

Lecture 21

Correlation and Regression Analysis

Readings: IntroStat Chapter 11; OpenIntro Chapter 8

STAT 8010 Statistical Methods I
June 15, 2023

Characterizing the
Relationship Between
Two Numerical
Variables

Regression Analysis

Simple Linear
Regression

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1 Characterizing the Relationship Between Two Numerical Variables

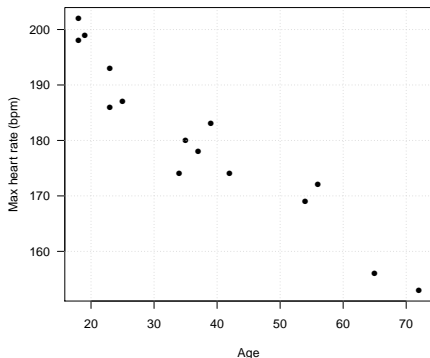
2 Regression Analysis

3 Simple Linear Regression

Motivated Example: Maximum Heart Rate vs. Age

Suppose we have 15 people of varying ages are tested for their maximum heart rate (MHR):

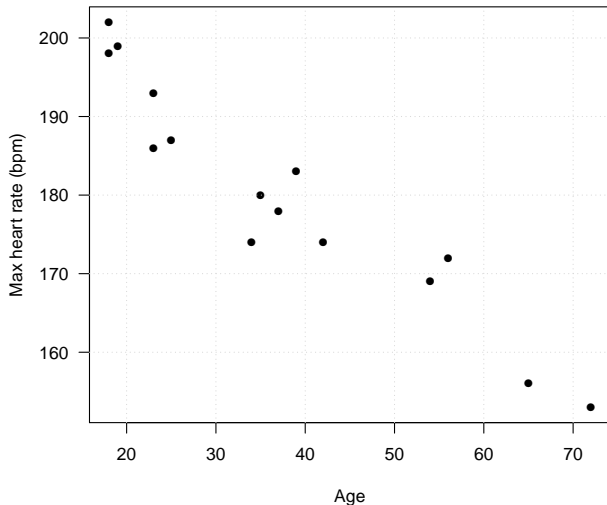
Age	18	23	25	35	65	54	34	56	72	19	23	42	18	39	37
MHR	202	186	187	180	156	169	174	172	153	199	193	174	198	183	178



Question: How to describe the relationship between maximum heart rate and age?

Scatterplot

A **scatterplot** is a useful tool to graphically display the relationship between **two numerical variables**. Each dot on the scatterplot represents one observation from the data



Typical questions we want to ask for a scatterplot:

- the **form** of relationship between two variables e.g. **linear**, **quadratic**, ...

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In the next few slides we will learn how to quantify the **strength** and **direction** of the **linear relationship** between two variables

- **Recall:** **Variance** is a measure of the variability of **one** **quantitative** variable

Variance, Covariance, and Correlation

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- The **sign** of the covariance shows the **direction** in the linear relationship between the variables
- The normalized version of the covariance, the **correlation** shows both the the **direction** and the **strength** of the linear relation

Correlation: Pearson Correlation Coefficient (ρ)

- We use ρ to denote the **population correlation** and r to denote the **sample correlation**
- The value of the correlation is between **-1** and **1**
- The **strength** of the linear relation:
 - If $\rho = 1$ (**-1**): the two variables have a **perfect positive (negative) linear relationship**
 - If $0.7 < |\rho| < 1$: we say the two variables have a **strong linear relationship**
 - If $0.3 < |\rho| < 0.7$: we say the two variables have a **moderate linear relationship**
 - If $0 < |\rho| < 0.3$: we say the two variables have a **weak linear relationship**
 - If $\rho = 0$: we say the two variables have **no linear relationship**

Scatterplot & Pearson Correlation Coefficient

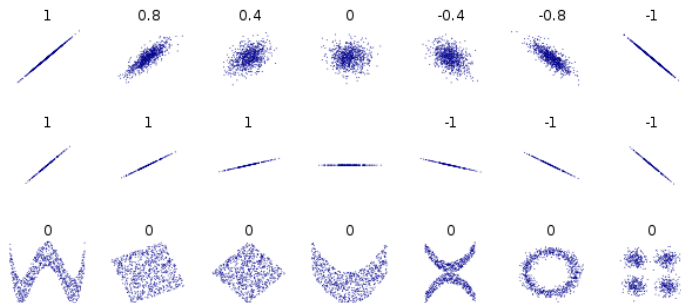


Figure: Image courtesy of Wikipedia at https://en.wikipedia.org/wiki/Correlation_and_dependence

● Recall: Variance

- Sample variance: $s_X^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}$
- Population variance: $\sigma_X^2 = E[(X - \mu_X)^2]$

● Covariance

- Sample covariance: $s_{X,Y} = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{n-1}$
- Population covariance: $\sigma_{X,Y} = E[(X - \mu_X)(Y - \mu_Y)]$

● Correlation

- Sample correlation: $r_{X,Y} = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^n (X_i - \bar{X})^2} \sqrt{\sum_{i=1}^n (Y_i - \bar{Y})^2}}$
or $\frac{s_{X,Y}}{s_X s_Y}$
- Population correlation: $\rho_{X,Y} = \frac{E[(X - \mu_X)(Y - \mu_Y)]}{\sqrt{E[(X - \mu_X)^2]} \sqrt{E[(Y - \mu_Y)^2]}}$
or $\frac{\sigma_{X,Y}}{\sigma_X \sigma_Y}$

A Toy Example

You wonder how sleep affects productivity. You take a sample of 4 of your friends and measure last night's sleep and today's productivity in hours. Here are the results:

Sleep (X)	Productivity (Y)
2	4
4	12
6	14
10	10

Calculate the **means**, **variances**, and **standard deviations** of each variable and the **correlation coefficient** of these two variables

Toy Example Cont'd

Solution.

Let X denote last night's sleep in hours and Y denote today's productivity in hours

$$\bullet \bar{X} = \frac{2+4+6+10}{4} = 5.5, \quad \bar{Y} = \frac{4+12+14+10}{4} = 10$$

$$\bullet s_X^2 = \frac{(2-5.5)^2 + (4-5.5)^2 + (6-5.5)^2 + (10-5.5)^2}{4-1} = \frac{35}{3}$$
$$s_Y^2 = \frac{(4-10)^2 + (12-10)^2 + (14-10)^2 + (10-10)^2}{4-1} = \frac{56}{3}$$

$$\bullet s_X = \sqrt{s_X^2} = \sqrt{\frac{35}{3}}, \quad s_Y = \sqrt{s_Y^2} = \sqrt{\frac{56}{3}}$$

$$\bullet r_{X,Y} = \frac{s_{X,Y}}{s_X s_Y}$$
$$s_{X,Y} = \frac{(2-5.5)(4-10) + (4-5.5)(12-10) + (6-5.5)(14-10) + (10-5.5)(10-10)}{3}$$
$$= \frac{20}{3} \Rightarrow r_{X,Y} = \frac{\frac{20}{3}}{\sqrt{\frac{35}{3}} \sqrt{\frac{56}{3}}} = \frac{20}{\sqrt{35 \times 56}} = 0.4518$$

Inference/Hypothesis Test on ρ

1 $H_0 : \rho = 0$ vs. $H_a : \rho \neq 0$

2 Test statistic: $t^* = r\sqrt{\frac{n-2}{1-r^2}}$

3 Under H_0 : $t^* \sim t_{df=n-2}$

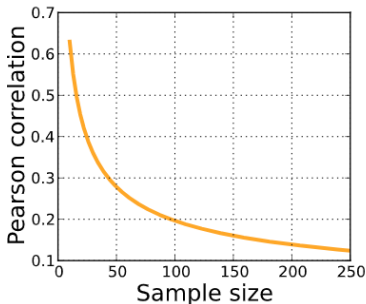
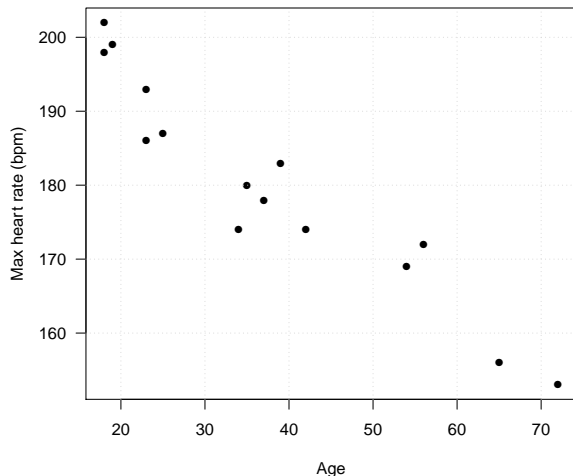


Figure: Image courtesy of Wikipedia

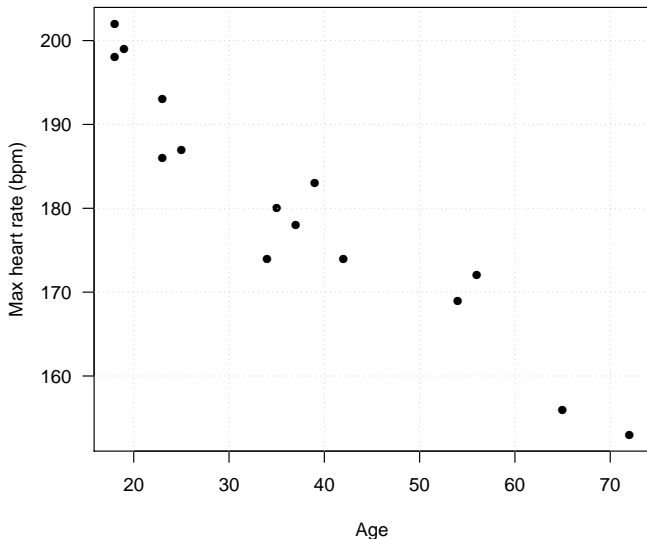
Maximum Heart Rate Example Revisited



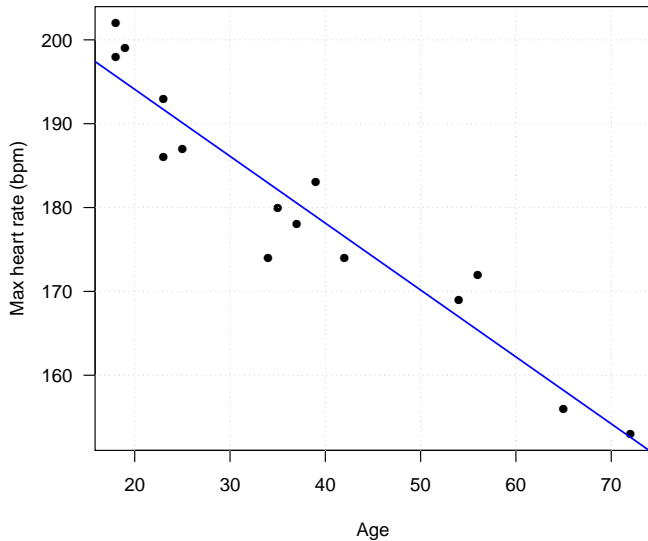
We may want to **predict** maximum heart rate for an individual based on his/her age \Rightarrow **Regression Analysis**

What is Regression Analysis?

Regression analysis: A set of statistical procedures for estimating the relationship between **response variable** and **predictor variable(s)**



Scatterplot: Is Linear Trend Reasonable?



Simple Linear Regression (SLR)

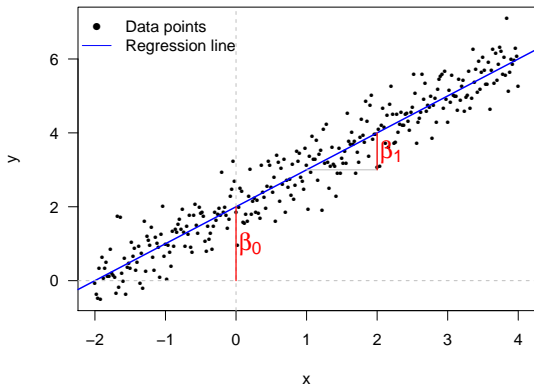
Y : dependent (response) variable; X : independent (predictor) variable

- In SLR we **assume** there is a **linear relationship** between X and Y :

$$Y = \beta_0 + \beta_1 X + \varepsilon$$

- We will need to estimate β_0 (intercept) and β_1 (slope)
- Then we can use the estimated regression equation to
 - make predictions
 - study the relationship between response and predictor
 - control the response

Regression equation: $Y = \beta_0 + \beta_1 X$



- β_0 : $E[Y]$ when $X = 0$
- β_1 : $E[\Delta Y]$ when X increases by 1

Assumptions about the Random Error ε

In order to estimate β_0 and β_1 , we make the following assumptions about ε

- $E[\varepsilon_i] = 0$
- $\text{Var}[\varepsilon_i] = \sigma^2$
- $\text{Cov}[\varepsilon_i, \varepsilon_j] = 0, \quad i \neq j$

Therefore, we have

$$E[Y_i] = \beta_0 + \beta_1 X_i, \text{ and}$$

$$\text{Var}[Y_i] = \sigma^2$$

The regression line $\beta_0 + \beta_1 X$ represents the **conditional expectation curve** whereas σ^2 measures the magnitude of the **variation** around the regression curve

Summary

In this lecture, we learned

- Scatterplot, Covariance, Correlation
- Regression Analysis and Simple Linear Regression

Next lecture we will learn how to estimate the regression parameters β_0, β_1 and how to quantify the estimation uncertainty