Statistical and Mathematical Methods for Data Analysis

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Textbooks

- ☐ Probability & Statistics for Engineers & Scientists,
 Ninth Edition, Ronald E. Walpole, Raymond H.
 Myer
- ☐ Elementary Statistics: Picturing the World, 6th Edition, Ron Larson and Betsy Farber
- ☐ Elementary Statistics, 13th Edition, Mario F. Triola

Reference books

- ☐ Probability and Statistical Inference, Ninth Edition, Robert V. Hogg, Elliot A. Tanis, Dale L. Zimmerman
- ☐ Probability Demystified, Allan G. Bluman
- □ Practical Statistics for Data Scientists: 50 Essential Concepts, Peter Bruce and Andrew Bruce
- ☐ Schaum's Outline of Probability, Second Edition, Seymour Lipschutz, Marc Lipson
- ☐ Python for Probability, Statistics, and Machine Learning, José Unpingco

References

☐ Probability & Statistics for Engineers & Scientists, Ninth edition, Ronald E. Walpole, Raymond H. Myer

☐ Elementary Statistics, Tenth Edition, Mario F. Triola

These notes contain material from the above resources.

Correlation

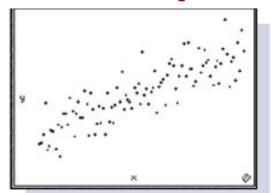
A correlation exists between two variables when one of them is related to the other in some way.

Exploring the Data

We can often see a **relationship between two variables** by constructing a **scatterplot**. When we examine a **scatterplot**, we should study the **overall pattern** of the plotted points. If there is a pattern, we should note its **direction**.

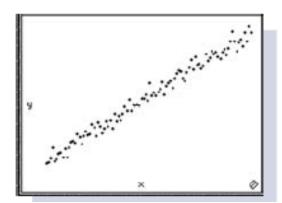
- ☐ An uphill direction suggests that as one variable increases, the other also increases.
- ☐ A downhill direction suggests that as one variable increases, the other decreases.
- ☐ We should look for **outliers**, which **are points that** lie very far away from all of the **other points**.

Scatter plots



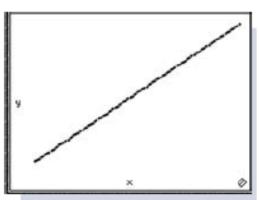
Positive correlation:

r = 0.851



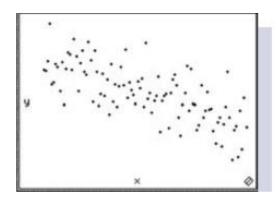
Positive correlation:

r = 0.991



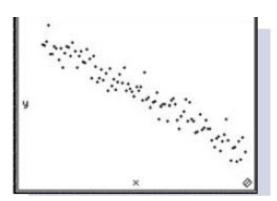
Perfect positive correlation:

r = 1



Negative correlation:

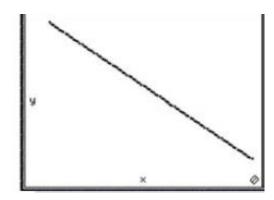
r = -0.702



Negative correlation:

r = -0.965

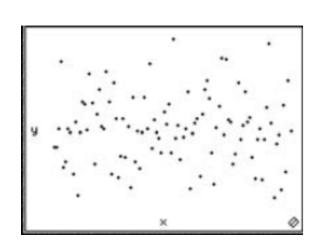
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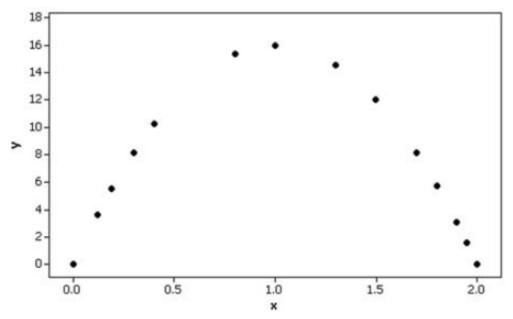
Perfect negative correlation:

r = -1

Scatter plots



No correlation: r = 0



Nonlinear relationship: r = -0.087

Palm reading



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Palm reading

- ☐ Some people believe that the length of their palm's lifeline can be used to predict longevity.
- ☐ In a letter published in the Journal of the American Medical Association, authors M. E. Wilson and L. E. Mather refuted that belief with a study of cadavers.
- ☐ Ages at death were recorded, along with the lengths of palm lifelines. The authors concluded that there is no significant correlation between age at death and length of lifeline. Palmistry lost, hands down.

Given any collection of sample paired data, the linear correlation coefficient r can always be computed, but the following requirements should be satisfied when testing hypotheses or making other inferences about r.

- 1. The sample of **paired** (x, y) data is a random sample of **independent quantitative data**.
- Visual examination of the scatterplot must confirm that the points approximate a straight-line pattern.
- 3. Any outliers must be removed if they are known to be errors. The effects of any other outliers should be considered by calculating r with and without the outliers included.

□ Note: Requirements 2 and 3 above are simplified attempts at checking this formal requirement: \Box The pairs of (x, y) data must have a bivariate normal distribution. (This assumption basically requires that for any fixed value of x, the corresponding values of y have a distribution that is bell-shaped, and for any fixed value of y, the values of x have a distribution that is bell-shaped.) ☐ This requirement is usually difficult to check, so for now, we will use Requirements 2 and 3 as listed

above.

Notation for the Linear Correlation Coefficient

☐ n: represents the number of pairs of data present.

☐ r: represents the linear correlation coefficient for a sample.

□ ρ: Greek letter **rho** used to represent the **linear correlation coefficient** for a **population**.

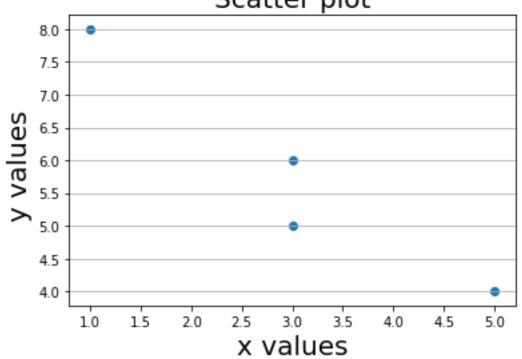
$$\mathbf{r} = \frac{\mathsf{n}(\sum \mathsf{x}\mathsf{y}) - (\sum \mathsf{x})(\sum \mathsf{y})}{\sqrt{\mathsf{n}(\sum \mathsf{x}^2) - (\sum \mathsf{x})^2} \sqrt{\mathsf{n}(\sum \mathsf{y}^2) - (\sum \mathsf{y})^2}}$$

Example Calculating *r* Using the simple random sample of data given in the table, find the value of the **linear** correlation coefficient *r*.

x	3	1	3	5
y	5	8	6	4

REQUIREMENT The data are a **simple random sample**. The accompanying **Python-generated scatterplot** shows **a pattern of points** that does appear to be a **straight-line pattern**. There are no outliers. We can proceed with the calculation of the linear correlation coefficient **r**.

Scatter plot



X	y	ху	x^2	y^2
3	5	15	9	25
1	8	8	1	64
3	6	18	9	36
5	4	20	25	16
$\sum x = 12$	$\sum y = 23$	$\sum xy = 61$	$\sum x^2 = 44$	$\sum y^2 = 141$

$$\mathbf{r} = \frac{\mathbf{n}(\sum \mathbf{x}\mathbf{y}) - (\sum \mathbf{x})(\sum \mathbf{y})}{\sqrt{\mathbf{n}(\sum \mathbf{x}^2) - (\sum \mathbf{x})^2} \sqrt{\mathbf{n}(\sum \mathbf{y}^2) - (\sum \mathbf{y})^2}}$$

$$r = \frac{4(61) - (12)(23)}{\sqrt{4(44) - (12)^2} \sqrt{4(141) - (23)^2}}$$
$$r = \frac{-32}{\sqrt{32}\sqrt{35}} = -0.956$$

☐ These calculations get quite messy with larger data sets, so it's fortunate that the linear correlation coefficient can be found automatically with many different calculators and computer programs

Interpreting the Linear Correlation Coefficient

- \square We need to interpret a calculated value of r, such as the value of -0.956 found in the preceding example.
- \Box The value of r must always fall between -1 and +1 inclusive.
- ☐ If *r* is close to 0, we conclude that there is no linear correlation between *x* and *y*, but if *r* is close -1 to or +1 we conclude that there is a linear correlation between *x* and *y*.

Properties of the Linear Correlation Coefficient *r*

- 1. The value of r is always between -1 and +1 inclusive. That is, $-1 \le r \le +1$
- 2. The value of r does not change if all values of either variable are converted to a different scale.
- **3.** The value of **r** is **not affected** by the choice of **x** or **y**. Interchange all x- and y-values and the value of **r** will not change.
- **4. r measures** the strength of a **linear relationship**. It is **not designed** to measure the **strength of a relationship** that is **not linear**.

Hypothesis Test for Correlation

Assume: r = 0.926, n = 8

1. We state our hypothesis as:

 H_0 : $\rho = 0$ (There is no linear correlation.)

 H_1 : $\rho \neq 0$ (There is a linear correlation.)

2. The level of significance is set $\alpha = 0.05$.

3. Test statistic to be used is
$$t_{cal} = \frac{r}{\sqrt{\frac{1-r^2}{n-2}}}$$

4. Calculations:

$$t_{cal} = \frac{0.926}{\sqrt{\frac{1 - (0.926)^2}{8-2}}} = 6.008$$

5. Critical region:

$$|\mathbf{t}_{cal}| > \mathbf{t}_{tab}$$
, where $\mathbf{t}_{tab} = \mathbf{t}_{(\alpha/2, n-2)}$

$$t_{tab} = t_{(0.0250, 6)} = 2.447$$

- **6. Conclusion:** Since t_{cal} is greater than the t_{tab} , so we reject H_o .
- ☐ There is **sufficient evidence** to support the claim of a linear correlation.

Examples: Applications of correlation

Buying a TV Audience The *New York Post* published the **annual** salaries (in millions) and the number of viewers (in millions), with results given below for Oprah Winfrey, David Letterman, Jay Leno, Kelsey Grammer, Barbara Walters, Dan Rather, James Gandolfini, and Susan Lucci, repsectively. Is there a correlation between salary and number of viewers?

Salary	100	14	14	35.2	12	7	5	1
Viewers	7	4.4	5.9	1.6	10.4	9.6	8.9	4.2

Examples: Applications of correlation

Parent Child Heights Listed below are heights (in inches) of mothers and heights (in inches) of their daughters (based on data from the National Health Examination Survey). Does there appear to be a linear correlation between mother's heights and the heights of their daughters?

Mother's height	63	67	64	60	65	67	59	60
Daughter's height	58.6	64.7	65.3	61.0	65.4	67.4	60.9	63.1

Basic Concepts of Regression

☐ In some cases, two variables are related in a deterministic way, meaning that given a value for one variable, the value of the other variable is automatically determined without any error.

For example, the **total cost y** of an item with a list price of x and a sales tax of 5% can be found by using the deterministic equation y = 1.05x. If an item is priced at \$100, its total cost is \$105.

Probabilistic Models

- ☐ In **probabilistic models**, meaning that one variable is **not determined** completely by the **other variable**.
- ☐ For example, a **child's height** is not determined completely by the **height of the father (or mother).**

□Sir Francis Galton (1822–1911) studied the phenomenon of heredity and showed that when tall or short couples have children, the heights of those children tend to regress, or revert to the more typical mean height for people of the same gender.

Notations

- The regression equation expresses a relationship between x (called the explanatory variable, or predictor variable, or independent variable) and (called the response variable, or dependent variable).
- The typical equation of a straight line y = mx + b is expressed in the form $\hat{y} = b_0 + b_1 x$ or $\hat{y} = a + bx$, where b_0 or a is the y-intercept and b_1 or b is the slope.

The given notation shows that b_0 and b_1 are sample statistics used to estimate the population parameters β_0 and β_1 .

We will use paired sample data to estimate the regression equation. Using only sample data, we can't find the exact values of the population parameters β_0 and β_1 , but we can use the sample data to estimate them with b_0 and b_1

1. The sample of paired (x, y) data is a random sample of quantitative data.

2. Visual examination of the scatterplot shows that the points approximate a straight-line pattern.

3. Any **outliers** must be **removed** if they are known to be errors. Consider the effects of any outliers that are not known errors.

Note: Requirements 2 and 3 above are simplified attempts at checking these formal requirements for regression analysis:

- ☐ For each fixed value of x, the corresponding values of y have a distribution that is bell-shaped.
- ☐ For the different fixed values of x, the distributions of the corresponding y-values all have the same variance.
- \Box For the different fixed values of x, the distributions of the corresponding y-values have means that lie along the same straight line.
- \Box The y values are independent.

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☐ Results are **not seriously affected** if departures from **normal distributions** and equal variances are not too extreme.

Definitions

Given a collection of paired sample data, the regression equation

$$\hat{y} = b_0 + b_1 x$$

algebraically describes the relationship between the two variables. The graph of the regression equation is called the regression line (or *line of best fit,* or *least-squares line*).

Notation for Regression Equation

	Population Parameter	Sample Statistic
y-intercept of regression equation	β_0	$\mathbf{b_0}$
Slope of regression equation	β_1	b ₁
Equation of the regression line	$Y = \beta_0 + \beta_1 x$	$\hat{y} = b_0 + b_1 x$

Finding the slope b_1 and y-intercept b_0 in the regression equation $\hat{y} = b_0 + b_1 x$

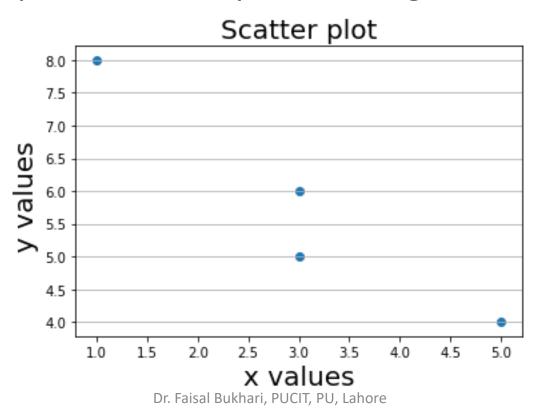
Slope	$\mathbf{b_1} = \frac{n(\sum xy) - (\sum x)(\sum y)}{n(\sum x^2) - (\sum x)^2}$
y-intercept:	$\mathbf{b_0} = \overline{y} - \mathbf{b_1} \overline{x}$ or $\mathbf{b_0} = \frac{(\sum y)(\sum x^2) - (\sum x)(\sum xy)}{n(\sum x^2) - (\sum x)^2}$

Example Finding the Regression Equation

Use the given sample data to find the regression equation.

X	3	1	3	5
y	5	8	6	4

REQUIREMENT The data are a simple random sample. The accompanying Python-generated scatterplot shows a pattern of points that does appear to be a straight-line pattern. There are no outliers. We can proceed to find the slope and intercept of the regression line.



X	y	ху	x^2	y^2
3	5	15	9	25
1	8	8	1	64
3	6	18	9	36
5	4	20	25	16
$\sum x = 12$	$\sum y = 23$	$\sum xy = 61$	$\sum x^2 = 44$	$\sum y^2 = 141$

$$\mathbf{b_1} = \frac{\mathsf{n}(\sum \mathsf{x}\mathsf{y}) - (\sum \mathsf{x})(\sum \mathsf{y})}{\mathsf{n}(\sum \mathsf{x}^2) - (\sum \mathsf{x})^2}$$

$$\mathbf{b_1} = \frac{4(61) - (12)(23)}{4(44) - (12)^2} = \frac{-32}{32} = -1$$

$$\overline{x} = \frac{12}{4} = 3$$

$$\overline{y} = \frac{23}{4} = 5.75$$

$$\mathbf{b_0} = \overline{\mathbf{y}} - \mathbf{b_1} \overline{\mathbf{x}}$$

$$b_0 = 5.75 - (-1)(3)$$

$$b_0 = 8.75$$

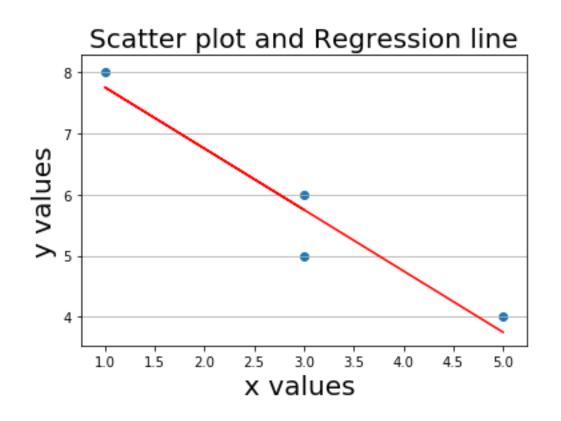
 \square Knowing the slope b_1 and y-intercept b_0 , we can now express the estimated equation of the regression line as

$$\hat{y} = b_0 + b_1 x$$

 $\hat{y} = 8.75 - 1x$

We should realize that this equation is an *estimate* of the true regression equation $Y = \beta_0 + \beta_1 x$. This estimate is based on one particular set of sample data, but another sample drawn from the same population would probably lead to a slightly different equation.

Scatter plot and Regression line



Using the Regression Equation for Predictions

Regression equations are often useful for *predicting* the value of one variable, given some particular value of the other variable.

☐ If the regression line fits the data quite well, then it makes sense to use its equation for predictions, provided that we don't go beyond the scope of the available values.

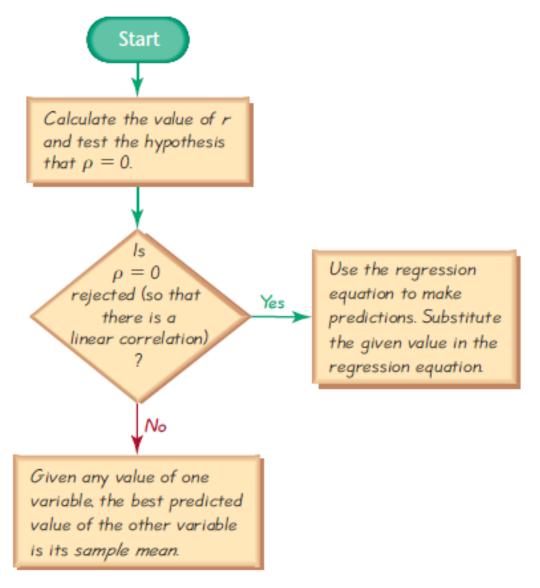
Using the Regression Equation for Predictions

In **predicting a value** of **y based** on some given value of **x** . . .

1. If there is **not** a linear correlation, the best predicted **y-value** is \overline{y} .

2. If there is a linear correlation, the best predicted y-value is found by substituting the x-value into the regression equation.

Procedure for Predicting



Guidelines for Using the Regression Equation

1. If there is no linear correlation, don't use the regression equation to make predictions.

2. When using the regression equation for predictions, stay within the scope of the available sample data. If you find a regression equation that relates women's heights and shoe sizes, it's absurd to predict the shoe size of a woman who is 10 ft tall.

Guidelines for Using the Regression Equation

3.A regression equation based **on old data** is not **necessarily valid now**. The regression equation relating **used-car prices** and **ages of cars** is no longer usable if it's based on data from the 1990s.

4.Don't make predictions about a population that is different from the population from which the sample data were drawn. If we collect sample data from men and develop a regression equation relating age and TV remotecontrol usage, the results don't necessarily apply to women. If we use state averages to develop a regression equation relating SAT math scores and SAT verbal scores, the results don't necessarily apply to individuals.