

III. Simple Regression.

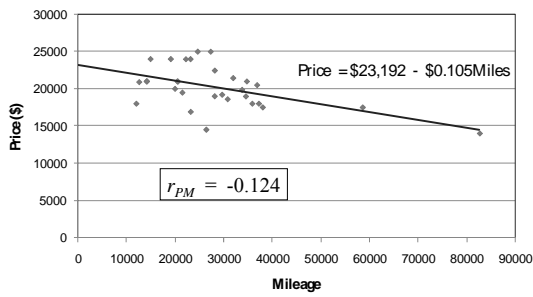
- A. Introduction
- B. Population Regression Equation
- C. Sample Regression Equation
- D. Ordinary Least Squares
- 7. Interpretations

- E. Classical Regression Model.
- F. Properties of OLS Estimators (**BLUE**)

Personalized Used 2004 Honda Accord
Samples.

- Personalized PRS questions in class tomorrow.
- Go to the course website.
- Download **your** data from the course website – same as the data used for Exam 1.
- Create an Excel Spreadsheet. Include:
 - Means, Standard Deviations.
 - Correlations
 - XY Scatter Diagram
 - OLS Regression Results
- Bring these with you to class – we'll use them for **Personalized PRS Questions**.

Relationship between Price and Mileage for Used 2004 Honda Accords: January 2006.



E. Classical Regression Model

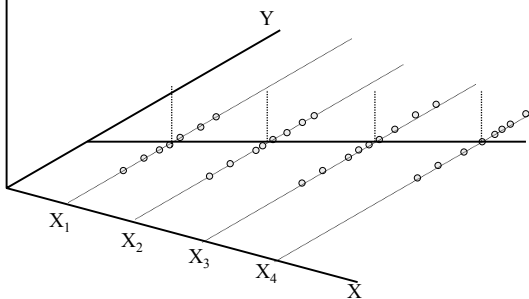
1. Introduction – CRM why do we need this?
 - What does OLS provide?
 - Is this enough?

2. CRM Assumptions:

- 1) $Y_i = \beta_0 + \beta_1 X_i + u_i$ is the *true model*.
- 2) X is *not a random variable*.
- 3) $E[u | X_i] = 0$, at each level of X .
- 4) $Var(u_i) = \sigma^2$, for all levels of X .
(or $E[u_i] = \sigma^2$ - **Homoskedasticity**)
- 5) $Cov(u_i, u_j) = 0$, for $i \neq j$.
(or $E(u_i u_j) = 0$ - **Non-autocorrelation**)
- 6) u is distributed normally.

f(u)

Our view of the population relationship with the CRM assumptions.



F. Properties of OLS

Estimators

- What properties do we want for our estimators?

- What are we trying to describe?
- Use expected values to determine:
Center –

Variance –

OLS Estimators

$$\hat{\beta}_1 = \frac{\sum X_i Y_i - n \bar{X} \bar{Y}}{\sum X_i^2 - n \bar{X}^2} = \frac{\sum x_i y_i}{\sum x_i^2} = \frac{\sum x_i Y_i}{\sum x_i^2}$$

$$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X}$$

Given the **CRM Assumptions**, the **Sampling Distributions** have the following **centers** and **variances**:

$$E[\hat{\beta}_0] = \beta_0 \quad \sigma_{\hat{\beta}_0} = \sqrt{\frac{\sigma^2 \sum X_i^2}{n \sum x_i^2}}$$

$$E[\hat{\beta}_1] = \beta_1 \quad \sigma_{\hat{\beta}_1} = \sqrt{\frac{\sigma^2}{\sum x_i^2}}$$
