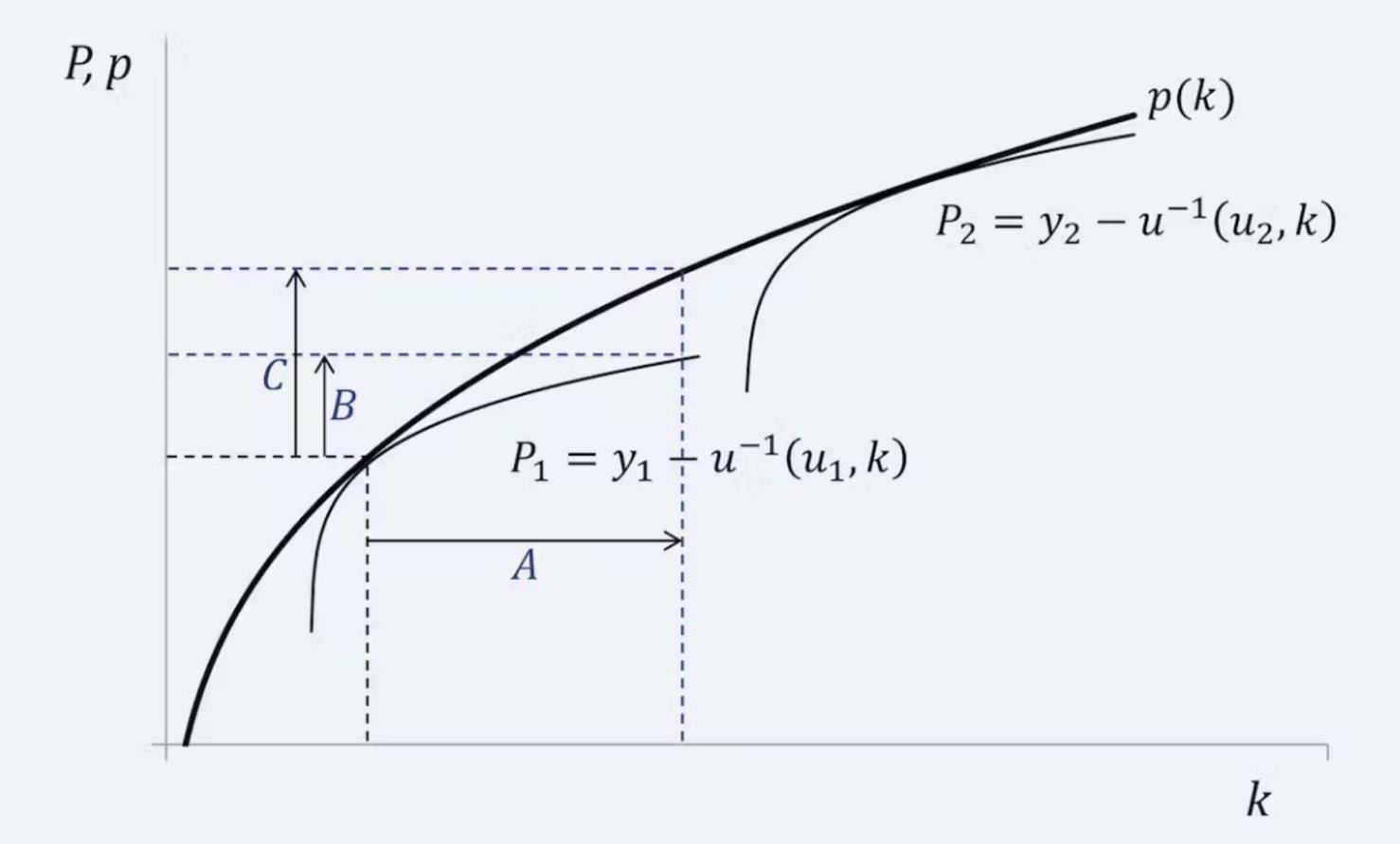


- Let's consider a small change in k
 - What is the actual willingness to pay?



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Structural (large) vs. marginal (small) changes in k



- Let's consider a large change in k
 - What is the willingness to pay?



4. Demand functions

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- To identify demand functions, Rosen (1974) suggested a three-step procedure
 - 1. Estimate a hedonic price function
 - 2. Calculate the implicit prices
 - 3. Estimate the inverse demand functions by a regression of marginal prices $\partial p/\partial k$ on the amount consumed of attribute k



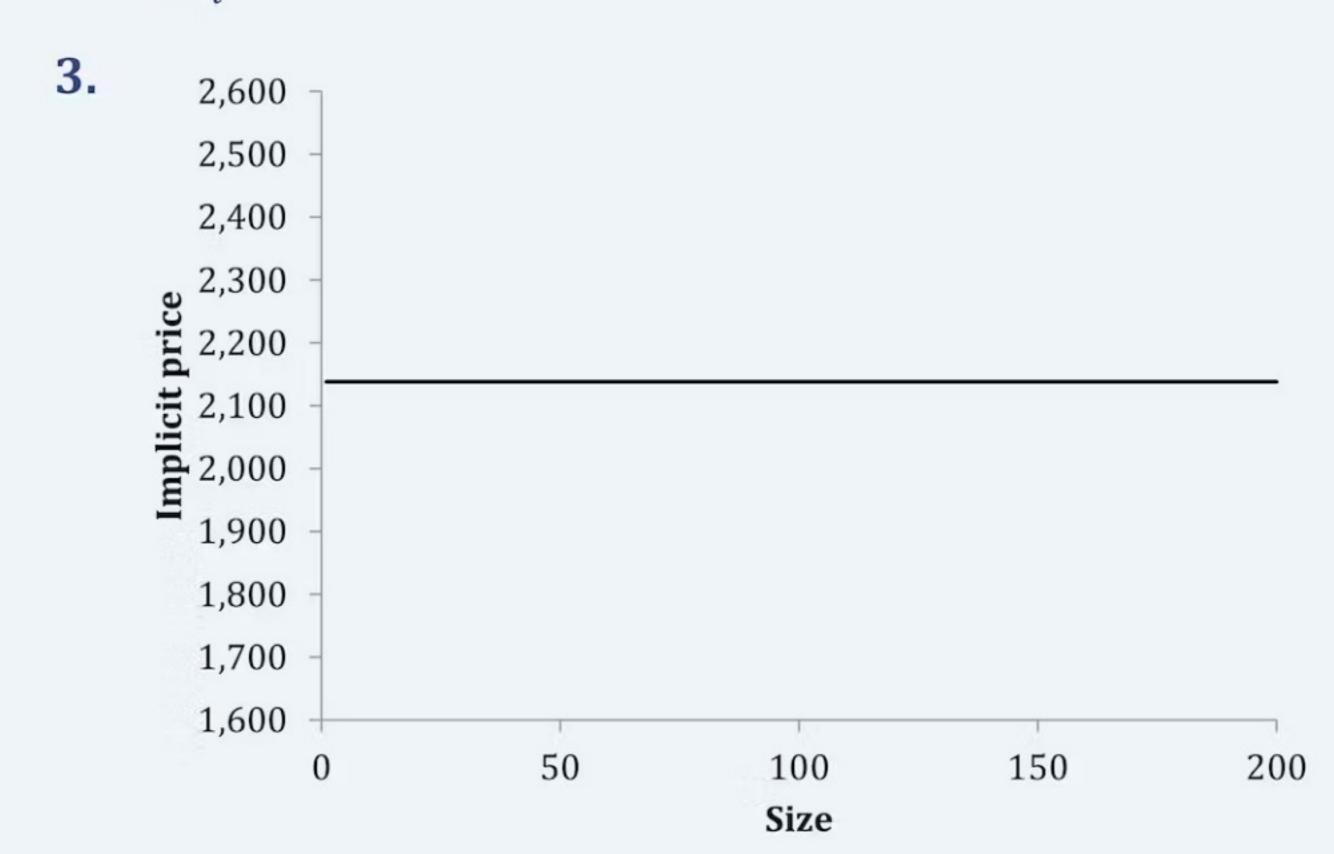
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Example linear hedonic price function:

1. Estimate hedonic price function

$$p_i = \alpha_0 + \alpha_1 k_i + \xi_i$$

$$2. \quad \frac{\partial p_i}{\partial k_i} = \hat{\alpha}_1 = 2139$$





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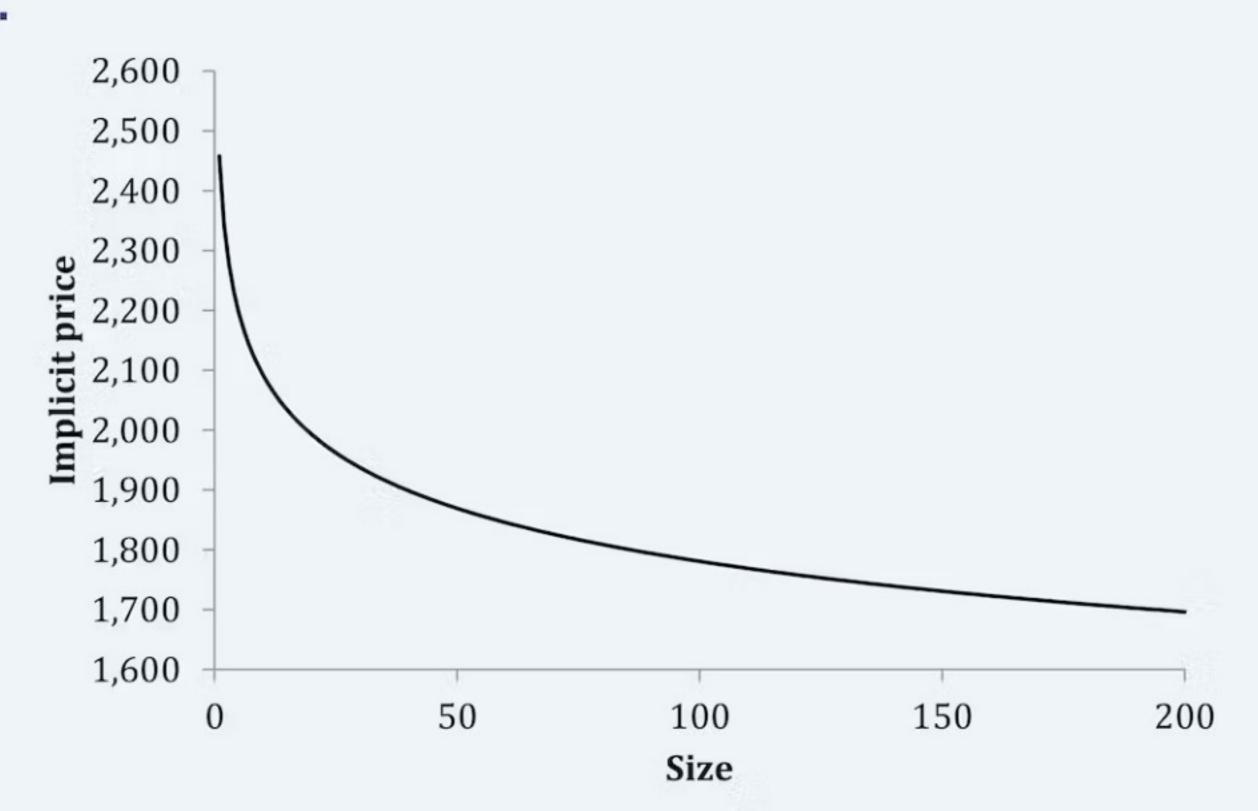
Example log-linear hedonic price function:

1. Estimate hedonic price function

$$\log p_i = \alpha_0 + \alpha_1 \log k_i + \xi_i$$

$$\mathbf{2.} \quad \hat{\alpha}_1 = 0.9276; \frac{\partial p_i}{\partial k_i} = \frac{\hat{\alpha}_1 p_i}{k_i}$$

3.





4. Demand functions

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- The procedure to obtain demand functions is misleading
 - Demand functions entirely depend on assumed functional form of hedonic price function



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- There are some solutions to the identification of demand functions:
 - Use multimarket data
 - > Utility functions are identical across markets but WTP is different
 - Use multiple observations for each individual
 - Bishop and Timmins (2018)
 - Nonparametric methods
 - Ekeland *et al.* (2004)
 - Bajari and Kahn (2005)



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Hedonic pricing (2)

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 Hedonic price functions are used to answer a lot of policy related questions

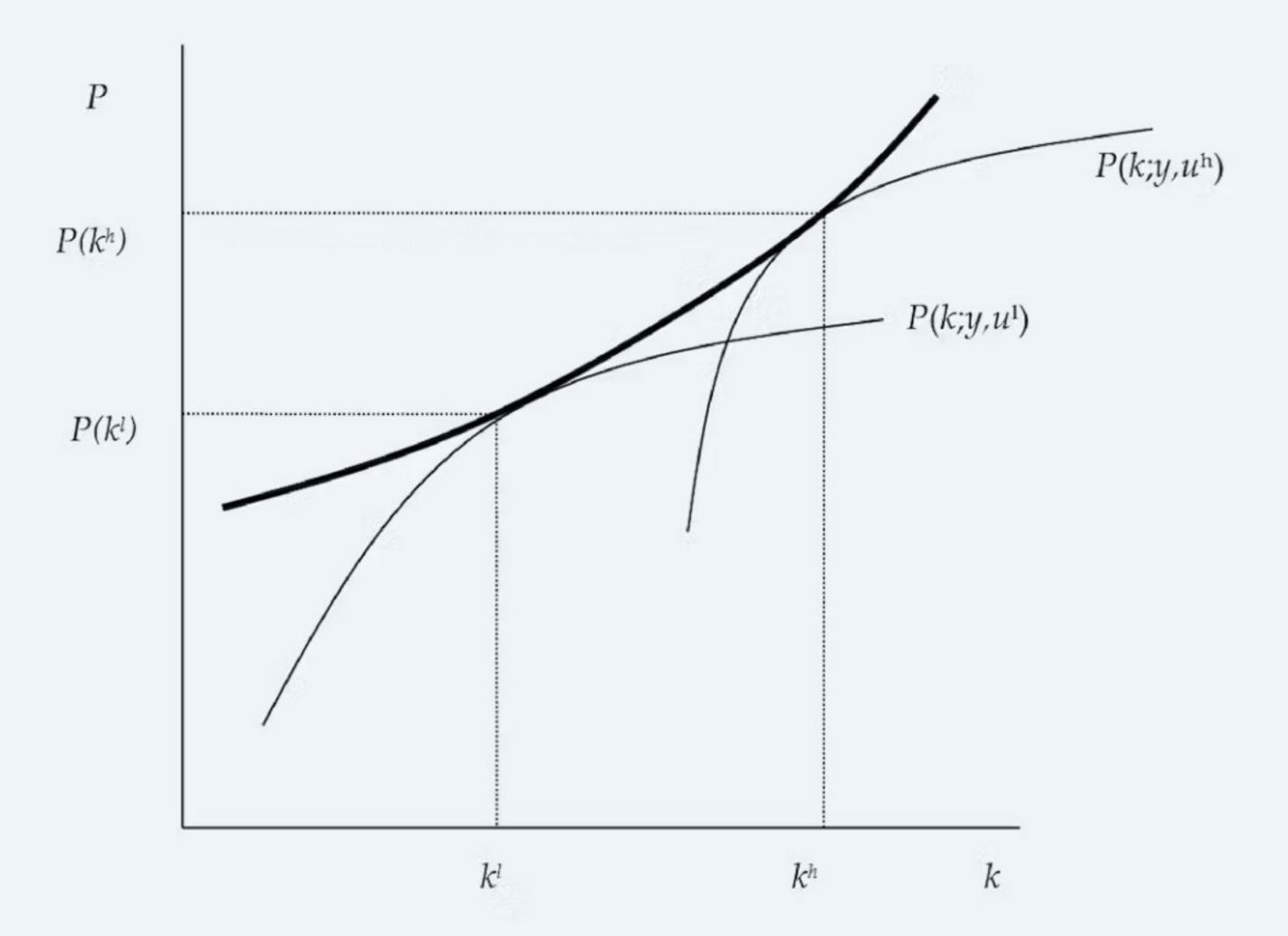
$$p_i = \alpha_0 + f(k_i) + \xi_i$$

- However:
 - 1. Misspecification
 - 2. Endogeneity



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 Recall: the hedonic price function is formed of different people attaching different values to a good





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• We have the following hedonic price function:

$$p_i = \alpha_0 + f(k_i) + \xi_i$$

- Functional form is unknown
 - May be highly nonlinear due to heterogeneity



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Logarithmic functions are often assumed:

$$\log p_i = \alpha_0 + \alpha_1 \log k_i + \xi_i$$
$$\log p_i = \alpha_0 + \alpha_1 k_i + \xi_i$$

Common feature of house price data: skewness
 ... issue of skewness is then addressed







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... But log-linear hedonic price functions can still be considered as restrictive

Recall that we had the following hedonic price function:

$$p_i = \alpha_0 + f(k_i) + \xi_i$$

- How should we estimate the above price function?
 - Nonparametric/semiparametric econometric techniques!
 - Put less structure on the data

- Nonparametric > no structure
- Semiparametric
 a bit of structure



2. Misspecification

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- Let's focus on two semiparametric estimation methods
 - Series approximation
 - Locally weighted regression (LWR)

- Series approximation
 - Estimate Taylor-series expansion
 - E.g., add k_i^2 and k_i^3 .
 - So, $\log p_i = \alpha_0 + \alpha_1 k_i + \alpha_2 k_i^2 + \alpha_3 k_i^3 + \xi_i$



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Series approximation

- Estimate Taylor-series expansion
- E.g., add k_i^2 and k_i^3 .

• So,
$$\log p_i = \alpha_0 + \alpha_1 k_i + \alpha_2 k_i^2 + \alpha_3 k_i^3 + \xi_i$$

Pros and cons

- Linear in parameters
- Problem if you have many explanatory variables
- Sometimes not flexible enough
- Performs poorly in 'boundary' regions



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- Locally weighted regression
 - Estimate for each observation a weighted regression
 - Let weights be higher for observations that are 'similar'
 - So, $\log p_i = \alpha_0 + \alpha_i k_i + \xi_i$ where *i* is the observation
 - Run weighted regression for each observation
 - Weights depend on value of k_i



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- Locally weighted regression
 - Weights depend on value of k:

$$w_i = \frac{1}{h\sqrt{2\pi\sigma_k}} e^{-\frac{1}{2}\left(\frac{k-k_i}{\sigma_k h}\right)^2}$$

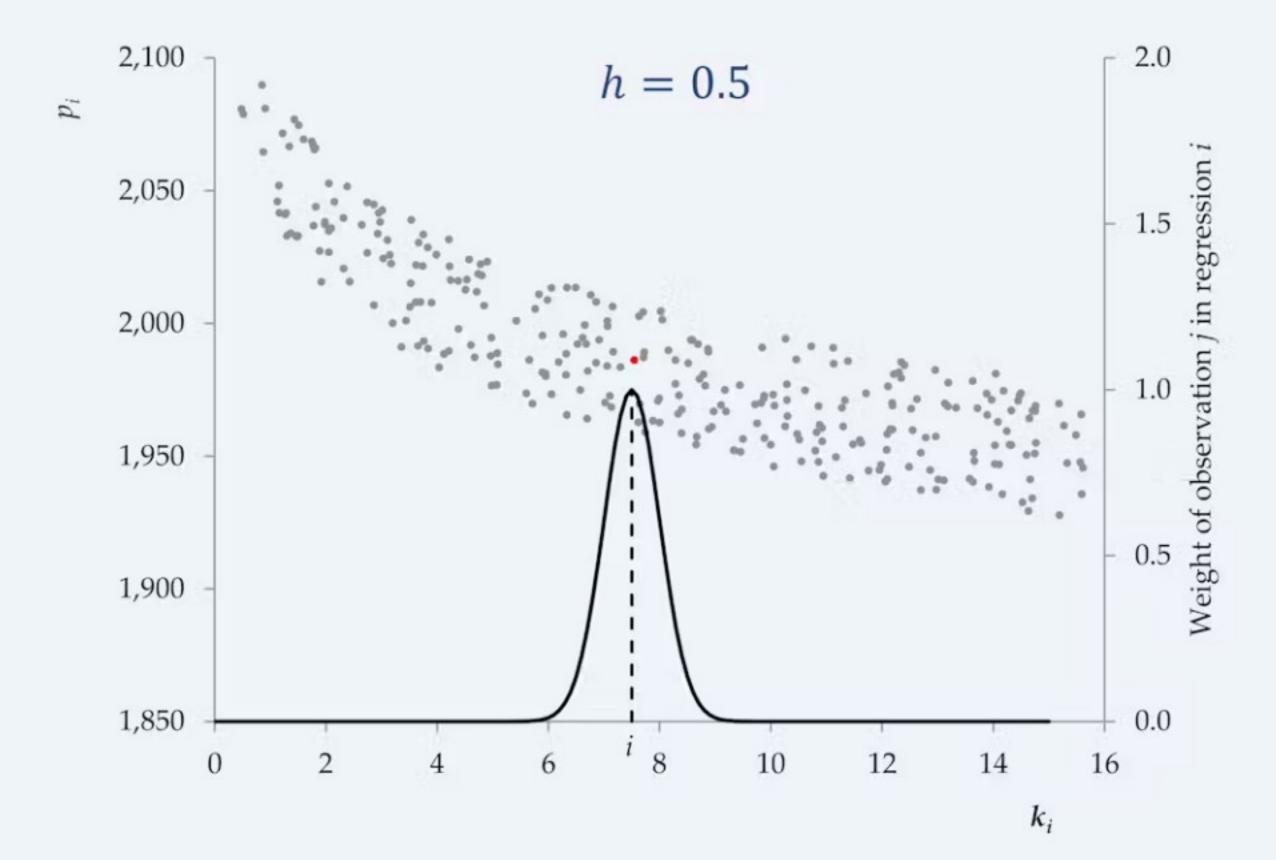
- Looks difficult, but weights are based on a normal distribution
- h is the bandwidth
 - h = 0, only take observation i into account
 - $h \to \infty$, identical to linear regression



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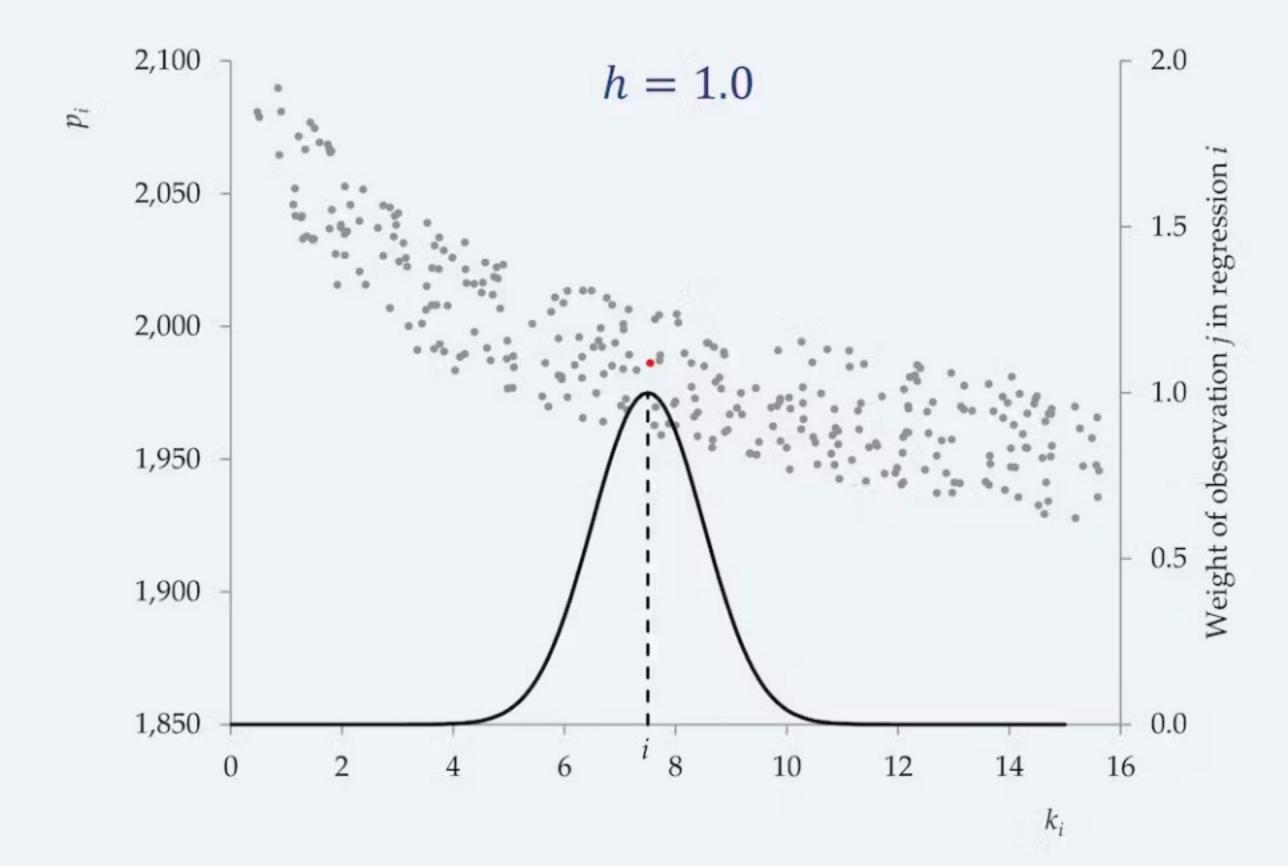




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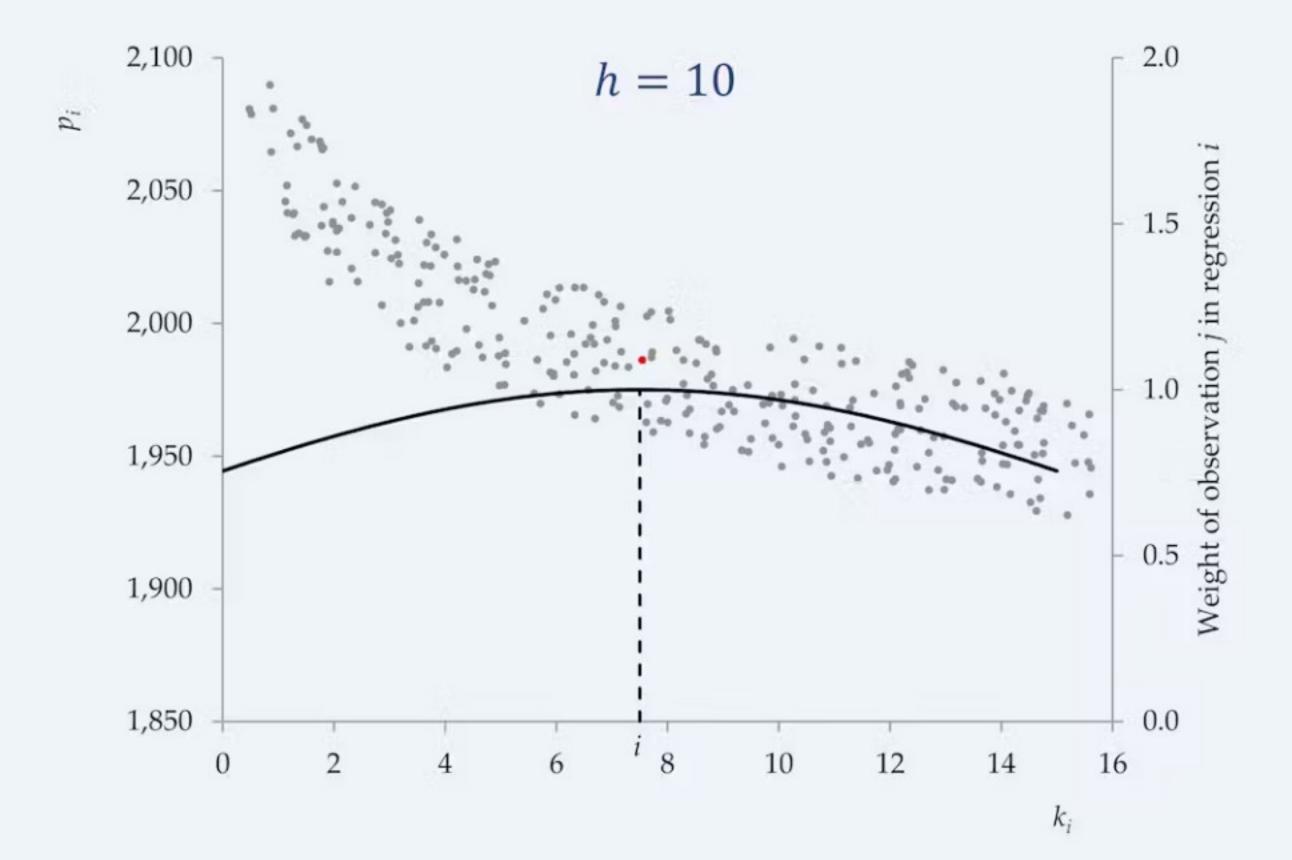




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- Locally weighted regression
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- Locally weighted regression
 - Very flexible
 - Easy to estimate implicit prices $(\partial p_i/\partial k_i = \alpha_i)$
 - Bandwidth is important parameter; determines smoothness
 - · Becomes popular in applied research
 - > Bajari and Kahn (2005)
 - > McMillen and Redfearn (2010)
 - Use for example NPREGRESS command in STATA

- But: computationally intensive!
 - It takes very long to estimate when $N \gtrsim 50000$



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• How to estimate partially linear functions?

$$p_i = f(k_i) + \beta x_i + \xi_i$$

- Use series estimation
- Use Robinson's procedure
 - > PLREG in STATA



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- Omitted variable bias (OVB) is a big issue when estimating hedonic price functions
 - So: $\log p_i = \alpha_0 + \alpha_1 \log k_i + \xi_i$
 - If $E[\xi_i | k_i] \neq 0$, α_1 will be inconsistent



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Hedonic pricing:

- 1. Understand what a hedonic price function
 - A description of the equilibrium prices of varieties of a heterogeneous good
- 2. Have basic knowledge about how a hedonic price function is linked to economic theory
 - Heterogeneous households have different WTPs for goods
- 3. Understand how to address misspecification and endogeneity when estimating hedonic price functions
 - Estimate log-linear hedonic price functions
 - Series approximation / LWR
 - Add controls and fixed effects
 - Consider IV/quasi experiments



Hedonic pricing (2)

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