

Lesson 10:

Simple Regression Analysis

Victoria Blanes-Vidal

Manuella Lech Cantuaria

The Maersk Mc-Kinney Møller Institute

Applied AI and Data Science

Lesson	Week	Date	TOPICS	Teacher
1	35	1/Sep	Introduction to the course Descriptive statistics –Part I	MLC
2	36	8/sep	Descriptive statistics –Part II	MLC
3	37	15/Sep	Probability distributions	MLC
4	38	22/Sep	Hypothesis testing (one sample)	VBV
5	39	29/Sep	Hypothesis testing (two samples)	VBV
6	40	6/Oct	ANOVA one-way	VBV
7	41	13/Oct	R class (Introduction to R and descriptive statistics) Point giving activity	MLC
-	42	20/Oct	NO CLASS (Autum holidays)	
8	43	27/Oct	R class (hypothesis testing + ANOVA)	MLC
9	44	3/Nov	ANOVA two-way	VBV
-	45	10/Nov	NO CLASS	
10	46	17/Nov	Simple regression analysis	VBV
11	47	24/Nov	Experimental design Point giving activity	VBV
12	48	1/Dec	Multiple regression analysis and questions	VBV+MLC

VBV = Victoria Blanes-Vidal

MLC = Manuella Lech Cantuaria

Lesson 10

1. **Regression analysis**
2. Scatter plot
3. Correlation coefficient
4. Statistical significance of the correlation coefficient
5. Correlation vs. Causation
6. Determining the regression line
7. Plotting the regression line
8. Prediction
9. Coefficient of determination
10. In-class activities



Regression analysis

Regression Analysis is a statistical method used to describe how one (or more) quantitative variable is related to another quantitative variable.

In regression, there are **two types of variables**:

Independent variable/s, also called explanatory variable/s or predictor variable/s (input): x

Dependent variable, also called a response variable (output): y

$x \longleftrightarrow y$

Simple regression analysis

$x_1, x_2, x_3, x_4, \dots \longleftrightarrow y$

Multiple regression analysis

Example: Basketball players



We want to study the relation between weight and height of NBA basketball players.

We randomly select 10 players and measure their height and weight.

Example: Basketball players

```
> Basketball_players <-  
read.table("C:/Users/vbv/Desktop/My_documents/Teaching/Teaching/Statistical_Data_Analysis/2022_2023/Exercises_in_R/Basketball_players.  
txt", header=TRUE)
```

```
> Basketball_players
```

	X_Player_height	Y_Player_weight
1	190.05	88.00
2	190.72	88.89
3	191.48	90.11
4	194.21	91.96
5	199.63	97.02
6	201.76	102.93
7	202.99	103.72
8	204.75	105.27
9	209.45	111.61
10	210.60	112.46



Lesson 10

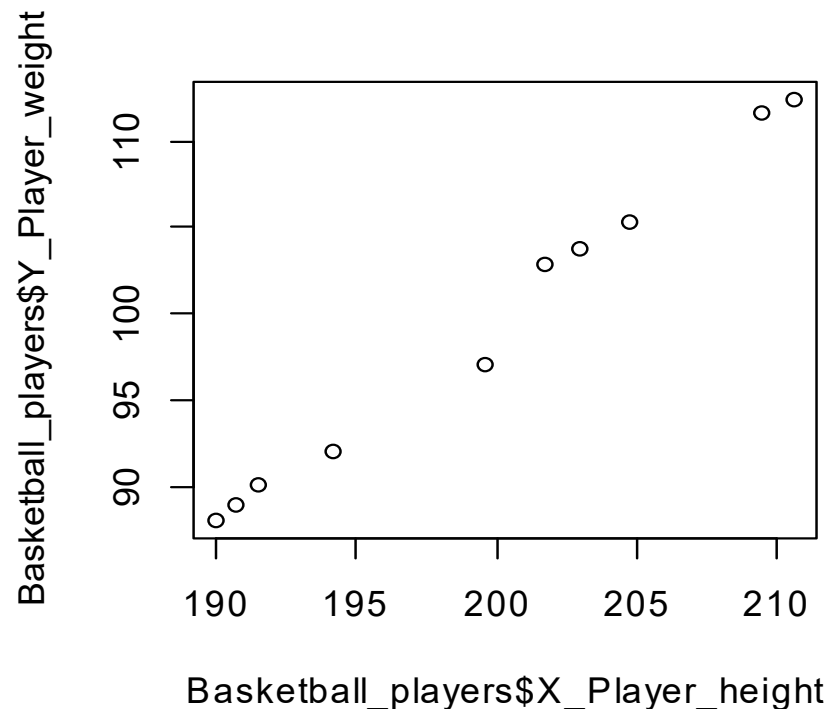
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10. **In-class activities**



Scatter plot

- The relation between two quantitative variables can be visualized by plotting a ***scatter plot***.
- A scatter plot is a graph of the x and y pairs.

```
> plot(Basketball_players$X_Player_height, Basketball_players$Y_Player_weight)
```



- Usually, the variable x is plotted on the horizontal axis, and the variable y is plotted on the vertical axis.

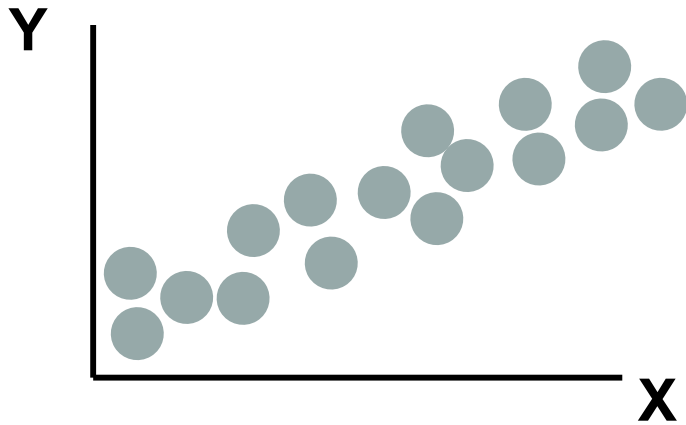
Scatter plot

- The scatter plot is a visual way to describe the nature of the relationship between the two variables.
- A scatterplot displays the:
 - Form: Linear vs. non-linear
 - Direction: Positive vs. Negative
 - Strength: Weak vs. Strong

of the relationship between two variables (x and y).

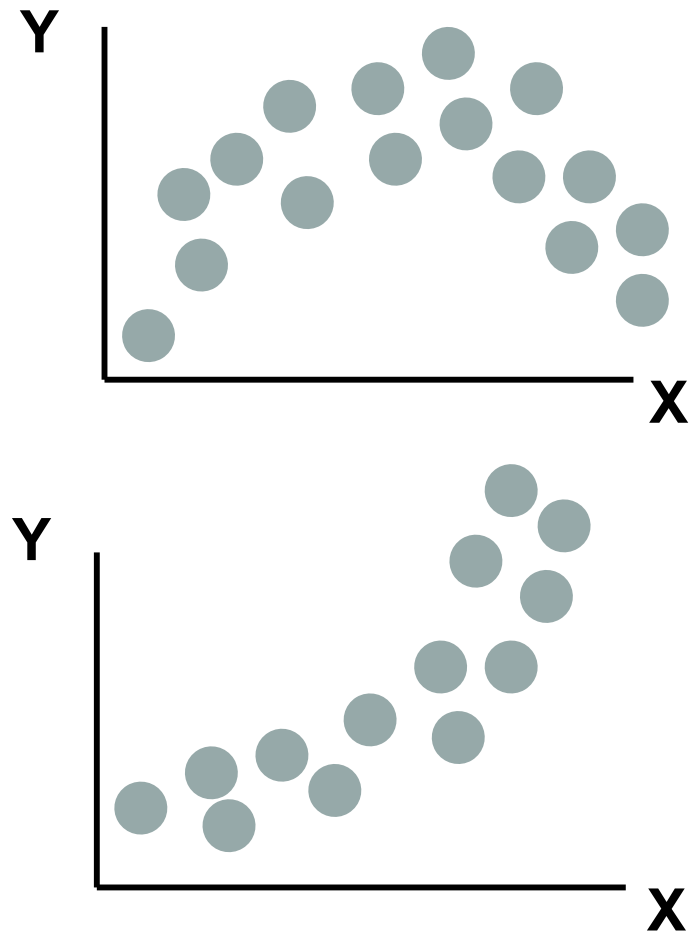
Form

Linear



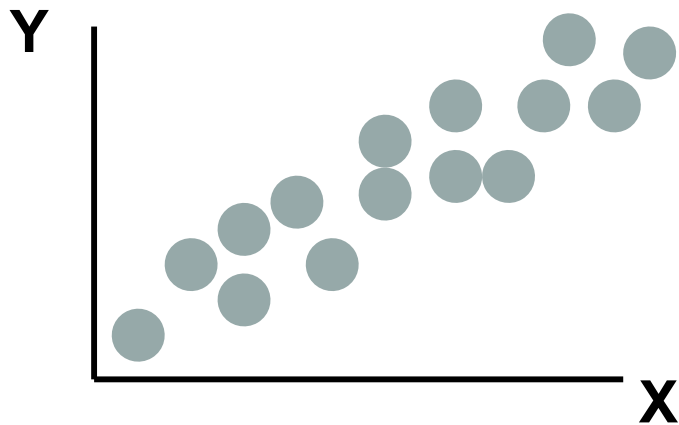
When the relationship between X and Y is linear, the scatter plot gives a **straight line**.

Non-linear



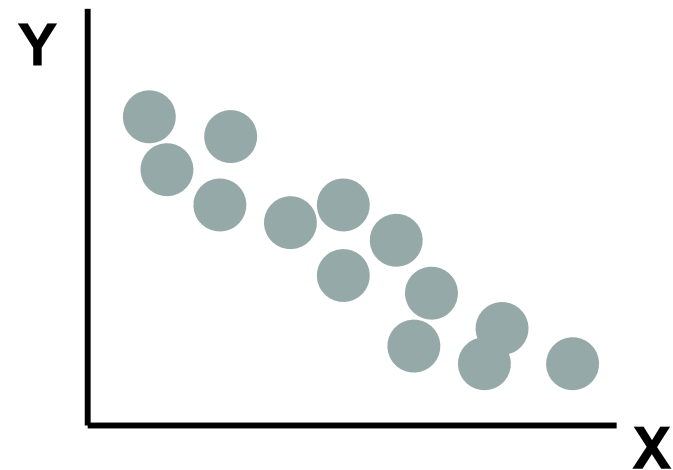
Direction

Positive



If there is a positive linear relationship,
as **X increases**, **Y increases**.

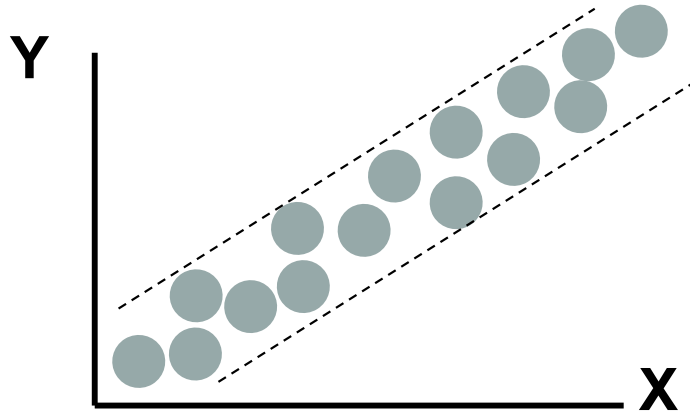
Negative



If there is a negative linear relationship,
as **X increases**, **Y decreases**.

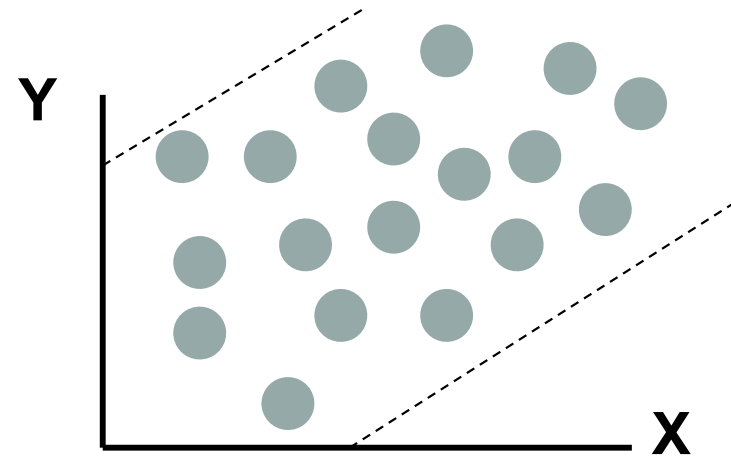
Strength

Strong

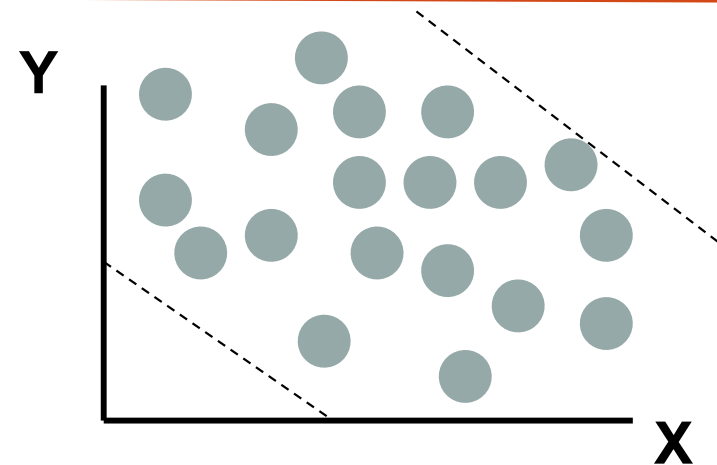
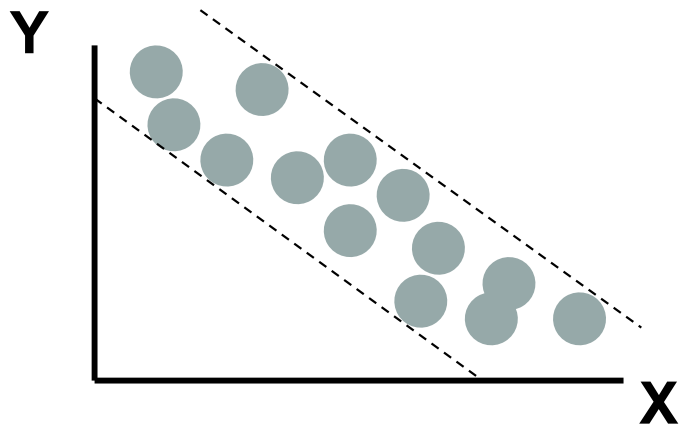


If there is a strong linear positive relationship, when the value of X increases, the value of Y increases in a reliable manner.

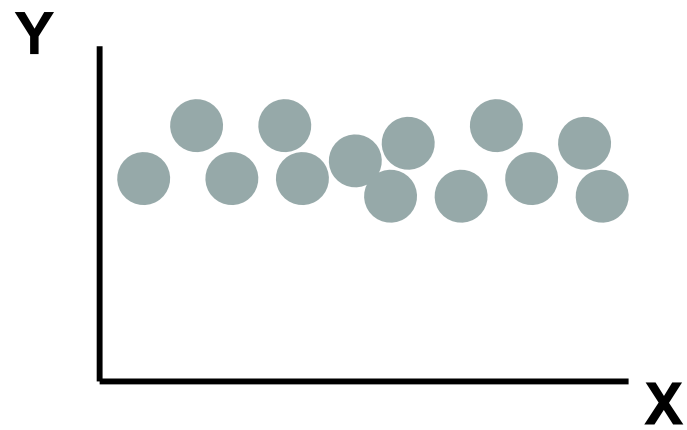
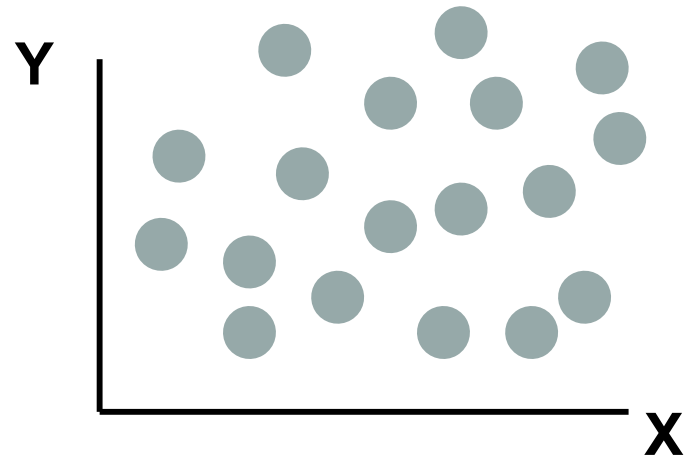
Weak



If there is a weak linear positive relationship, when the value of X increases, the other variable tends to increase as well, but in a weak or unreliable manner.

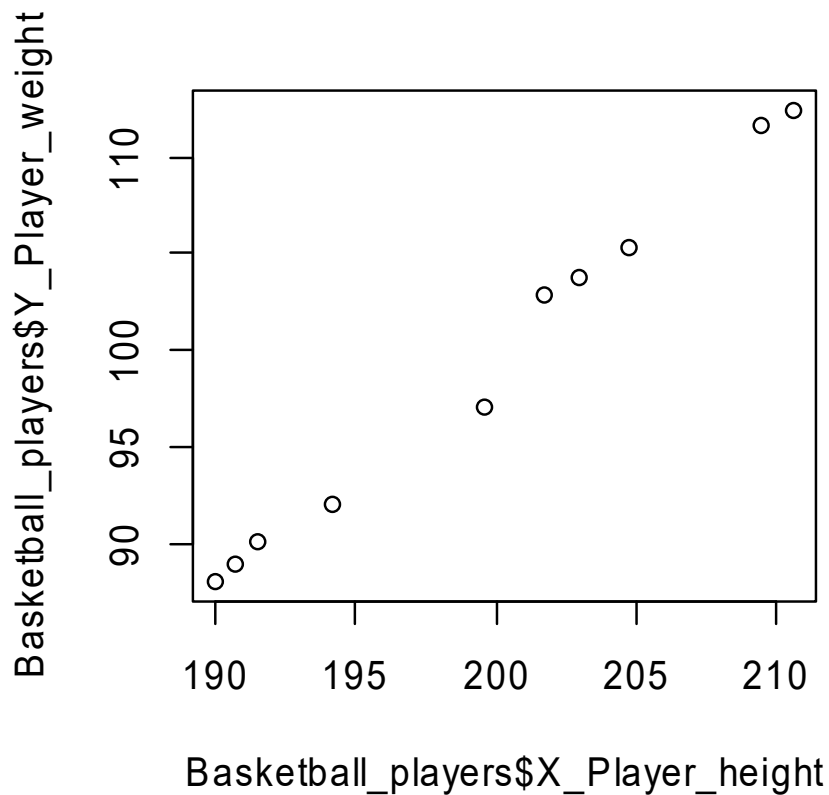


No relationship



Example: Basketball players

```
> plot(Basketball_players$X_Player_height, Basketball_players$Y_Player_weight)
```



What is the form, direction and strength of the relationship between basketball players height and weight?

Form

Linear

Non-linear

Direction

Positive

Negative

Strength

Weak

Strong

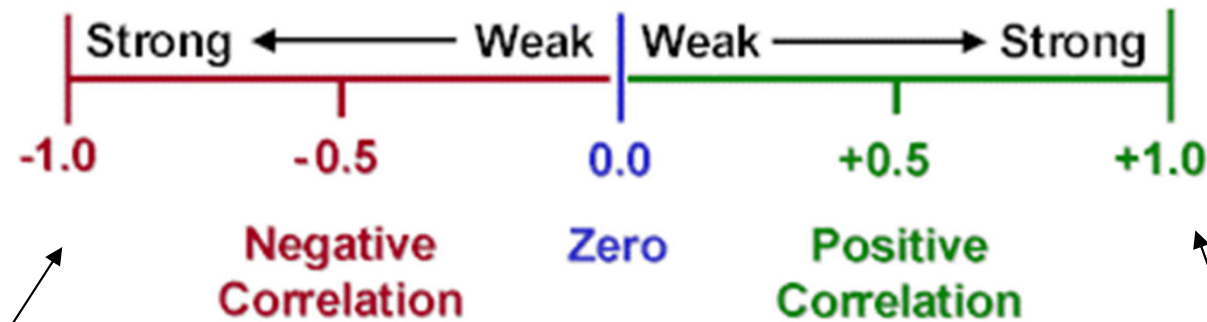
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Correlation coefficient

- The **correlation coefficient (r)** measures the strength and direction of a linear relationship between two variables.
- The range of the correlation coefficient (r) is **from -1 to 1**.

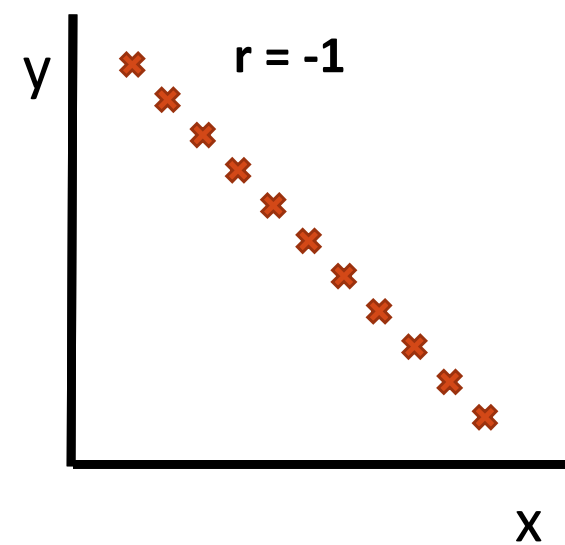
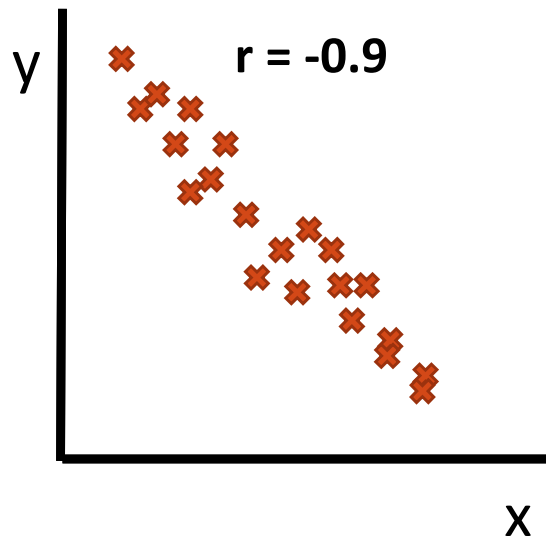
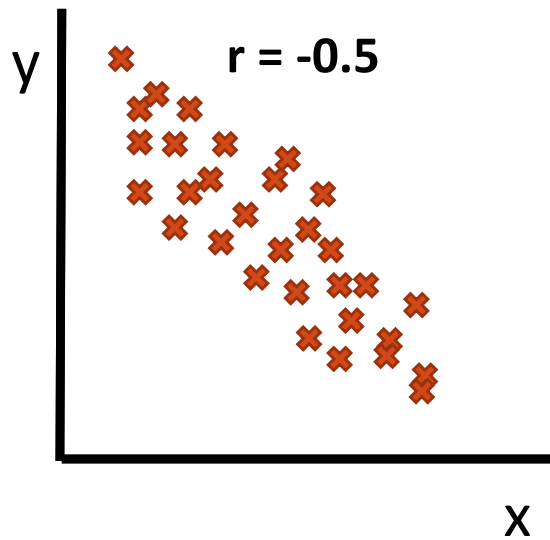
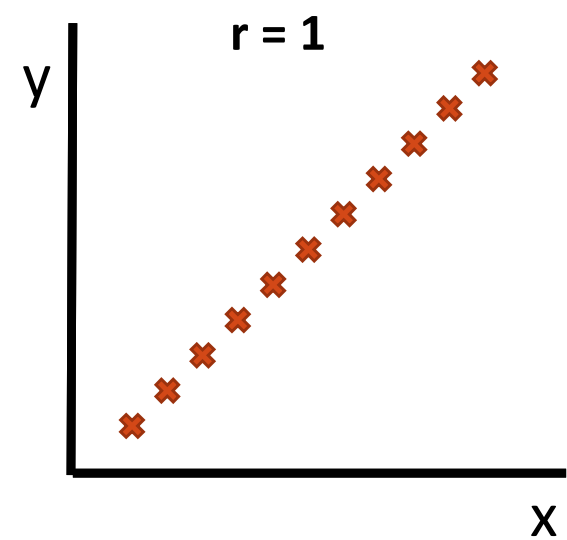
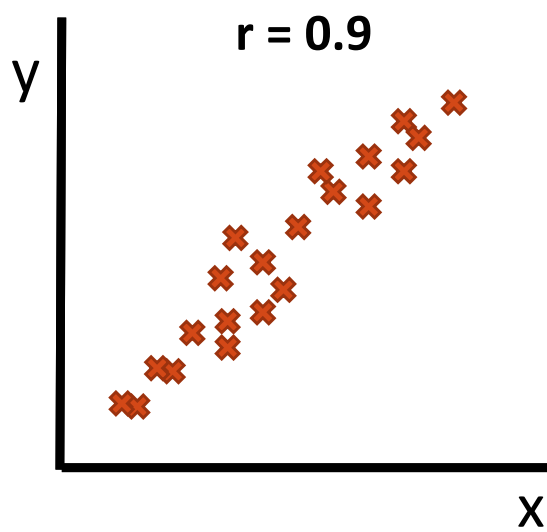
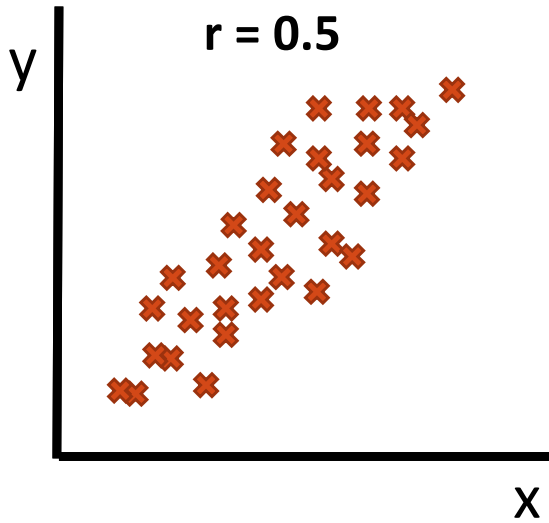


If there is a strong negative linear relationship between the variables, the value of r will be close to -1.

When there is no linear relationship between the variables or only a very weak relationship, the value of r will be close to 0.

If there is a strong positive linear relationship between the variables, the value of r will be close to 1.

Scatter plots and correlation coefficients



Example: Basketball players

```
> cor(Basketball_players$X_Player_height,Basketball_players$Y_Player_weight)
[1] 0.9948415
```

Based on the correlation coefficient ($r = 0.994$), the direction and strength of the relationship between basketball players height and weight is...

Direction

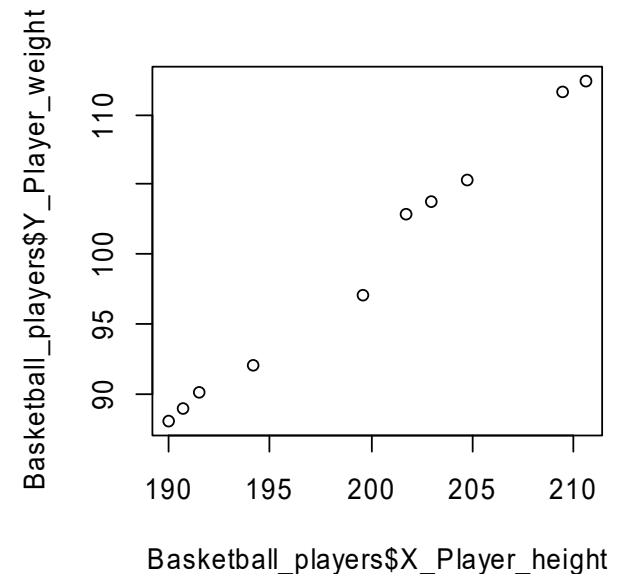
Positive

Negative

Strength

Weak

Strong



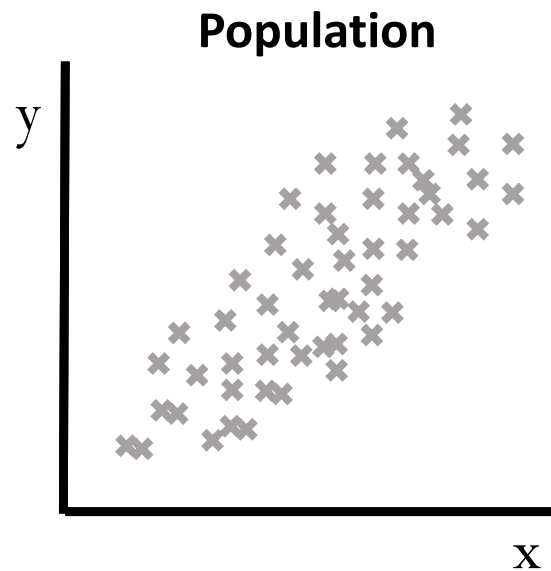
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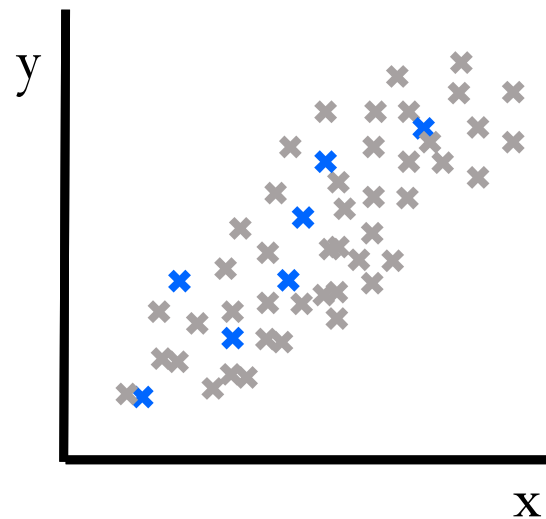


Statistical Significance of the Correlation Coefficient

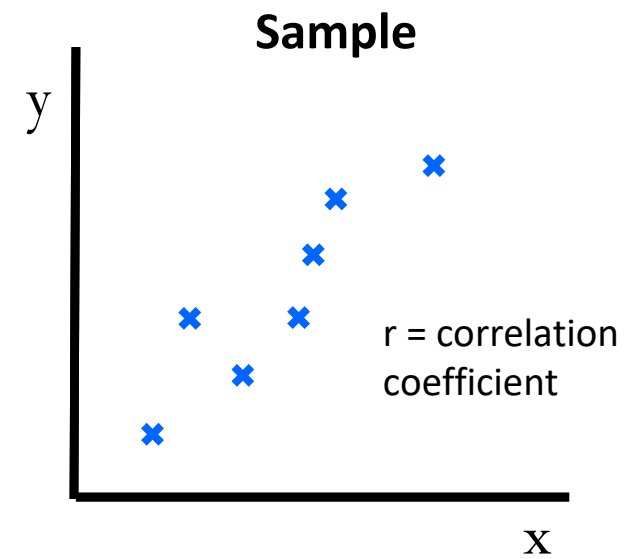
- What we would like to know is whether there is a relation between X and Y in the population (not in the sample).
- However, the correlation coefficient (r) is calculated from data obtained from a sample (not from the population).



Out of all members of the population...



We have selected (sampled) only some of them, and measured X and Y only for those



We have calculated the correlation coefficient (r) using the X and Y measured in that sample (7 points)

Statistical Significance of the Correlation Coefficient

If the correlation coefficient in the sample (r) shows that there is a relationship between X and Y in the sample:

Does that mean that there is a relation between X and Y in the population?

We need to use a hypothesis test, where:

Null hypothesis (H_0):

There is not a relation between X and Y in the population

Alternative hypothesis (H_1):

There is a relation between X and Y in the population

Statistical Significance of the Correlation Coefficient



```
> Basketball_players
```

	X_Player_height	Y_Player_weight
1	190.05	88.00
2	190.72	88.89
3	191.48	90.11
4	194.21	91.96
5	199.63	97.02
6	201.76	102.93
7	202.99	103.72
8	204.75	105.27
9	209.45	111.61
10	210.60	112.46

```
> cor.test(Basketball_players$X_Player_height, Basketball_players$Y_Player_weight)
```

Pearson's product-moment correlation

data: Basketball_players\$X Player height and Basketball_players\$Y_Player_weight
 t = 27.738, df = 8, **p-value = 3.079e-09**
 alternative hypothesis: true correlation is not equal to 0
 95 percent confidence interval:
 0.9775002 0.9988253
 sample estimates:

cor
0.9948415

This is the correlation coefficient between height and weight in the sample.

This is the p-value of the statistical significance of the correlation coefficient.

In this case, the hypothesis test shows that there is a significant relation between X and Y in the population (p-value<0.05).

Note that the strength and the significance of the relationship between X and Y, are two different things:

A relationship can be **strong** for the sample (r close to 1 or -1), and yet **not significant** for the population, because the result of the statistical test **depends also on the sample size**.

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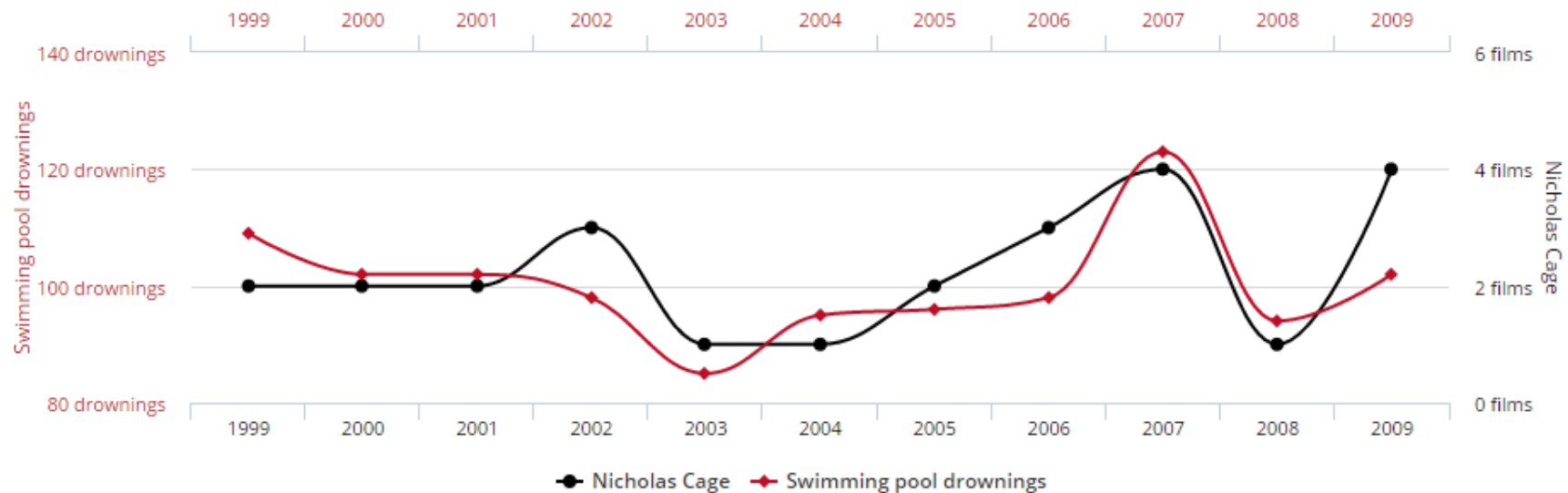
Correlation vs. causation

When analyzing data, many people confuse the concepts of correlation and causation.

- **Correlation** - When researchers find a correlation, which can also be called an association, what they are saying is that they found a **significant relationship between two variables**.
- **Causation** - When an article says that causation was found, this means that the researchers found that **changes in one variable they measured, *directly caused* changes in the other**.

Number of people who drowned by falling into a pool correlates with Films Nicolas Cage appeared in

Correlation: 66.6% ($r=0.666004$)



tylervigen.com

Data sources: Centers for Disease Control & Prevention and Internet Movie Database

```
> Nicolas_cage_drowning <-  
read.table("C:/Users/vbv/Desktop/My_documents/Teaching/Teaching/Statistical_Data_Analysis/2022_2023/Exercises_in_R/Nicolas_cage.txt", header=TRUE)
```

```
> Nicolas_cage_drowning  
  Nicolas_cage_films People_drowned  
1                   2             110  
2                   2             103  
3                   2             101  
4                   3              98  
5                   1              85  
6                   1              93  
7                   2              95  
8                   3              98  
9                   4             122  
10                  1              95  
11                  4             102
```

```
cor.test(Nicolas_cage_drowning$Nicolas_cage_films,Nicolas_cage_drowning$People_drowned)
```

Pearson's product-moment correlation

data: Nicolas_cage_drowning\$Nicolas_cage_films and
Nicolas_cage_drowning\$People_drowned

t = 2.6502, df = 9, p-value = 0.02647

alternative hypothesis: true correlation is not equal to 0

95 percent confidence interval:

0.1031497 0.9032183

sample estimates:

cor
0.6620586

There is a significant positive relation between Nicolas Cage films and People drowned in swimming pools (p-value<0.05).

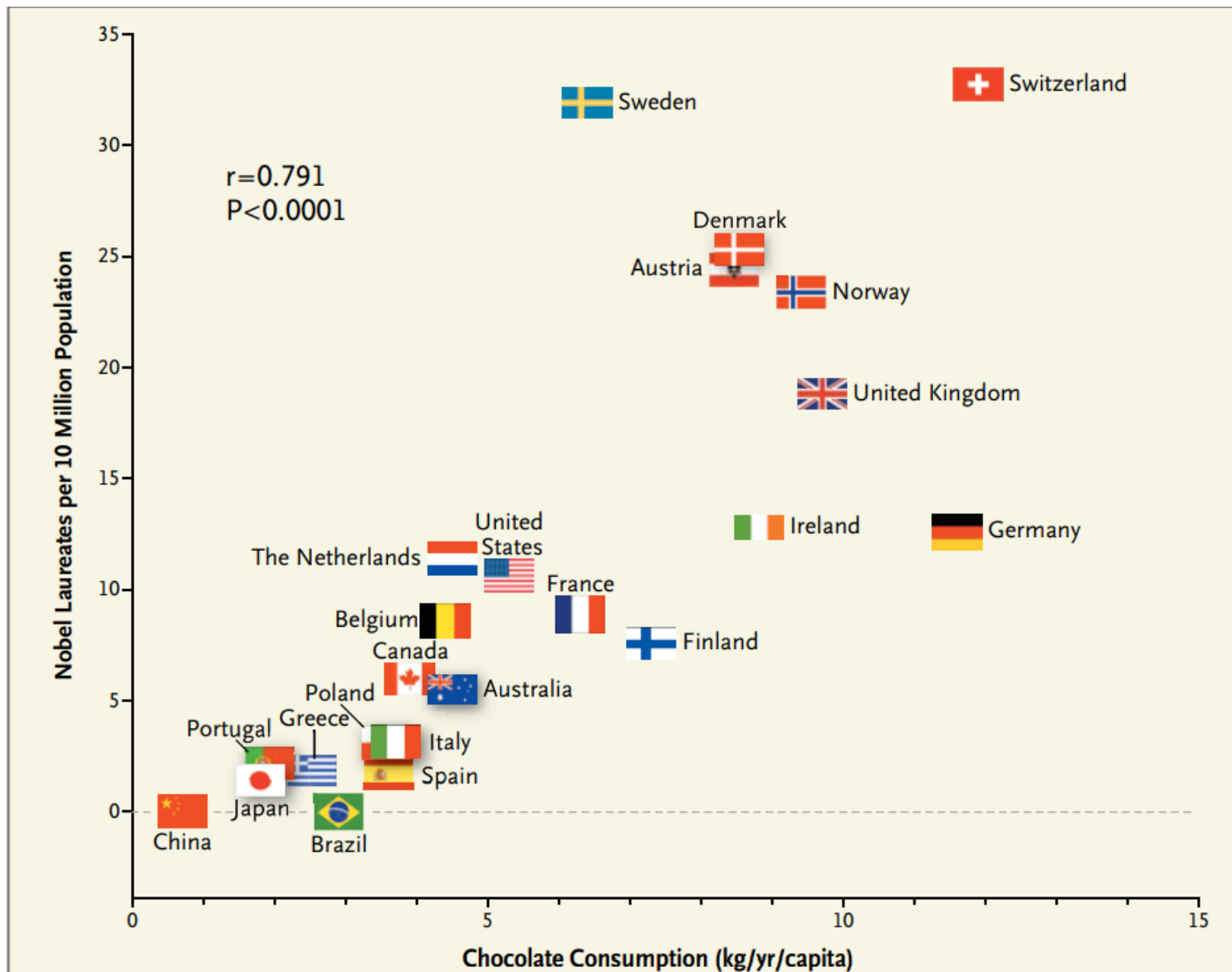


Figure 1. Correlation between Countries' Annual Per Capita Chocolate Consumption and the Number of Nobel Laureates per 10 Million Population.

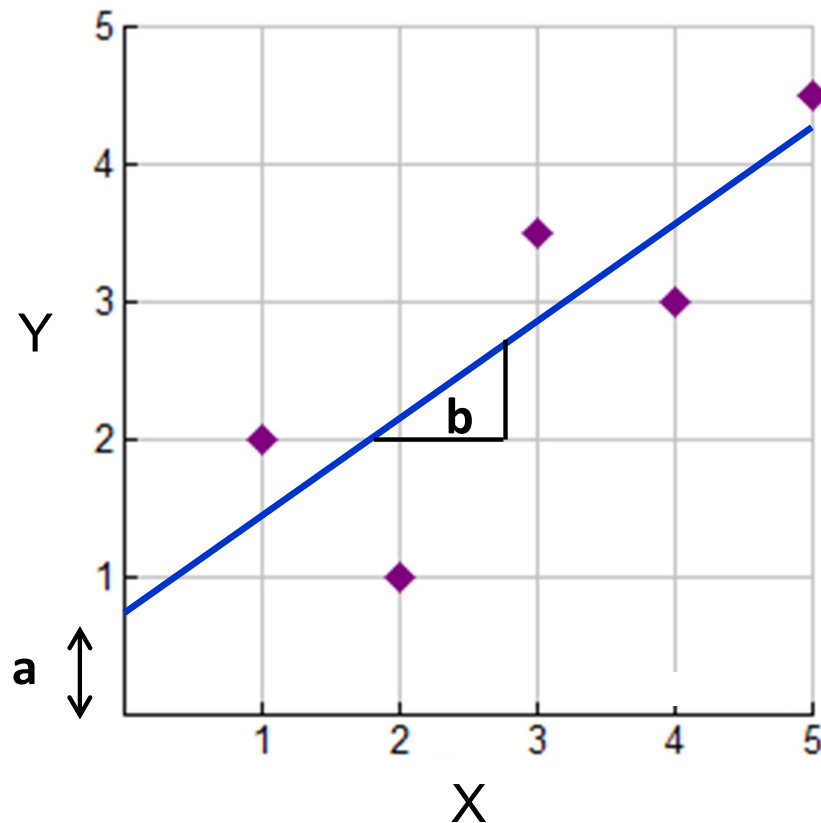
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Determining the regression line

- If the value of the correlation coefficient is significant, the next step is to determine the equation of the **regression line**.



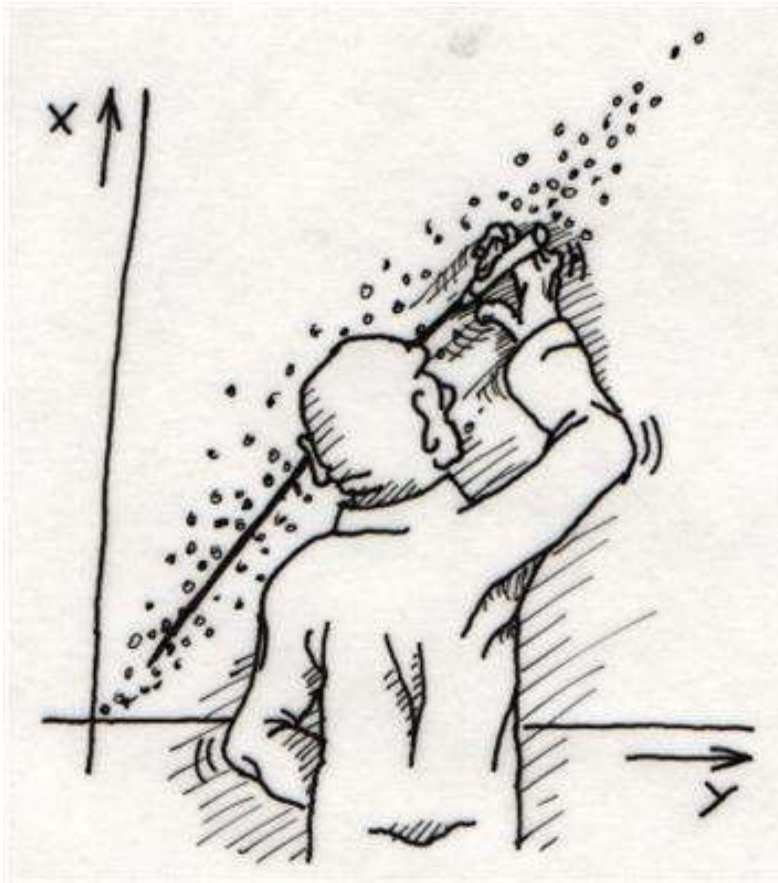
The equation of the regression line is:

$$Y = a + b * X$$

a = intercept

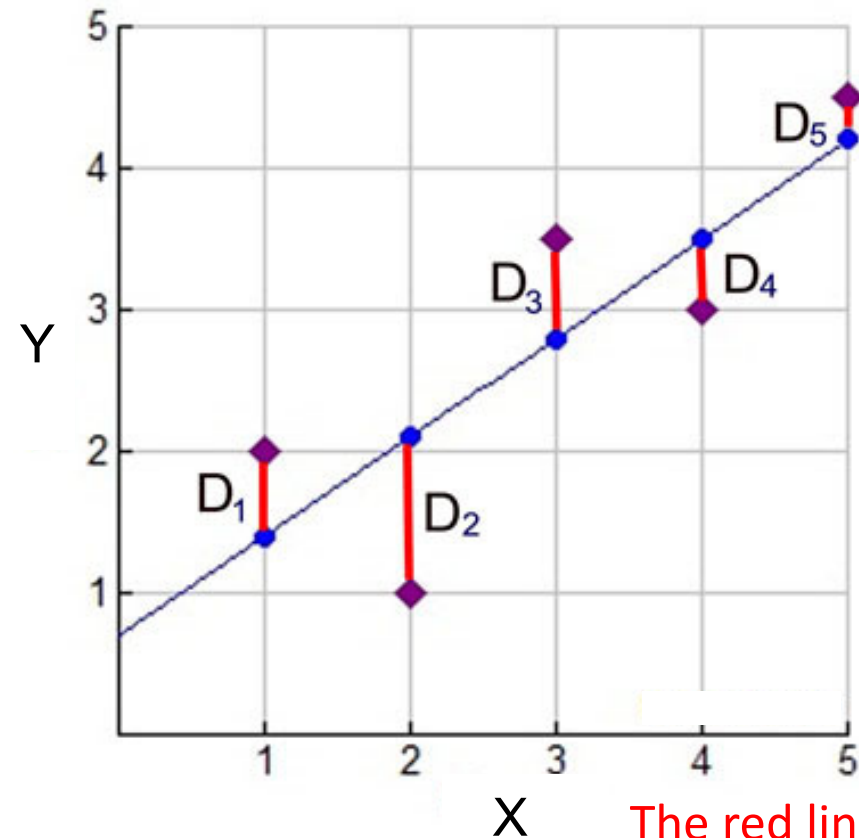
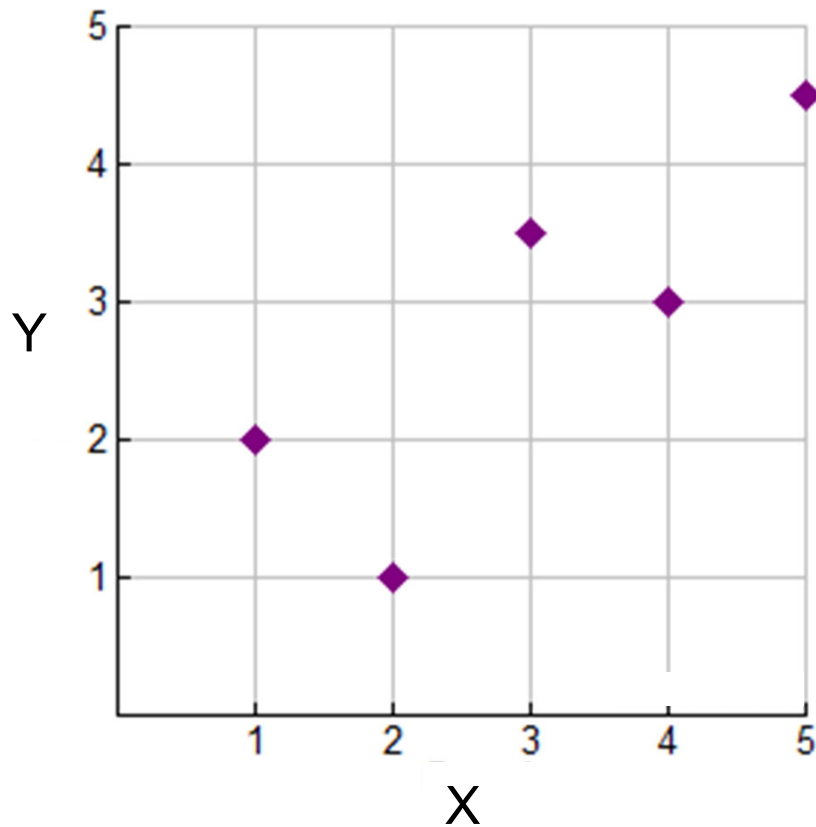
b = slope

Determining the regression line



Out of all the possible straight lines we could draw...

which one is the regression line?



The red line segments are called **residuals**.

The regression line is:
the line with the minimum sum of the squares of the residuals.

$$\min \left(\sum D_n^2 \right)$$

In the example: $\min(D_1^2 + D_2^2 + D_3^2 + D_4^2 + D_5^2)$

Determining the regression line

```
> Regression_Basketball <- lm(Y_Player_weight~X_Player_height,  
data=Basketball_players)  
> summary(Regression_Basketball)
```

Call:

```
lm(formula = Y_Player_weight ~ X_Player_height, data =  
Basketball_players)
```

Residuals:

Min	1Q	Median	3Q	Max
-2.2566	-0.1474	0.3176	0.4659	1.0847

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-141.47114	8.68205	-16.30	2.03e-07 ***
X_Player_height	1.20597	0.04348	27.74	3.08e-09 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.9969 on 8 degrees of freedom

Multiple R-squared: 0.9897, Adjusted R-squared: 0.9884

F-statistic: 769.4 on 1 and 8 DF, p-value: 3.079e-09



$$Y = a + b * X$$

$$a = -141.5$$

$$b = 1.206$$

$$Y = -141.5 + 1.206 * X$$

$$\text{Weight} = -141.5 + 1.206 * \text{Height}$$

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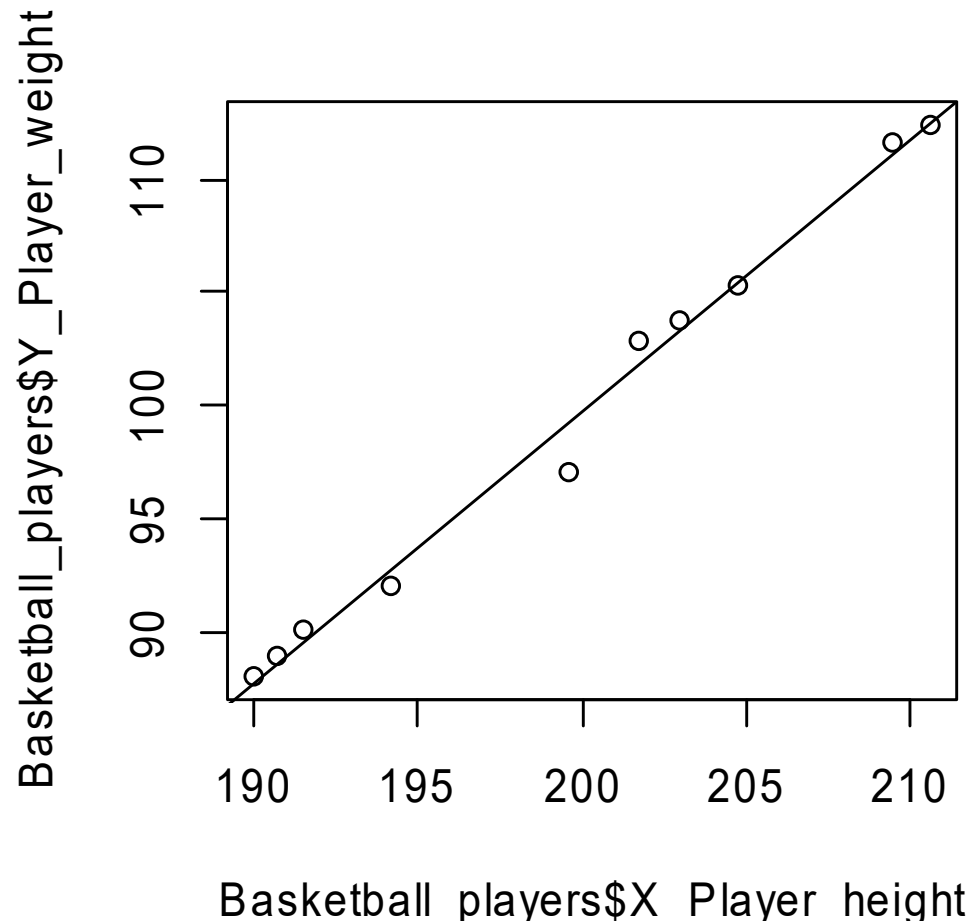


Plotting the regression line

Example: Basketball players



```
> plot(Basketball_players$X_Player_height,Basketball_players$Y_Player_weight)  
> abline(Regression_Basketball)
```



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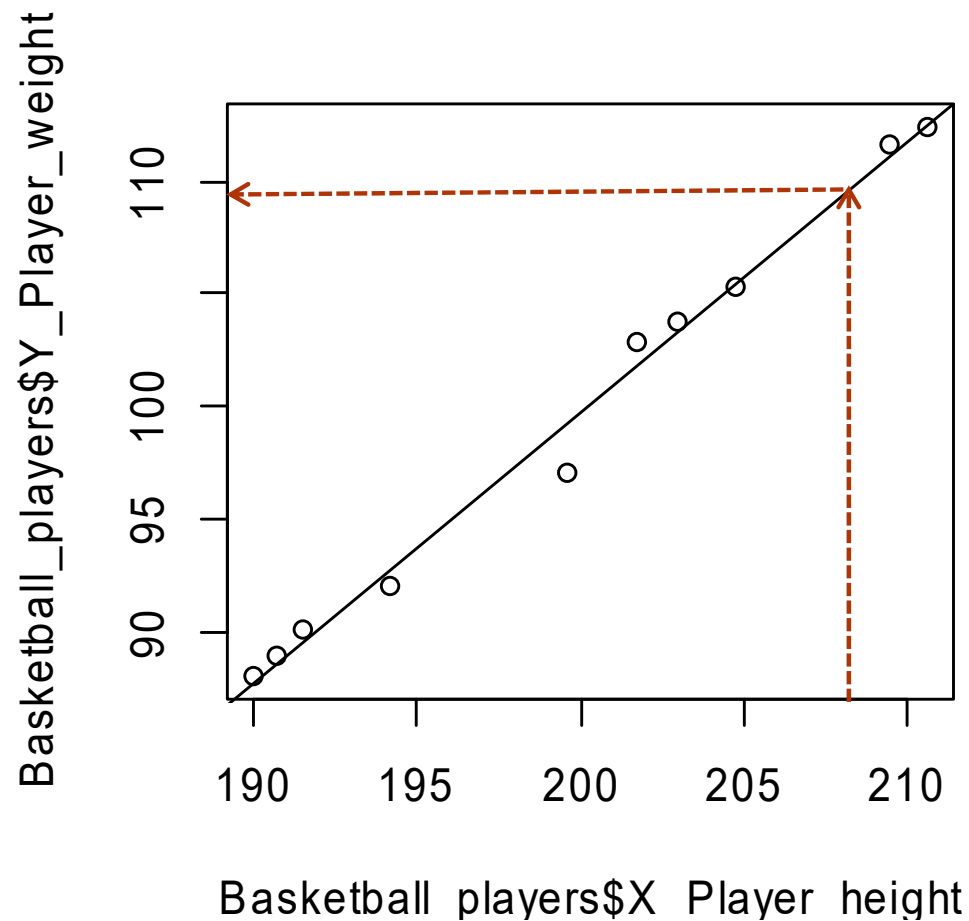
Prediction

The equation of the **regression line** can be used for **prediction**.

Example: Basketball players

We measure the height of a new basketball player: Height = 208 cm

What is the predicted weight?



The equation of the **regression line** can be used for **prediction**.

Example: Basketball players

We measure the height of a new basketball player: Height = 208 cm

What is the predicted weight?



$$\text{Weight} = -141.5 + 1.206 * \text{Height}$$

$$\text{Weight} = -141.5 + 1.206 * 208$$

$$\text{Weight} = 109.35 \text{ kg}$$

```
> predict(Regression_Basketball, data.frame(X_Player_height=208))  
      1  
109.3706
```

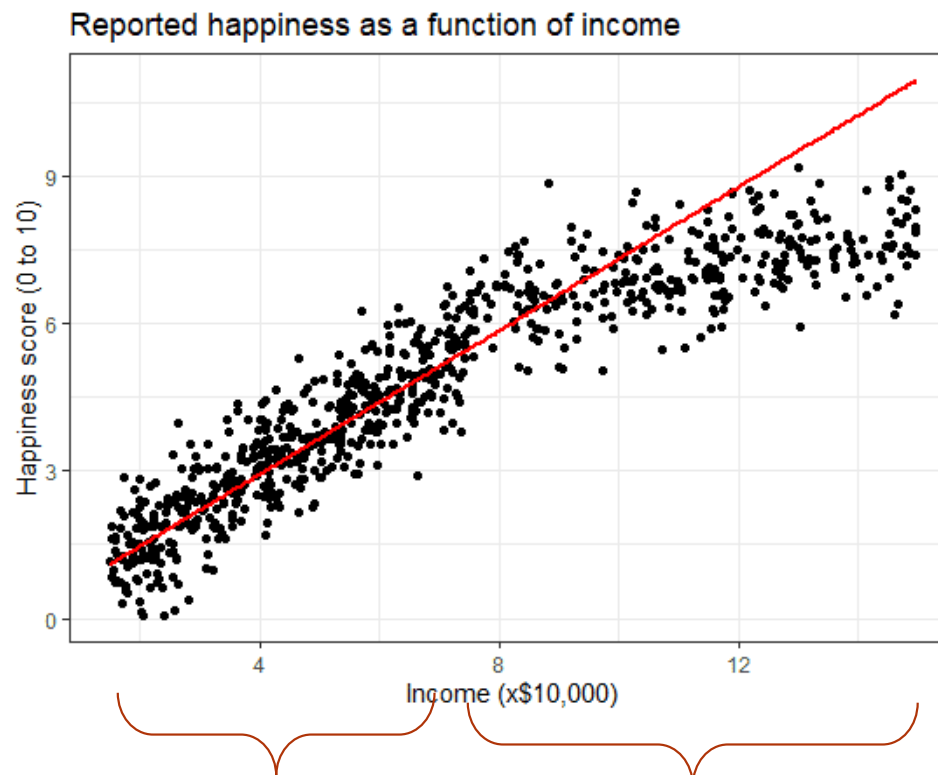
Prediction

Can you predict values outside the range of your data?

No.

We said that the regression equation can be used to **predict** the value of the dependent variable at certain values of the independent variable.

However, this is only true for the range of values where we have actually measured.



The regression line was calculated using these values

It cannot be used to predict Y for this range of values of X

Lesson 10

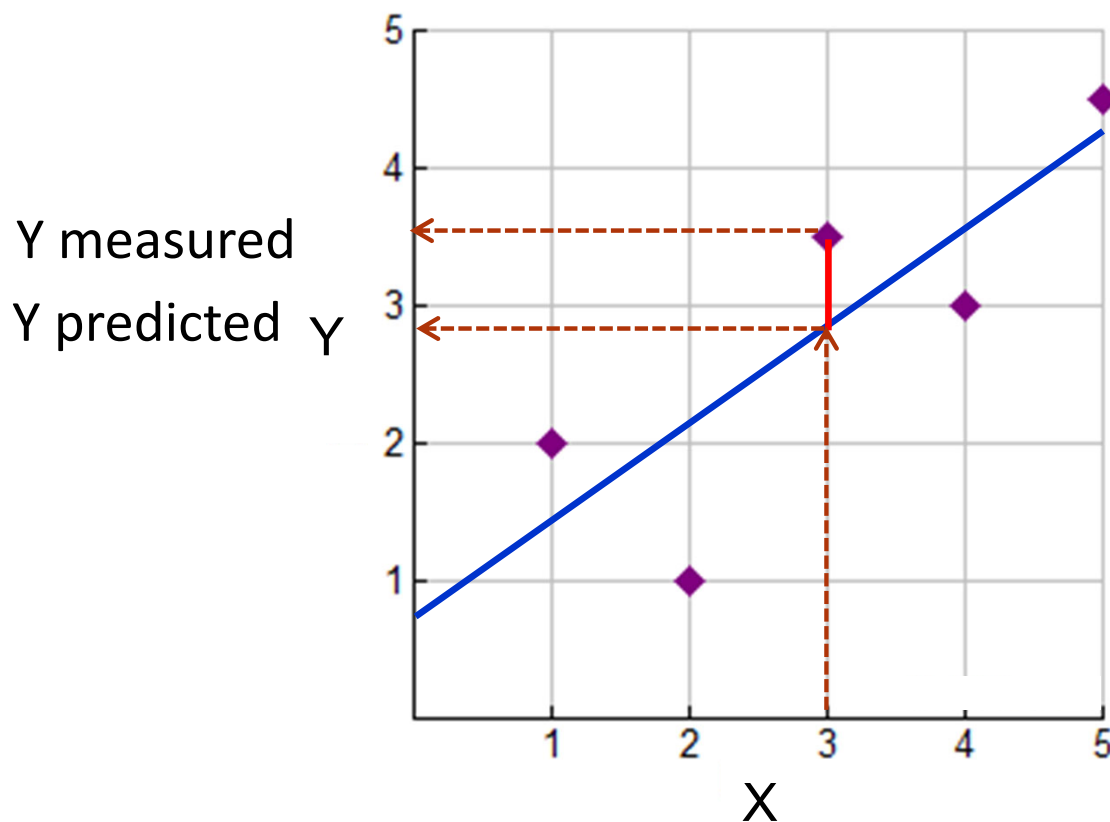
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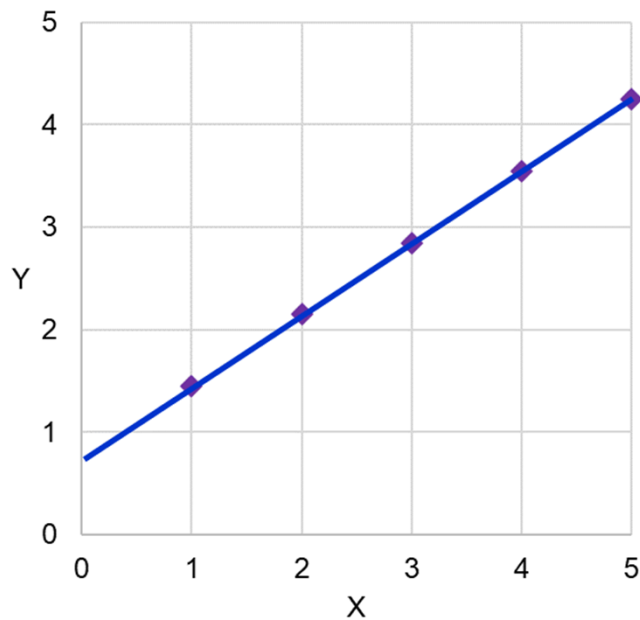
Coefficient of determination (R^2)

There is a difference between the location of a specific point in the scatter plot (Y measured), and what the regression equation predicts (Y predicted).

This is called **residual**.



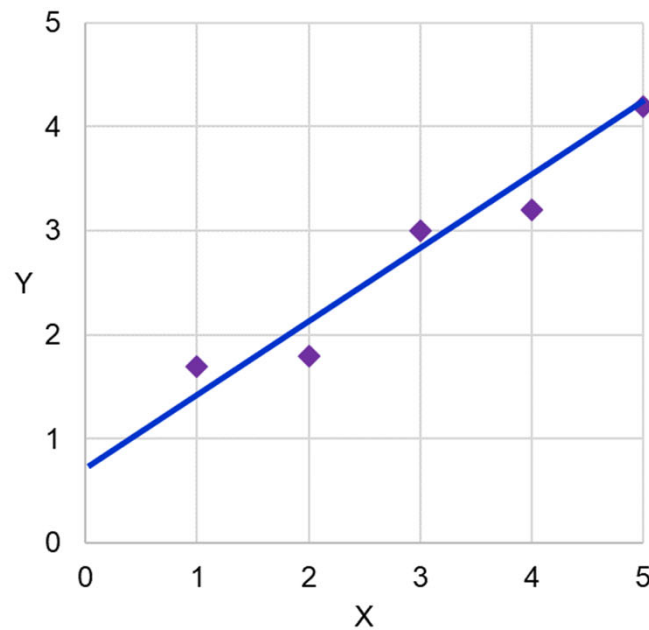
The size of these **residuals** provide information about the existence of other variables (variables other than X), that also affect the Y (and that have not been included in the regression analysis).



If the residuals are zero,
that means that:

"X explains ALL the variation
we see in Y"

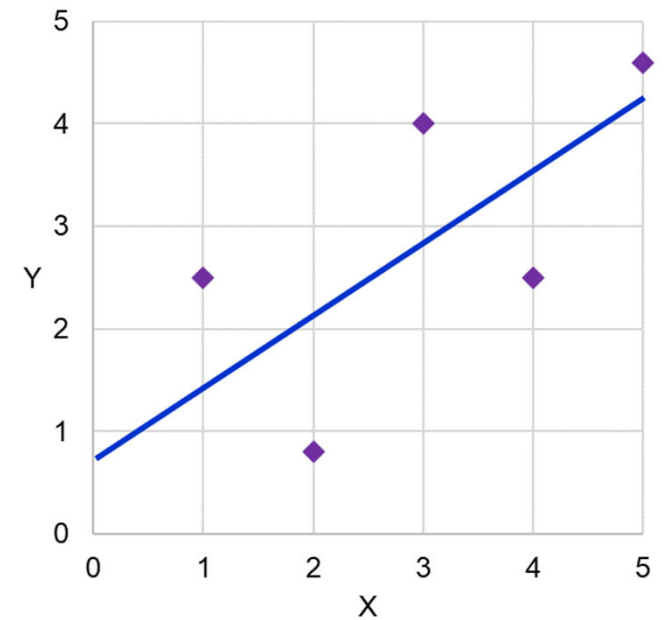
X perfectly predicts Y



If the residuals are small,
that means that:

"X explains almost all the
variation we see in Y"

X is good at predicting Y, but
it is not perfect
(because there are other
variables, other than X, that
also affect Y and these other
variables are not included in
this analysis)

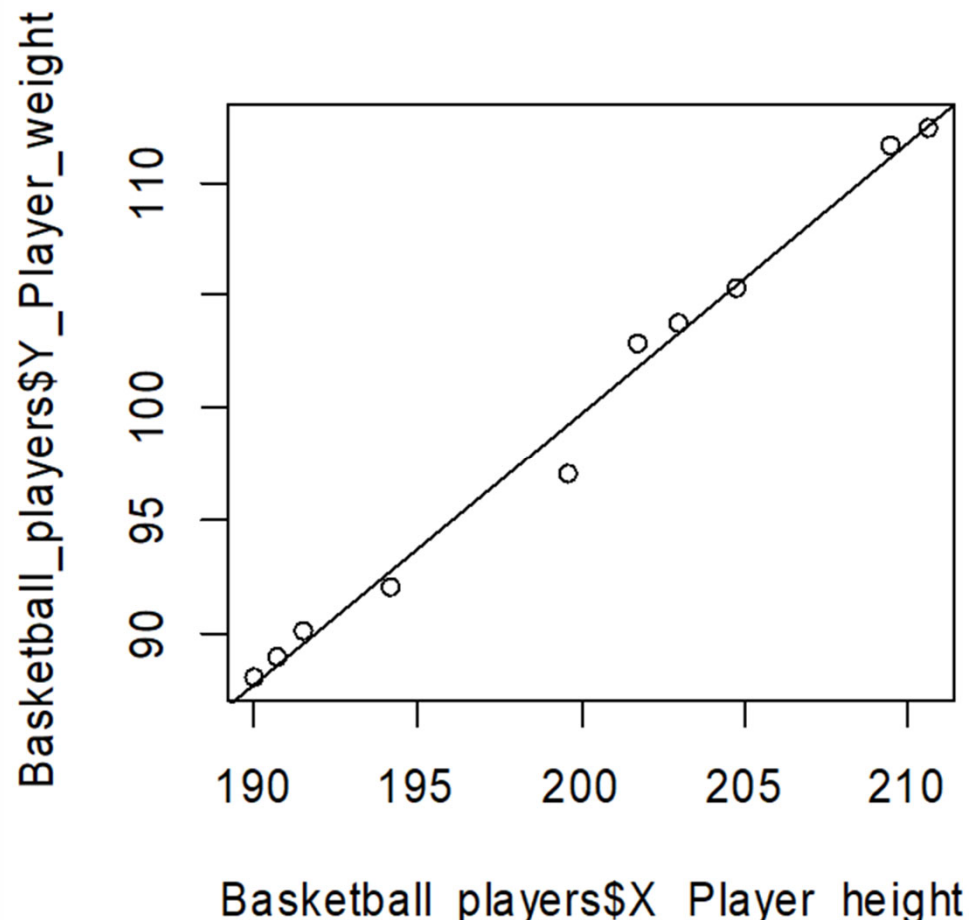


If the residuals are large,
that means that:

"X explains very little of the
variation we see in Y".

X is not good at predicting Y

Example: Basketball players



The residuals are small, that means that:

- The weight of basketball players (Y) can be almost fully explained by their height (X).
- The variable height can predict very well the weight of the player, but the prediction is not perfect (there are some residuals), because:

There are other variables (variables other than height), that also affect weight, for example: diet, exercise, genetics,...

(and these other variables are not included in the simple regression analysis).

Coefficient of determination (R^2)

- The coefficient of determination (R^2) is the ratio of the explained variation to the total variation.

$$R^2 = \frac{\text{Explained variation}}{\text{Total variation}}$$

Coefficient of determination (R^2)

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Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.9969 on 8 degrees of freedom

Multiple R-squared: 0.9897, Adjusted R-squared: 0.9884

F-statistic: 769.4 on 1 and 8 DF, p-value: 3.079e-09

Coefficient of determination (R^2)

The coefficient of determination is: $R^2 = 0.9897$

The coefficient of determination (R^2) is the ratio of the explained variation to the total variation.

Interpreting R^2 :

If $R^2 = 0.9897$, that means that:

98.97% of the variation of weight of basketball players can be explained by their height.

The remaining 1.03% is unexplained; that is, it is not explained by X (height). It could be explained by other variables such as diet, exercise, genetics,...

Coefficient of determination (R^2)

Another way to obtain the value of the coefficient of determination R^2 is to square the correlation coefficient (r).

$$(\text{Correlation coefficient})^2 = \text{Coefficient of determination}$$

```
>
cor.test(Basketball_players$X_Player_height, Basketball_players$Y_Player_weight)
Pearson's product-moment correlation

data: Basketball_players$X_Player_height and Basketball_players$Y_Player_weight
t = 27.738, df = 8, p-value = 3.079e-09
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 0.9775002 0.9988253
sample estimates:
```

cor
0.9948415

$$0.99484^2 = 0.9897$$

The coefficient of determination is: $R^2 = 0.9897$

Lesson 10

1. Regression analysis
2. Scatter plot
3. Correlation coefficient
4. Statistical significance of the correlation coefficient
5. Correlation vs. Causation
6. Determining the regression line
7. Plotting the regression line
8. Prediction
9. Coefficient of determination
10. In-class activities





We want to study the relationship between the number of times a student is absent in the class and the final grade.

We register data on number days absent and final grade of 20 students.

The correlation coefficient is $r = -0.975$ ($p\text{-value} < 0.05$)

That means:

- There is a strong positive relationship, and this relationship is statistically significant (at $\alpha = 0.05$).
- There is a strong negative relationship, and this relationship is statistically significant (at $\alpha = 0.05$).
- There is a strong negative relationship in the sample, but this relation is not statistically significant (at $\alpha = 0.05$).

TRUE

FALSE

TRUE

FALSE

TRUE

FALSE



We want to study the relationship between the number of times a student is absent in the class and the final grade.

We register data on number days absent and final grade of 20 students.

The correlation coefficient is $r = -0.975$ ($p\text{-value} < 0.05$)

That means:

- About 97.5% of the variation of final grades can be explained by the number of times a student is absent in class. The other 2.5% is unexplained; that is, it is not explained by x , but it could be explained by other variables such as amount of time studied, intelligence, etc.
- About 95% of the variation of final grades can be explained by the number of times a student is absent in class. The other 5% is unexplained; that is, it is not explained by x , but it could be explained by other variables such as amount of time studied, intelligence, etc.

TRUE

FALSE

TRUE

FALSE



Instagram App Users, Smartphone Users and Total Population

Below, we see the number of Instagram users, smartphone users and total population of 9 countries.

```
> Instagram <-  
read.table("C:/Users/vbv/Desktop/My_documents/Teaching/Teaching/Statistical_Data_Analysis/2022_2023/Exercises_in_R/Instagram.txt",  
header=TRUE)  
> Instagram
```

	Country	Total_population	Smartphone_owners	Instagram_users
1	USA	329.0	260.0	120
2	Brazil	212.0	96.9	77
3	Indonesia	270.0	83.9	63
4	Russia	144.0	95.4	44
5	Turkey	83.0	44.8	38
6	Japan	127.0	72.6	29
7	United_Kingdom	67.0	55.5	24
8	Mexico	132.0	65.6	24
9	Germany	82.4	65.9	21

```
> cor.test(Instagram$Total_population, Instagram$Instagram_users)
```

Pearson's product-moment correlation

data: Instagram\$Total_population and Instagram\$Instagram_users

t = 5.9425, df = 7, p-value = 0.0005744

alternative hypothesis: true correlation is not equal to 0

95 percent confidence interval:

0.6341979 0.9819273

sample estimates:

cor

0.9135475

```
> Regression_Instagram <- lm(Instagram_users~Total_population,  
data=Instagram)  
> summary(Regression_Instagram)
```

Call:

```
lm(formula = Instagram_users ~ Total_population, data = Instagram)
```

Residuals:

Min	1Q	Median	3Q	Max
-21.9064	-8.7790	0.6185	11.2082	15.6494

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-4.07543	10.10060	-0.403	0.698633
Total_population	0.32956	0.05546	5.943	0.000574 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 14.26 on 7 degrees of freedom

Multiple R-squared: 0.8346, Adjusted R-squared: 0.8109

F-statistic: 35.31 on 1 and 7 DF, p-value: 0.0005744

```
> predict(Regression_Instagram,data.frame(Total_population=175))
```

```
1  
53.59797
```

There is a weak positive relationship between total population and number of Instagram users.

TRUE

FALSE

About 83.5% of the variation in the number of Instagram users in a country can be explained by the number of habitants.

TRUE

FALSE

The predicted number of Instagram users in a country with 175 million habitants is 76.3 millions.

TRUE

FALSE

Videogames critics score and number of copies sold.



We want to study the relationship between critics score of videogames and number of copies sold.



We collect data from 30 videogames on:

- Rank - Ranking of overall sales
- Name - The games name
- Genre - Genre of the game
- Publisher - Publisher of the game
- Critic score - Score of the game on metacritic.com
- Copies sold – in millions

```
> Video_games <-
read.table("C:/Users/vbv/Desktop/My_documents/Teaching/Teaching/Statistical_Data_Analysis/2022_2023/Exercises_in_R/Video_games.txt", header=TRUE)
> Video_games
```

	Rank	Name	Genre	Publisher	Critic_score	Copies_sold_mill
1	1	Wii_Sports	Sports	Nintendo	7.7	82.86
2	2	Super_Mario_Bros.	Platform	Nintendo	9.4	40.24
3	3	Mario_Kart_Wii	Racing	Nintendo	8.2	37.14
4	4	PlayerUnknown's_Battlegrounds	Shooter	PUBG_Corporation	8.5	36.60
5	5	Wii_Sports_Resort	Sports	Nintendo	8.0	33.09
6	6	Pokemon_RGB_Version	Role-Playing	Nintendo	9.4	31.38
7	7	New_Super_Mario_Bros.	Platform	Nintendo	9.1	30.80
8	8	Tetris	Puzzle	Nintendo	7.3	30.26
9	9	New_Super_Mario_Bros_Wii	Platform	Nintendo	8.6	30.22
10	10	Minecraft	Misc	Mojang	9.3	30.01
11	11	Duck_Hunt	Shooter	Nintendo	7.1	28.31
12	12	Wii_Play	Misc	Nintendo	5.9	28.02
13	13	Kinect_Adventures	Party	Microsoft_Game_Studios	6.7	24.00
14	14	Nintendogs	Simulation	Nintendo	8.4	23.96
15	15	Mario_Kart_DS	Racing	Nintendo	9.1	23.60
16	16	Pokemon_Gold_Silver	Role-Playing	Nintendo	9.2	23.10
17	17	Wii_Fit	Sports	Nintendo	7.9	22.67
18	18	Wii_Fit_Plus	Sports	Nintendo	8.0	21.13
19	19	Super_Mario_world	Platform	Nintendo	8.5	20.61
20	20	Grand_Theft_Auto_V	Action	Rockstar_Games	9.4	20.00
21	21	Brain_Age	Misc	Nintendo	8.1	19.01
22	22	Garrys_Mod	Misc	Unknown	6.5	18.58
23	23	Super_Mario_Land	Platform	Nintendo	7.0	18.14
24	24	Mario_Kart_7	Racing	Nintendo	8.2	18.11
25	25	Pokemon_Diamond_Pearl	Role-Playing	Nintendo	8.6	17.67
26	26	Grand_Theft_Auto_San_Andreas	Action	Rockstar_Games	9.5	17.30
27	27	Super_Mario_Bros_3	Platform	Nintendo	7.5	17.28
28	28	Pokemon_X/Y	Role-Playing	Nintendo	8.9	16.37
29	29	Pokemon_Ruby_Sapphire	Role-Playing	Nintendo	8.8	16.22
30	30	Pokemon_Sun_Moon	Role-Playing	Nintendo	9.0	16.14


```
> cor.test(Video_games$Critic_score, Video_games$Copies_sold_mill)
```

Pearson's product-moment correlation

data: Video_games\$Critic_score and Video_games\$Copies_sold_mill

t = -0.30182, df = 28, p-value = 0.765

alternative hypothesis: true correlation is not equal to 0

95 percent confidence interval:

-0.4088272 0.3096771

sample estimates:

cor

-0.05694534

```
> Regression_videogames <- lm(Copies_sold_mill~Critic_score,  
data=Video_games)  
> summary(Regression_videogames)
```

Call:

```
lm(formula = Copies_sold_mill ~ Critic_score, data = Video_games)
```

Residuals:

Min	1Q	Median	3Q	Max
-9.799	-8.319	-3.112	4.289	56.009

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	32.6746	20.8329	1.568	0.128
Critic_score	-0.7563	2.5059	-0.302	0.765

Residual standard error: 12.92 on 28 degrees of freedom

Multiple R-squared: 0.003243, Adjusted R-squared: -0.03236

F-statistic: 0.09109 on 1 and 28 DF, p-value: 0.765

There is a significant positive relationship between critics score of videogames and number of copies sold.

TRUE

FALSE

We can use critics score to predict the number of copies sold.

TRUE

FALSE