

REM Tutorial 3: Regression I - Hedonic

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Hedonic (House) Pricing

- ▶ Heterogeneous goods → buying a house always means selecting a whole bundle of housing characteristics
- ▶ No price per standard quantity or quality of housing available (observable market price is an outcome of the unobservable utility from house ownership and use)
- ▶ **Hedonic price equation:** observable market price is modeled as a function of observable housing characteristics

$$P = \alpha + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_n X_n \quad (1)$$

with:

P = market price

α = constant term

X_i = characteristic i ($i = 1, \dots, n$), e.g., size, age, neighbourhood

β_i = **implicit increase in price** for an additional unit of characteristic i

- ▶ Various applications in empirical research (see, e.g., academic GfT: https://www.youtube.com/watch?v=0_NreXsVvVY&t=12s
Paper: <https://link.springer.com/article/10.1007/s11116-022-10295-8>)

Today's Agenda

| Learning Outcomes | R Functions | Packages | Data |
|---|--|---|-------------------|
| Be able to plot kernel density estimates | <code>max(); density()</code> <code>polygon()</code> | | Lucas_County_data |
| Know how to add a curve to a plot | <code>curve(); dnorm()</code> <code>mean(); sd()</code> | | |
| Be able to perform the Jarque-Bera test in R | <code>jarque.bera.test()</code> | <code>tseries</code> <code>quadprog</code> | |
| Be able to run an OLS regression using R & assess regression output | <code>lm()</code> <code>summary()</code> | | |
| Be able to produce a two-way scatter plot & remove scientific notation from axes & add a fitted regression line to the plot | <code>plot()</code> <code>options()</code> <code>abline()</code> | | |
| Understand/recap common functional forms | | | |
| Know how to include & interpret a quadratic term in a linear regression model | <code>I()</code> | | |
| Be able to plot a non-linear fitted line & determine minimum/maximum point | <code>data.frame(); seq()</code> <code>min(); max()</code> <code>nrow(); predict()</code> <code>as.data.frame()</code> <code>lines(); abs()</code> | | |
| Know how to interpret coefficient estimates for categorical variables | <code>as.factor()</code> <code>levels()</code> | | |
| Know how to produce contingency tables | <code>table()</code> | | |

Practice

Work through **Tutorial 3 - Workbook** to solve the practical problems provided in **Tutorial 3 - Problem Set**.

Follow up

I'm adding some code of what I showed in class on alternative ways to generate squared terms etc. (not contained in Workbook):

Alternatively, we could generate age squared manually and add this variable to regression equation:

```
dat1$age2 <- (dat1$age)^2 # gen. age squared and add to dat1  
fit2.2 <- lm(price ~ age + age2, data = dat1)  
summary(fit2.2) # same result
```

Or we use the poly function to generate age^2 age^3 etc. and write result to **dat1**:

```
age.poly <- poly(dat1$age, 3, raw = TRUE) # evaluates raw, not  
orthogonal, polynomials to the constant polynomial (degree=0)  
head(age.poly) # to view: 1st col is age, 2nd is age2, 3rd is age3  
# add to dat1 in one step  
dat1$age3 <- poly(dat1$age, 3, raw = TRUE)[,3]
```