

Lecture 6

Simple Linear Regression V & Introduction to Multiple Linear Regression

Reading: Chapter 11, 12

STAT 8020 Statistical Methods II
September 2, 2019

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Agenda

Simple Linear
Regression V &
Introduction to
Multiple Linear
Regression

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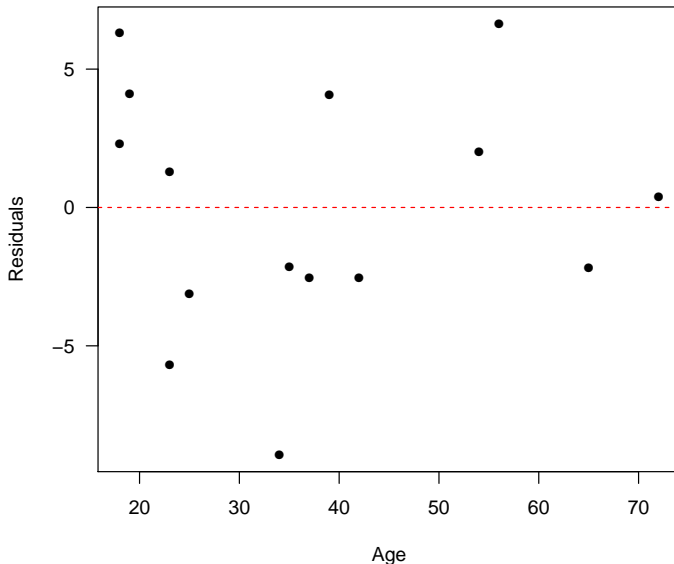
Regression
Diagnostics and
Remedies

Multiple Linear
Regression

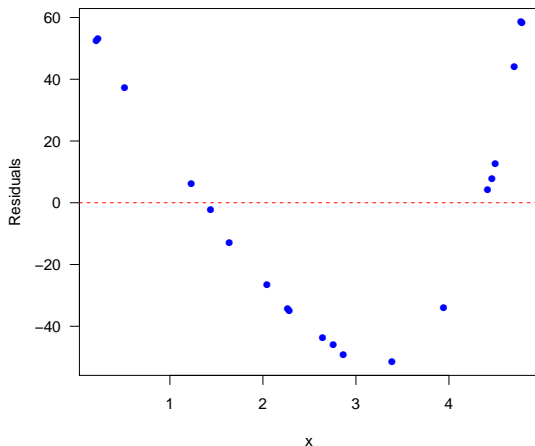
1 Regression Diagnostics and Remedies

2 Multiple Linear Regression

MaxHeartRate vs. Age Residual Plot Revisited



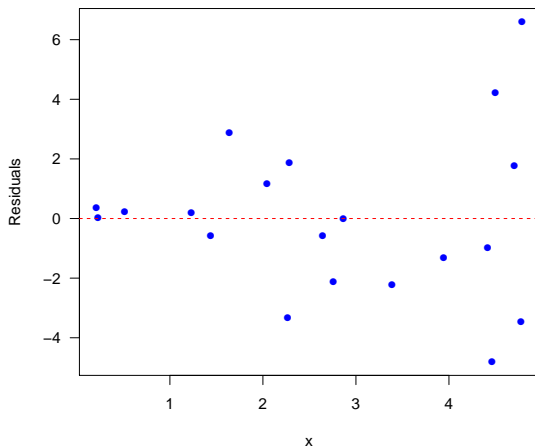
A Non-Linear Pattern



Possible Remedies:

- Transform X
- Nonlinear regression

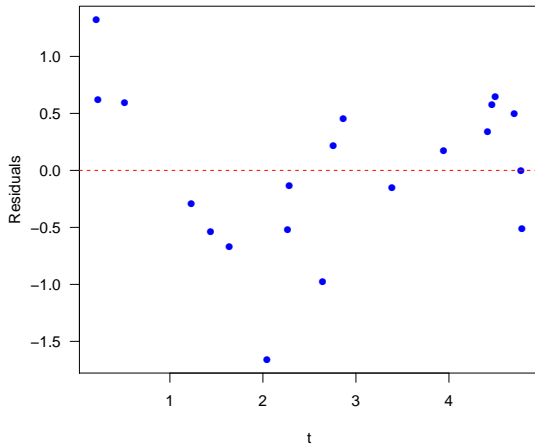
Non-Constant Variance



Possible Remedies:

- Transform Y
- Weighted least squares

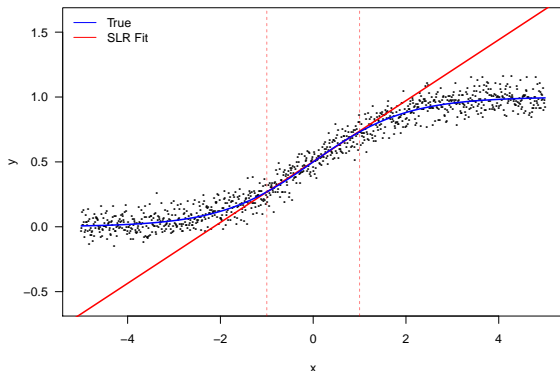
Correlated Errors



A Possible Remedy:

- Allow correlated errors in SLR

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

- **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**

Multiple Linear Regression

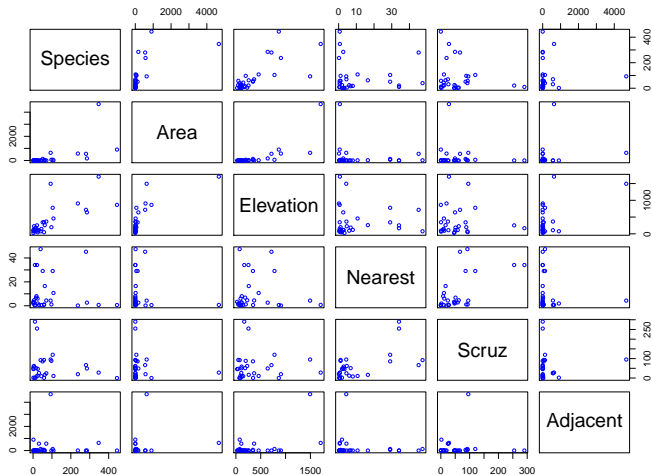
Goal: To model the relationship between two or more explanatory variables (X 's) and a response variable (Y) by fitting a **linear equation** to observed data:

$$Y_i = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \cdots + \beta_{p-1} X_{p-1} + \varepsilon_i, \quad \varepsilon_i \stackrel{i.i.d.}{\sim} N(0, \sigma^2)$$

Example: Species diversity on the Galapagos Islands. We are interested in studying the relationship between the number of plant species (Species) and the following geographic variables: Area, Elevation, Nearest, Scrub, Adjacent.



How Do Geographic Variables Affect Species Diversity?



$$\text{Species} = \beta_0 + \beta_1 \text{Area} + \beta_2 \text{Elevation} + \beta_3 \text{Nearest} + \beta_4 \text{Scrub} + \beta_5 \text{Adjacent} + \text{error}$$

Fit a Multiple Linear Regression using R

```
lm(formula = Species ~ Area + Elevation + Nearest + Scrub + Adjacent,  
    data = gala)
```

Residuals:

Min	1Q	Median	3Q	Max
-111.679	-34.898	-7.862	33.460	182.584

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	7.068221	19.154198	0.369	0.715351
Area	-0.023938	0.022422	-1.068	0.296318
Elevation	0.319465	0.053663	5.953	3.82e-06
Nearest	0.009144	1.054136	0.009	0.993151
Scrub	-0.240524	0.215402	-1.117	0.275208
Adjacent	-0.074805	0.017700	-4.226	0.000297

(Intercept)

Area

Elevation ***

Nearest

Scrub

Adjacent ***

Signif. codes:

0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 60.98 on 24 degrees of freedom

Multiple R-squared: 0.7658, Adjusted R-squared: 0.7171

F-statistic: 15.7 on 5 and 24 DF, p-value: 6.838e-07

Multiple Linear Regression in Matrix Notation

$$\mathbf{Y} = \begin{pmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_n \end{pmatrix}, \quad \mathbf{X} = \begin{pmatrix} 1 & X_{1,1} & X_{2,1} & \cdots & X_{p-1,1} \\ 1 & X_{1,2} & X_{2,2} & \cdots & X_{p-1,2} \\ \vdots & \cdots & \ddots & \vdots & \\ 1 & X_{1,n} & X_{2,n} & & X_{p-1,n} \end{pmatrix}$$

We can express MLR as

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon},$$

where $\boldsymbol{\beta} = (\beta_0, \dots, \beta_{p-1})^T$ and $\boldsymbol{\varepsilon} = (\varepsilon_1, \dots, \varepsilon_n)^T$

Error Sum of Squares (SSE) = $\sum_{i=1}^n (Y_i - \beta_0 - \sum_{j=1}^{p-1} \beta_j X_j)^2$
can be expressed in Matrix notation as:

$$(\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})^T (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})$$

Similar to SLR, we will discuss

- Estimation
- Inference
- Diagnostics and Remedies

We will also discuss some new topics

- Model Selection
- Multicollinearity