Biostatistics Week II

Ege Ülgen, M.D.

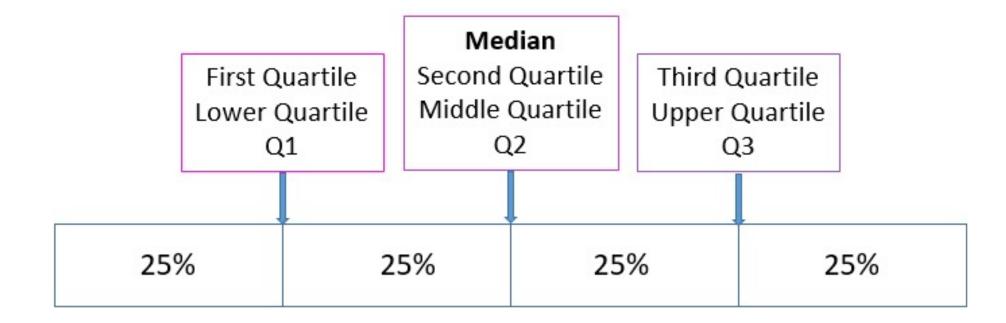
14 October 2021



Describing Distributions

- Shape
- Center
- (Measures of position)
- Spread
- Outliers

Quartiles



Quartiles

Recovery duration of 8 patients treated with a novel drug:
 30, 20, 24, 40, 65, 70, 10, 62

10, 20, 24,
$$30$$
, 40, 62, 65, 70 $Q_2 = 35$

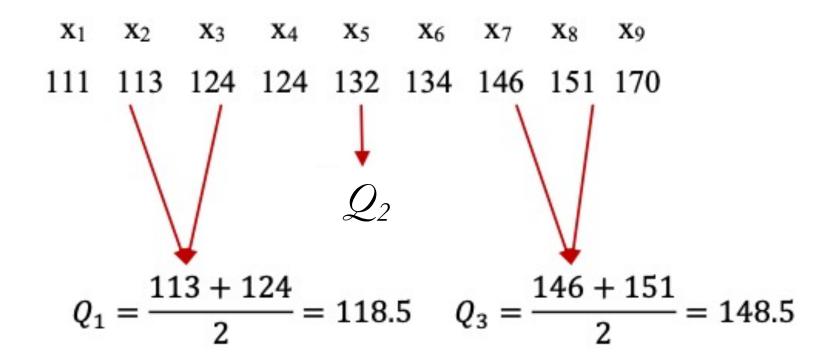
$$x_1$$
 x_2 x_3 x_4 x_5 x_6 x_7 x_8

10 20 24 30 40 62 65 70

 $Q_1 = \frac{20+24}{2} = 22$ $Q_3 = \frac{62+65}{2} = 63.5$

Quartiles

• Systolic blood pressure measurements of 9 patients: 151, 124, 132, 170, 146, 124, 113, 111, 134



Percentiles - Definition

100 * p percentile (0 ≤ p ≤ 1) is the data value for which:

- at least 100 * p of the data values are less than or equal to it
- at least 100 * (1 p) of the data values are greater than or equal to it

* If there are two values that satisfy the above conditions, the average of these values is taken as the 100 * p percentile

Percentiles - Algorithm

- 1. Sort data *X* in ascending order
- 2. Calculate $n \times p$
- 3. If np is not an integer, return $X_{ceiling(np)}$
- 4. Else (if $n \times p$ is an integer), return $(X_{np} + X_{np+1})/2$

Percentiles – simple example

- Original data: 13, 14, 12, 11, 19, 15, 18, 16, 17, 20 (n = 10)
- Sorted data: 11, 12, 13, 14, 15, 16, 17, 18, 19, 20
- 25th percentile (1st quartile): 13 (10 * 0.25 = 2.5 >> 3)
- 50th percentile (median): 15.5 (10 * 0.5 = 5 >> 5 & 6)
- 75th percentile (3rd quartile): 18 (10 * 0.75 = 7.5 >> 8)
- 90th percentile: 19.5 (10 * 0.9 = 9 >> 9 & 10)
- 95th percentile: 20 (10 * 0.95 = 9.5 >> 10)
- 97.5th percentile: 20 (10 * 0.975 = 9.75 >> 10)

Percentiles – another example

- Sorted data: 171, 174, 178, 181, 183, 183, 183, 184, 187, 188, 191, 191, 191, 192, 193, 193, 193, 194, 194, 195, 196, 196, 199, 200, 200, 200, 202, 204, 204, 205, 206, 211, 212, 213, 213, 216, 220, 221, 221, 227
- 25th percentile (1st quartile, Q1): 189.5 (40 * 0.25 = 10)
- 50th percentile (median, Q2): 195.5 (40 * 0.5 = 20)
- 75th percentile (3rd quartile, Q3): 205.5 (40 * 0.75 = 30)
- 90th percentile : 218 (40 * 0.9 = 36)
- 95th percentile: 221 (40 * 0.95 = 38)
- 97.5th percentile: 224 (40 * 0.975 = 39)

Quantiles – general formula

$$Q(q) = (1 - \gamma)X_j + \gamma X_{j+1}$$
 where:

- $\frac{j-m}{n} \le q \le \frac{j-m+1}{n}$
- $m \in \mathbb{R}$
- $0 \le \gamma \le 1$ and γ is a functi_q 1 of j and g
- j = floor(qn + m) and g = qn + m j

Type 7²:
$$\gamma = g$$
 and $m = 1 - p$

Type 2: m = 0 and $\gamma = 0.5$ when g = 0 and $\gamma = 1$ when g > 0

² By default, R uses **Type 7**

Quantiles – A simple example (type 7)

Original data: 13, 14, 12, 11, 19, 15, 18, 16, 17, 20 (n = 10)

Sorted data: 11, 12, 13, 14, 15, 16, 17, 18, 19, 20

$$p = 0.25$$
,

- -m=1-p=0.75,
- j = floor(n * p + m) = floor(10 * 0.25 + 0.75) = 3
- g = p * n + m j = 0.25 * 10 + 0.75 3 = 0.25
- $\gamma = g = 0.25$
- Q(0.25) = (1 0.25) * 13 + 0.25 * 14 = 13.25

Quantiles – A simple example (type 7)

Original data: 13, 14, 12, 11, 19, 15, 18, 16, 17, 20 (n = 10)

Sorted data: 11, 12, 13, 14, 15, 16, 17, 18, 19, 20

$$p = 0.5$$
,

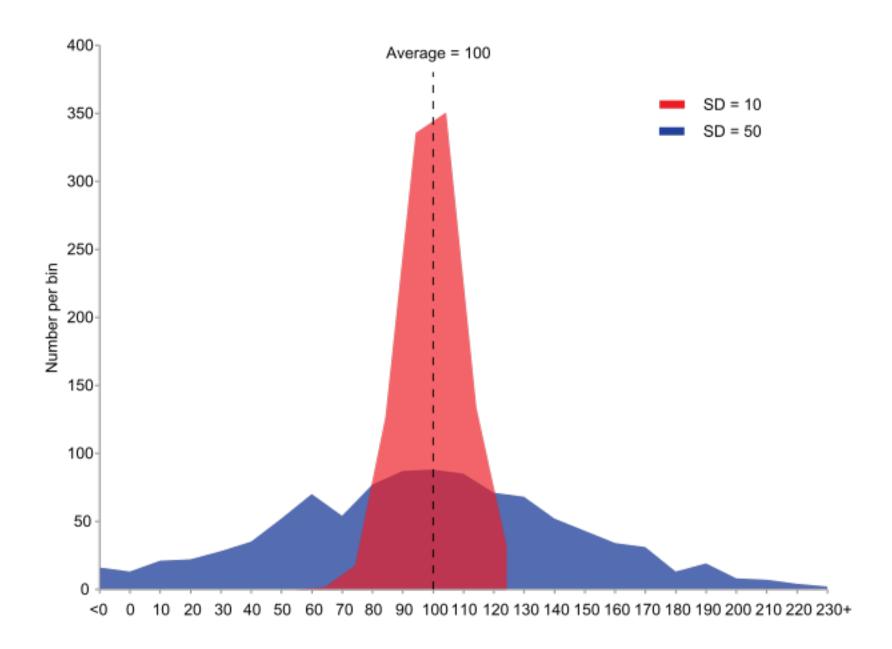
- -m=1-p=0.5,
- j = floor(pn + m) = floor(10 * 0.5 + 0.5) = 5
- -g = pn + m j = 0.5 * 10 + 0.5 5 = 0.5
- $\gamma = g = 0.5$
- Q(0.5) = (1 0.5) * 15 + 0.5 * 16 = 15.5

Describing Distributions

- Shape
- Center
- Spread
- Outliers

Measures of Spread

- The distances of the values to the center differ
 - The degree of these differences constitute the spread of the distribution
- Two distributions may have the same mean/median/mode and differ in terms of spread



Range

The difference between the maximal and minimal value

$$R = maximum - minimum$$

e.g., The ages of 12 arthritis patients:

30, 12, 15, 22, 40, 55, 20, 58, 25, 60, 23, 72

$$R = 72 - 12 = 60$$

Inter-Quartile Range

- The range quantifies the variability by using the range covered by all the data
- the Inter-Quartile Range (IQR) measures the spread of a distribution by describing the range covered by the middle 50% of the data

$$IQR = Q3 - Q1$$



Inter-Quartile Range

• Recovery durations of 8 patients in days: 30, 20, 24, 40, 65, 70, 10, 62

10, 20, 24, <u>30</u>, <u>40</u>, 62, 65, 70

$$x_1$$
 x_2 x_3 x_4 x_5 x_6 x_7 x_8

10 20 24 30 40 62 65 70

 $Q_1 = \frac{20+24}{2} = 22$ $Q_3 = \frac{62+65}{2} = 63$

$$IQR = 63.5 - 22 = 41.5$$

Variance and Standard Deviation

- Variance
 - A measure of how distant observations are from the mean
 - Population variance: σ^2
 - Sample variance: s²
- Because the unit of variance is quadratic, standard deviation is more widely used
- Standard deviation (sd)
 - Defined as the square-root of variance
 - Population sd: σ
 - Sample sd: s

Sample Variance and Standard Deviation

$$s^{2} = \frac{\sum_{j=1}^{n} (x_{j} - \bar{x})^{2}}{n-1}$$

Sample Variance and Standard Deviation

Ages of 6 patients in a study:

10, 15, 22, 26, 31, 40

$$\overline{x} = (10 + 15 + 22 + 26 + 31 + 40) / 6 = 24$$

$$s^2 = \frac{(10-24)^2 + (15-24)^2 + (22-24)^2 + (26-24)^2 + (31-24)^2 + (40-24)^2}{6-1} = 118$$

$$s = \sqrt{s^2} = \sqrt{118} = 10.863$$

Sample Variance and Standard Deviation

If y = x + c, where c is a constant, var(y) = var(x)If z = x * c, where c is a constant, $var(z) = c^2 var(x)$

Describing Distributions

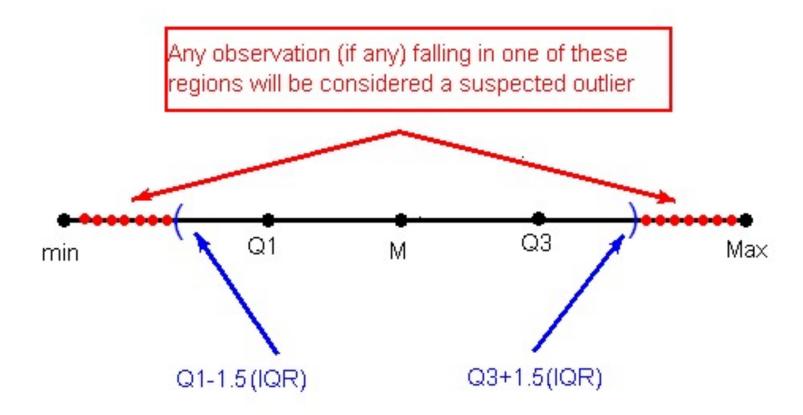
- Shape
- Center
- Spread
- Outliers

Outliers

Extreme observations that are distant from the rest of the data

- For
 - Lower Limit = $Q_1 1.5 * IQR$
 - Upper Limit = $Q_3 + 1.5 * IQR$
- Outliers are defined as any value(s) larger than the upper limit or smaller than the lower limit

Outliers



Outliers – Cholesterol Level Example

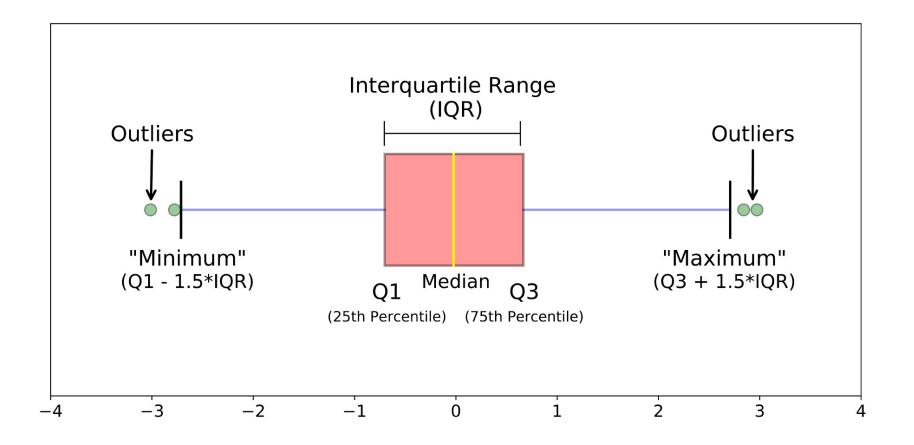
- Sorted data: 171, 174, 178, 181, 183, 183, 183, 184, 187, 188, 191, 191, 191, 192, 193, 193, 193, 194, 194, 195, 196, 196, 199, 200, 200, 200, 202, 204, 204, 205, 206, 211, 212, 213, 213, 216, 220, 221, 221, 227
- 25th percentile (1st quartile, Q_1): 189.5 (40 * 0.25 = 10)
- 75th percentile (3rd quartile, Q_3): 205.5 (40 * 0.75 = 30)
- IQR = 205.5 189.5 = 16
- LL = Q_1 1.5 * IQR = 189.5 1.5 * 16 = 165.5
- UL = Q_3 + 1.5 * IQR = 205.5 + 1.5 * 16 = 229.5

No outliers

Outliers – Cholesterol Level Example (cont.)

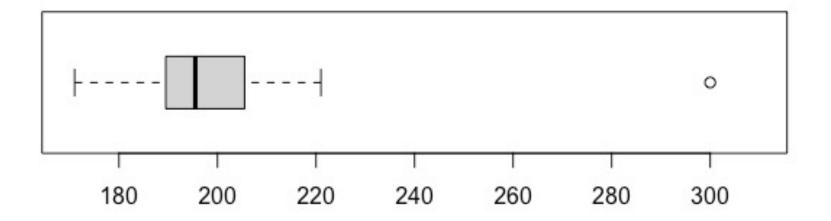
- Sorted data: 171, 174, 178, 181, 183, 183, 183, 184, 187, 188, 191, 191, 191, 192, 193, 193, 193, 194, 194, 195, 196, 196, 199, 200, 200, 200, 202, 204, 204, 205, 206, 211, 212, 213, 213, 216, 220, 221, 221, 300
- 25th percentile (1st quartile, Q_1): 189.5 (40 * 0.25 = 10)
- 75th percentile (3rd quartile, Q_3): 205.5 (40 * 0.75 = 30)
- IQR = 205.5 189.5 = 16
- LL = Q_1 1.5 * IQR = 189.5 1.5 * 16 = 165.5
- UL = Q_3 + 1.5 * IQR = 205.5 + 1.5 * 16 = 229.5
- 300 > UL => outlier

Box Plot



Box Plot – Example

• 171, 174, 178, 181, 183, 183, 183, 184, 187, 188, 191, 191, 191, 192, 193, 193, 193, 194, 194, 195, 196, 196, 199, 200, 200, 200, 202, 204, 204, 205, 206, 211, 212, 213, 213, 216, 220, 221, 221, 300

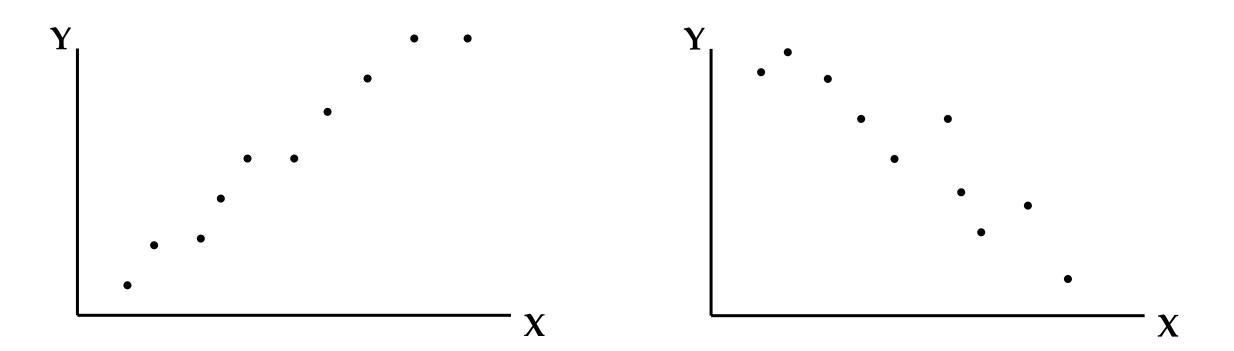


Left-Skewed Right-Skewed Symmetric \mathbf{Q}_1 \mathbf{Q}_2 \mathbf{Q}_3 $\mathbf{Q}_1 \ \mathbf{Q}_2 \ \mathbf{Q}_3$ $\mathbf{Q}_2 \ \mathbf{Q}_3$

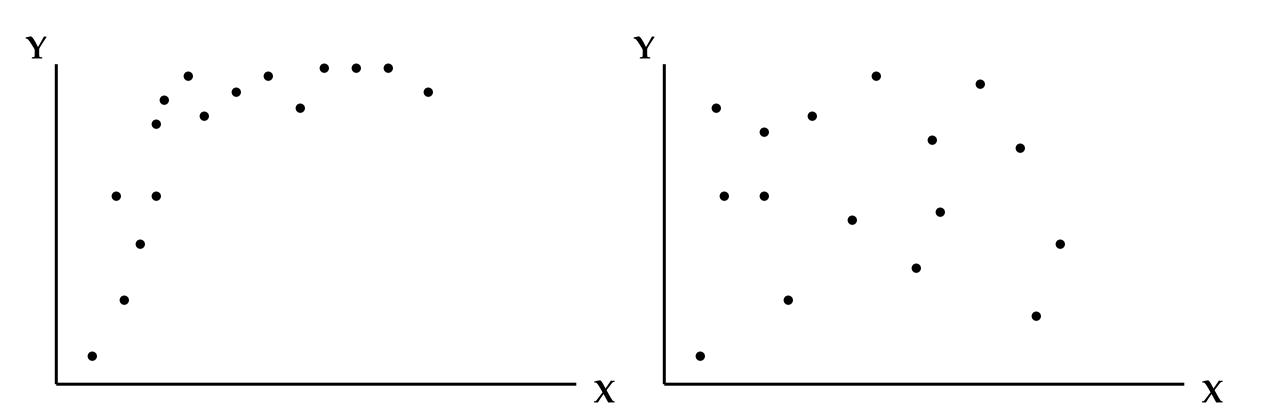
Exploratory Data Analysis (EDA)

- Examining Distributions exploring data one variable at a time.
- Examining Relationships exploring data two variables at a time.

Relationship between two variables



Relationship between two variables



Sample Covariance

A measure of how two variables change together

$$Cov(X,Y) = \frac{1}{n-1} \sum_{i=1}^{n} (X - \bar{X})(Y - \bar{Y})$$

Sample Covariance

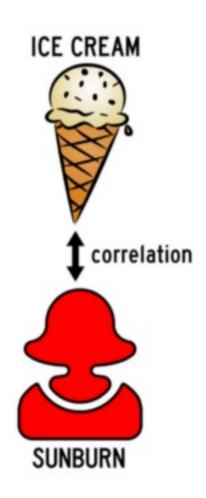
Proporties:

- $Cov(X,Y) \in \mathbb{R}$
- Cov(X,Y) = Cov(Y,X)
- Cov(X, X) = Var(X)
- $Cov(aX, bY) = abCov(X, Y), a, b \in \mathbb{R}$
- $Cov(X + a, Y + b) = Cov(X, Y), a, b \in \mathbb{R}$

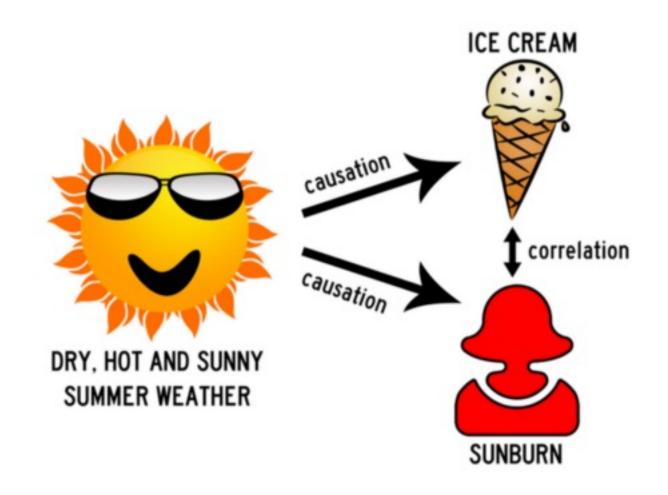
Correlation

- Correlation is a bivariate analysis that measures the strength of association between two variables and the direction of the relationships
- In terms of the strength of relationship, the value of the correlation coefficient varies between +1 and -1
- Correlation does not mean causation

Correlation does not mean causation



Correlation does not mean causation



Correlation Coefficient

A statistic that measures the relationship between two variables

- Pearson's r
 - Measures linear relationship
 - Both variables have to be normally distributed
- Spearman's ρ
 - Measures monotonic relationship
 - Based on rank non-parametric

Pearson Correlation Coefficient

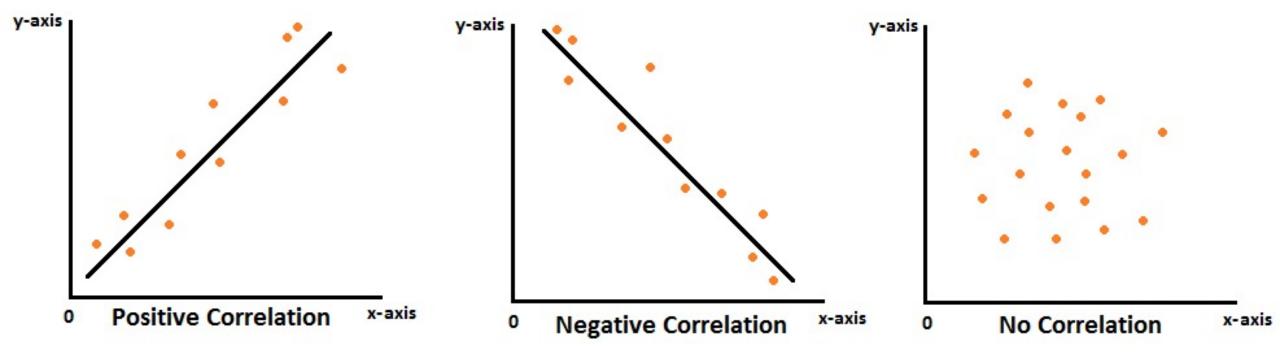
$$r_{X,Y} = \frac{Cov(X,Y)}{\sqrt{Var(X)Var(Y)}}$$

- A measure of the linear correlation between two variables X and Y
- takes values between -1 and 1
- unitless
- $r_{X,Y} = r_{Y,X}$
- r_{X,Y} = 0 means no linear relationship

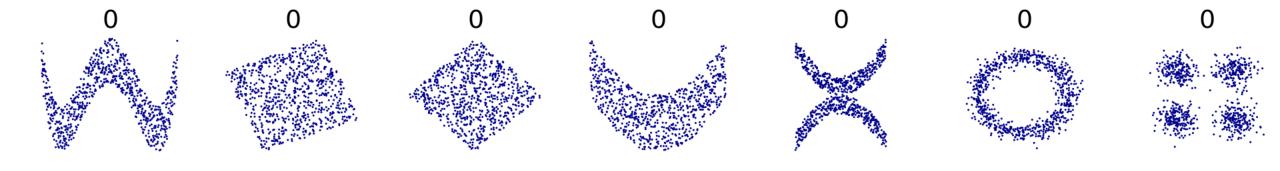
Pearson Correlation Coefficient

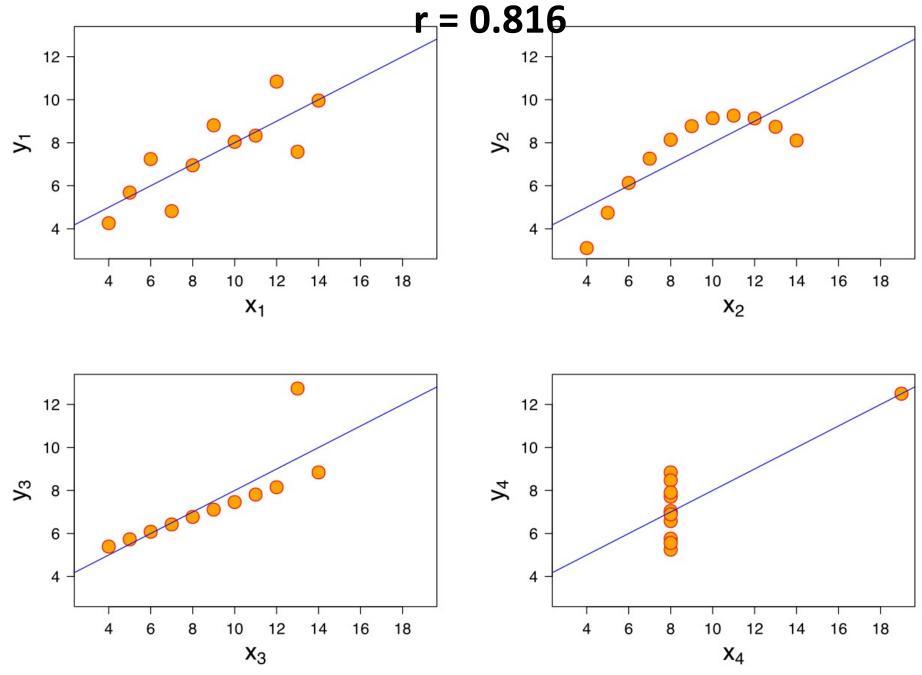
Cohen's (1988) conventions to interpret effect size:

- -|r| = 0.10 0.29: Weak
- -|r| = 0.30 0.49: Moderate
- *-* |r| ≥ 0.50: Strong



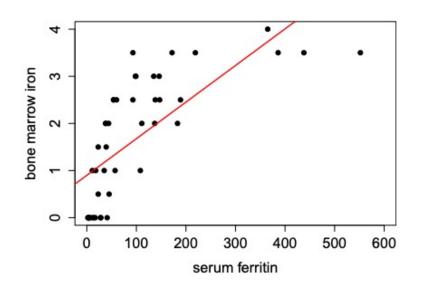
Suresh S. From correlation to causation [Internet]. Medium. 2020 [cited 2021 Oct 8]. Available from https://towardsdatascience.com/from-correlation-to-causation-49f566eea954





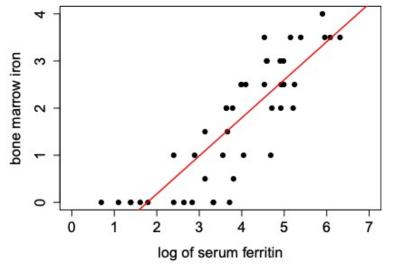
https://en.wikipedia.org/wiki/Correlation_and_dependence

Example: Relation between blood serum content of Ferritin and bone marrow content of iron.



$$r = 0.72$$

- Transformation to linear relation?
- Frequently a transformation to the normal distribution helps.

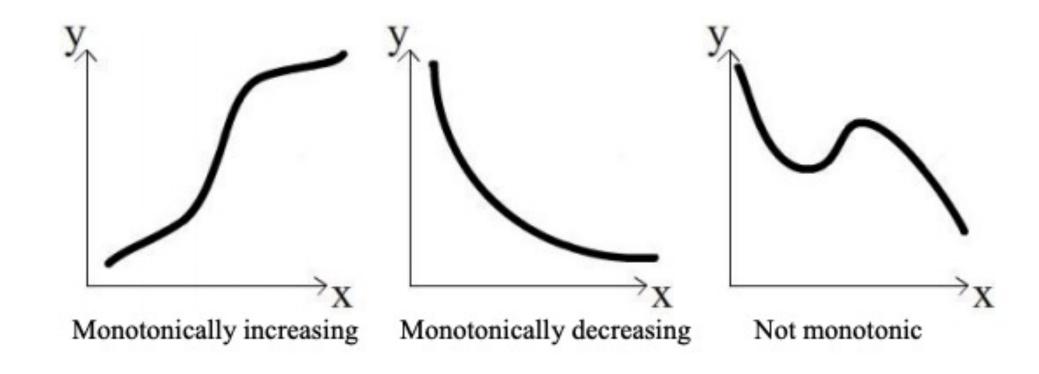


$$r = 0.85$$

Spearman Rank Correlation

- It assesses how well the relationship between two variables can be described using a monotonic function
- It does not carry any assumptions about the distribution of the data and is the appropriate correlation analysis when the variables are measured on a scale that is at least ordinal

Spearman Rank Correlation



Spearman Rank Correlation

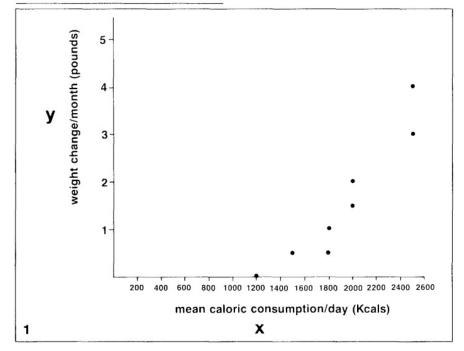
$$\rho = 1 - \frac{6\sum_{i=1}^{n} d_i^2}{n(n^2 - 1)}$$

- d_i := the difference between the ranks of corresponding variables (i.e., $d = X_i Y_i$)
- n := number of observations

TABLE 1. Sample data: Caloric consumption versus weight change

Patient	(X) Mean Caloric Consumption/Day	(Y) Weight Change/ Month
1	1,200	0.0
2	1,500	0.5
3	1,800	0.5
4	2,000	1.5
5	2,500	4.0
6	1,800	1.0
7	2,500	3.0
8	2,000	2.0

FIGURE 1. Scatter diagram for sample data given in Table 1 (caloric consumption vs weight change).



There is a strong positive relationship between mean caloric consumption/day and weight change/month

$$r = 0.94 \text{ or}$$

 $\rho = 0.97$

Units

- Mean: same unit with the data
- Median: same unit with the data
- Mode: same unit with the data
- Quantiles: same unit with the data
- Variance: square of the unit of the data
- Standard deviation: same unit with the data
- Covariance: square of the unit of the data
- Correlation: unitless

Brief Summary

- Quantiles can be used to partition the data and calculate specific positions
- The most used measures of spread are:
 - IQR
 - Variance and standard deviation
- Outliers can be defined based on Q1, Q3 and IQR
- Box plots can be used to display the distribution of a continuous variable
 - displays Q1, median, Q3, outliers
- The relationship between two variables can be visualized using scatter plots
- The relationship between two variables can be assessed using correlation
 - Pearson
 - Spearman