

Lecture 39

Simple Linear Regression: ANOVA & Coefficient of Determination

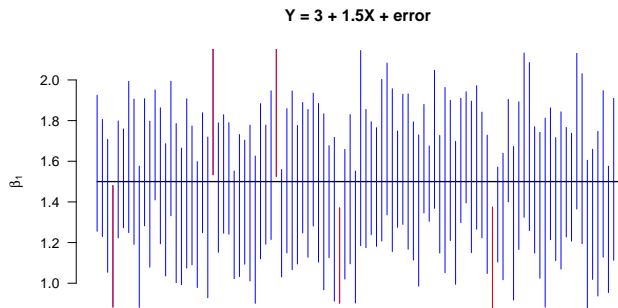
STAT 8010 Statistical Methods I

December 4, 2019

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Understanding Confidence Intervals

- Suppose $Y = \beta_0 + \beta_1 X + \varepsilon$, where $\beta_0 = 3$, $\beta_1 = 1.5$ and $\sigma^2 \sim N(0, 1)$
- We take 100 random sample each with sample size 20
- We then construct the 95% CI for each random sample (\Rightarrow 100 CIs)



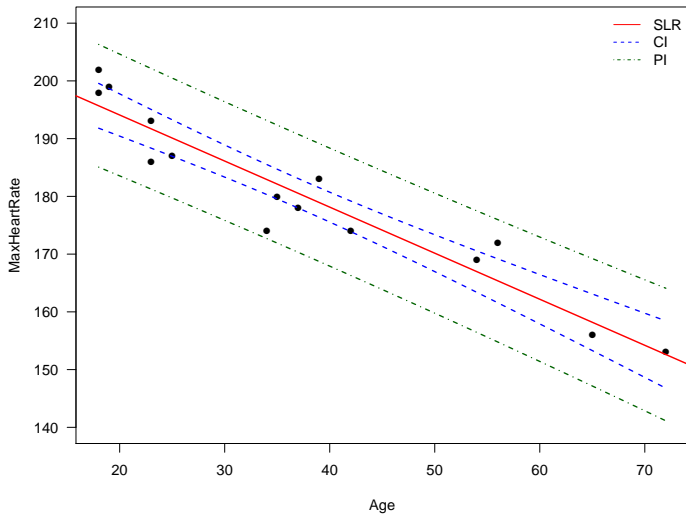
Confidence Intervals vs. Prediction Intervals

Simple Linear
Regression: ANOVA
& Coefficient of
Determination

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Review of Last Class

Analysis of Variance
(ANOVA) Approach to
Regression



Partitioning Sums of Squares

- Total sums of squares in response

$$SST = \sum_{i=1}^n (Y_i - \bar{Y})^2$$

- We can rewrite SST as

$$\begin{aligned} \sum_{i=1}^n (Y_i - \bar{Y})^2 &= \sum_{i=1}^n (Y_i - \hat{Y}_i + \hat{Y}_i - \bar{Y})^2 \\ &= \underbrace{\sum_{i=1}^n (Y_i - \hat{Y}_i)^2}_{\text{Error}} + \underbrace{\sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2}_{\text{Model}} \end{aligned}$$

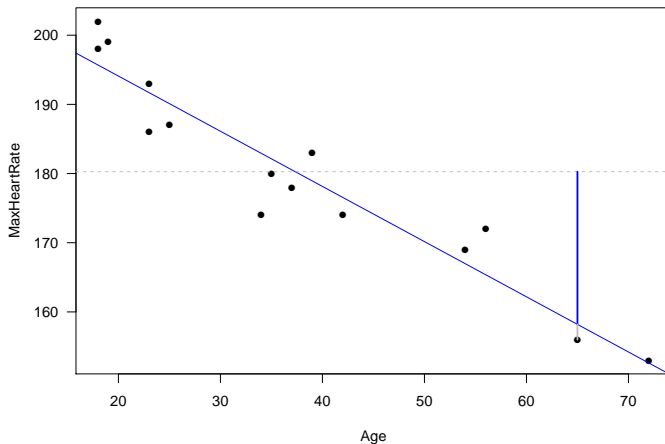
Partitioning Total Sums of Squares

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Total Sum of Squares: SST

- If we ignored the predictor X , the \bar{Y} would be the best (linear unbiased) predictor

$$Y_i = \beta_0 + \varepsilon_i \quad (1)$$

- SST is the sum of squared deviations for this predictor (i.e., \bar{Y})
- The **total mean square** is $SST/(n - 1)$ and represents an unbiased estimate of σ^2 under the model (1).

Regression Sum of Squares: SSR

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- SSR: $\sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2$
- Degrees of freedom is 1 due to the inclusion of the **slope**, i.e.,

$$Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i \quad (2)$$

- “Larg” MSR = SSR/1 suggests a linear trend, because

$$E[MSE] = \sigma^2 + \beta_1^2 \sum_{i=1}^n (X_i - \bar{X})^2$$

- SSE is simply the sum of squared residuals

$$SSE = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$$

- Degrees of freedom is $n - 2$ (Why?)
- SSE large when |residuals| are “large” $\Rightarrow Y_i$ ’s vary substantially around fitted regression line
- $MSE = SSE/(n - 2)$ and represents an unbiased estimate of σ^2 **when taking X into account**

ANOVA Table and F test

Source	df	SS	MS
Model	1	$SSR = \sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2$	$MSR = SSR/1$
Error	$n - 2$	$SSE = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$	$MSE = SSE/(n-2)$
Total	$n - 1$	$SST = \sum_{i=1}^n (Y_i - \bar{Y})^2$	

- **Goal:** To test $H_0 : \beta_1 = 0$
- Test statistics $F^* = \frac{MSR}{MSE}$
- If $\beta_1 = 0$ then F^* should be near one \Rightarrow reject H_0 when F^* "large"
- We need sampling distribution of F^* under $H_0 \Rightarrow F_{1,n-2}$, where $F(d_1, d_2)$ denotes a F distribution with degrees of freedom d_1 and d_2

F Test: $H_0 : \beta_1 = 0$ vs. $H_a : \beta_1 \neq 0$

```
fit <- lm(MaxHeartRate ~ Age)
anova(fit)
```
```

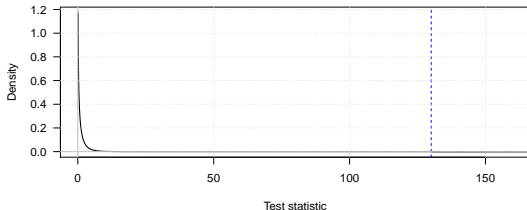


### Analysis of Variance Table

Response: MaxHeartRate

|           | Df | Sum Sq    | Mean Sq | F value |
|-----------|----|-----------|---------|---------|
| Age       | 1  | 2724.50   | 2724.50 | 130.01  |
| Residuals | 13 | 272.43    | 20.96   |         |
|           |    | Pr(>F)    |         |         |
| Age       |    | 3.848e-08 | ***     |         |

Null distribution of F test statistic



## ANOVA Table and F-Test

### Analysis of Variance Table

Response: MaxHeartRate

|           | Df      | Sum Sq  | Mean Sq   |
|-----------|---------|---------|-----------|
| Age       | 1       | 2724.50 | 2724.50   |
| Residuals | 13      | 272.43  | 20.96     |
|           | F value |         | Pr(>F)    |
| Age       | 130.01  |         | 3.848e-08 |

## Parameter Estimation and T-Test

Coefficients:

|             | Estimate  | Std. Error | t value | Pr(> t ) |
|-------------|-----------|------------|---------|----------|
| (Intercept) | 210.04846 | 2.86694    | 73.27   | < 2e-16  |
| Age         | -0.79773  | 0.06996    | -11.40  | 3.85e-08 |

- **Pearson Correlation:**  $r = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^n (X_i - \bar{X})^2 \sum_{i=1}^n (Y_i - \bar{Y})^2}}$
- $-1 \leq r \leq 1$  measures the strength of the **linear relationship** between  $Y$  and  $X$
- We can show

$$r = \hat{\beta}_1 \sqrt{\frac{\sum_{i=1}^n (X_i - \bar{X})^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2}},$$

this implies

$$\beta_1 = 0 \text{ in SLR} \Leftrightarrow \rho = 0$$

## Coefficient of Determination $R^2$

- Defined as the proportion of total variation explained by SLR

$$R^2 = \frac{\sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2} = \frac{SSR}{SST} = 1 - \frac{SSE}{SST}$$

- We can show  $r^2 = R^2$ :

$$\begin{aligned} r^2 &= \left( \hat{\beta}_{1,LS} \sqrt{\frac{\sum_{i=1}^n (X_i - \bar{X})^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2}} \right)^2 \\ &= \frac{\hat{\beta}_{1,LS}^2 \sum_{i=1}^n (X_i - \bar{X})^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2} \\ &= \frac{SSR}{SST} \\ &= R^2 \end{aligned}$$

## Maximum Heart Rate vs. Age: $r$ and $R^2$

Simple Linear  
Regression: ANOVA  
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Review of Last Class

Analysis of Variance  
(ANOVA) Approach to  
Regression

```
> summary(fit)$r.squared
[1] 0.9090967
> cor(Age, MaxHeartRate)
[1] -0.9534656
```

### Interpretation:

There is a strong negative linear relationship between `MaxHeartRate` and `Age`. Furthermore,  $\sim 91\%$  of the variation in `MaxHeartRate` can be explained by `Age`.

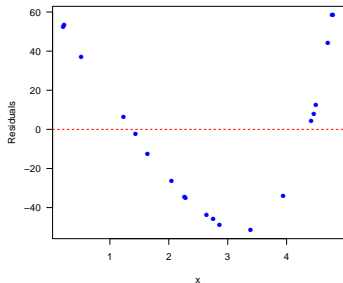
# Residual Plot Revisited

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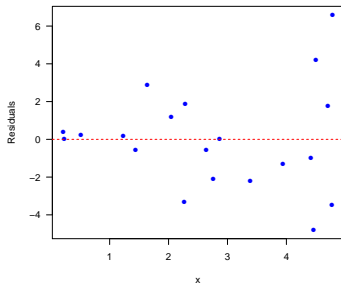
Review of Last Class

Analysis of Variance  
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⇒ Nonlinear relationship

- Transform  $X$
- Nonlinear regression



⇒ Non-constant variance

- Transform  $Y$
- Weighted least squares

# Summary

Simple Linear  
Regression: ANOVA  
& Coefficient of  
Determination



Review of Last Class

Analysis of Variance  
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Regression

In this lecture, we learned **ANOVA Approach to Regression**  
and **Coefficient of Determination**

Next time: **Review**