

Lecture 26

Simple Linear Regression: ANOVA Approach to Regression

Text: Chapter 11

STAT 8010 Statistical Methods I
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Simple Linear Regression: ANOVA Approach to Regression

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Review of Last Class

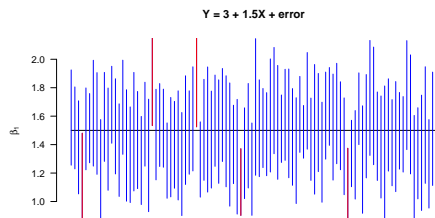
Analysis of Variance (ANOVA) Approach to Regression

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Notes

Understanding Confidence Intervals

- Suppose $Y = \beta_0 + \beta_1 X + \epsilon$, 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)



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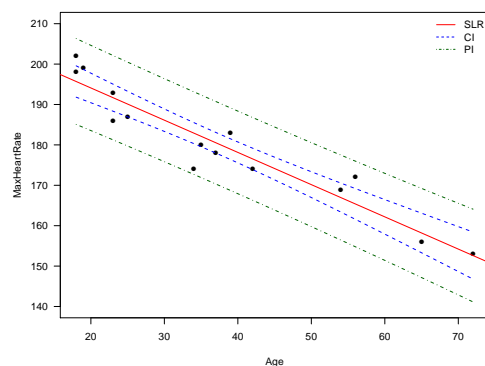
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Notes

Confidence Intervals vs. Prediction Intervals



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Notes

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

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

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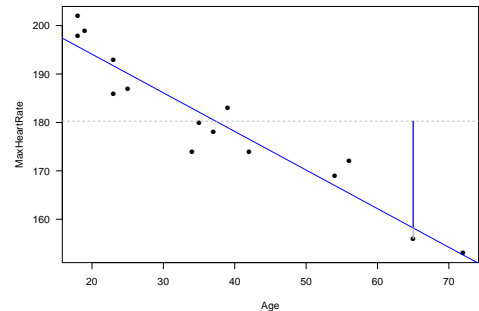
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Notes

Partitioning Total Sums of Squares



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Notes

Total Sum of Squares: SST

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

$$Y_i = \beta_0 + \varepsilon_i \tag{1}$$

- 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).

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Regression Sum of Squares: SSR

- 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.,
$$Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i \tag{2}$$
- “Large” MSR = SSR/1 suggests a linear trend, because

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

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Error Sum of Squares: SSE

- 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**

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Notes

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

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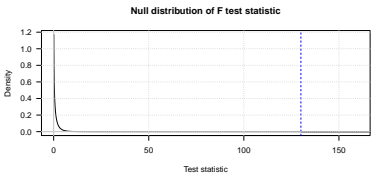
F Test: $H_0 : \beta_1 = 0$ vs. $H_a : \beta_1 \neq 0$

```
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	***	



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Notes

SLR: F-Test vs. T-test

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

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Notes

Correlation and Simple Linear Regression

- **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

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

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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 $r^2 = R^2$:

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

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Notes

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

```
> 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, ~91% of the variation in MaxHeartRate can be explained by Age.

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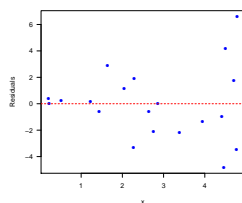
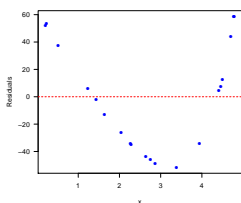
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Notes

Residual Plot Revisited



⇒ Nonlinear relationship

⇒ Non-constant variance

- Transform X

- Transform Y

- Nonlinear regression

- Weighted least squares

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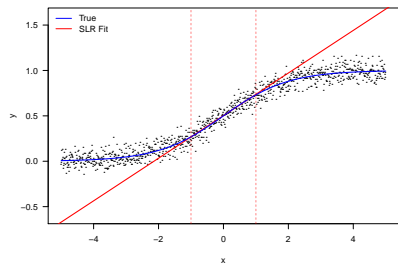
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Notes

Extrapolation in SLR



Extrapolation beyond the range of the given data can lead to **seriously biased estimates** if the **assumed relationship does not hold the region of extrapolation**

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Summary of SLR

- **Model:** $Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i$
- **Estimation:** Use the **method of least squares** to estimate the parameters
- **Inference**
 - Hypothesis Testing
 - Confidence/prediction Intervals
 - ANOVA
- **Model Diagnostics and Remedies**

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