

, 26, 27, 28, 29, 30

**Stat 231 - Statistics**

**Spring 2017**

Lecture 25, 26, 27, 28, 29, 30: June 26 - July 10, 2017

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## 25.1 Measurement Bias Testing

**Objective :** To test whether a scale is biased

**Experiment :** We take an object of a known weight (10) and measure of using the scale n times.

$Y_i$  = ith reading of the scale

Let  $S$  = bias of our scale and  $R_i$  error on the ith measurement

$$Y_i = \text{true weight} + S + R_i$$

**Conventions :**

- If d is to the right of the median of the  $x^2$   
 $p - \text{value} = 2P(D \geq d)$
- If d is to the left of the median  $p - \text{value} = 2P(D \leq d)$

## 25.2 Testing Using The Likelihood

Suppose we are not able to find D fro some problem. Assume n is large. We can then use the likelihood ratio test statistic as our discrepancy measure.

$$\triangle(\theta) = -2\log \frac{L(\theta)}{L(\hat{\theta})}$$

$$D = \left| \frac{\tilde{\theta} - \theta_o}{\sqrt{\frac{\theta(1-\theta)}{n}}} \right|$$

1. Find  $L(\theta)$  and calculate  $\hat{\theta}$
2. Calculate  $\triangle(\theta_o)$
3. Calculate the p-value where  $d = \triangle$

## 25.3 Introduction To Regression Models

A linear regression model assumes that  $E(y)$  is a linear function of  $x$

$$E(Y) = \alpha + \beta X$$

## 25.4 Simple Linear Regression Model

### Assumptions

- Given  $x_i$ 's,  $Y_i$ 's have a Normal distribution.
- The mean of the  $Y_i$ 's are a linear function of the  $x_i$ 's
- The variance of the  $Y_i$ 's are constant  $\sigma^2$  (unknown)  
 $\sigma^2$  is independent of  $x$
- $\sigma^2$  is not a function of  $x$  and  $Y_i$ 's are independent

### 25.4.1 Model

$$Y_i \sim G(\alpha + \beta x_i, \sigma)$$

$$\text{Alternate Form : } Y_i = \alpha + \beta x_i + R_i$$

### 25.4.2 Gauss-Markov Assumptions

- $Y_i$ 's independent
- $Y_i$ 's are Gaussian
- $E(Y_i) = \alpha + \beta x_i$
- $Var(Y_i) = \sigma^2 \forall x$
- $E(Y_i)$  is not linear  $\longrightarrow$  Non-linear regression models
- $V(Y_i) \neq \sigma^2 \longrightarrow$  Heteroscedastic Models
- More than one explanatory variable  $\longrightarrow$  Multivariable Regression

### 25.4.3 Questions

- Interpretation of  $\alpha, \beta, \sigma$  ?  
 $\beta$  is the increase in the mean of  $Y$  if  $x$  goes by 1 unit  
 $\alpha = E(Y)$  when  $x = 0$
- MLE for  $\alpha, \beta, \sigma$

$$- \hat{\alpha} = \bar{y} - \hat{\beta}\bar{x} \text{ where } \hat{\beta} = \frac{S_{xy}}{S_{xx}}$$

$$- \hat{\sigma}^2 = \frac{1}{n}[X_{yy} - \hat{\beta}S_{xy}]$$

$$- S_{yy} = \sum (y_i - \bar{y})^2$$

$$- S_{xx} = \sum (x_i - \bar{x})^2$$

$$- S_{xy} = \sum (x_i - \bar{x})(y_i - \bar{y})$$

$$S^2 = \frac{1}{n-2}[S_{yy} - \beta S_{xy}]$$

$$\text{Let } a_i = \frac{1}{S_{xx}} \cdot (x_i - \bar{x})$$

$$\hat{p} = \sum a_i y_i$$

#### 25.4.4 Confidence interval for p

- $E(\tilde{p}) = \alpha\sigma + \beta = \beta$

- $V(\beta) = \frac{\sigma^2}{S_{xx}}$

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$$\frac{\tilde{\beta} - \beta}{\frac{S}{\sqrt{S_{xx}}}} \sim T_{n-2}$$

- Coverage Interval

$$\tilde{\beta} \pm t^* \frac{S}{\sqrt{S_{xx}}}$$

- D value for hypothesis testing

$$D = \left| \frac{\tilde{\beta} - \beta_0}{\frac{S}{\sqrt{S_{xx}}}} \right|$$