

Correlation

Colby Community College

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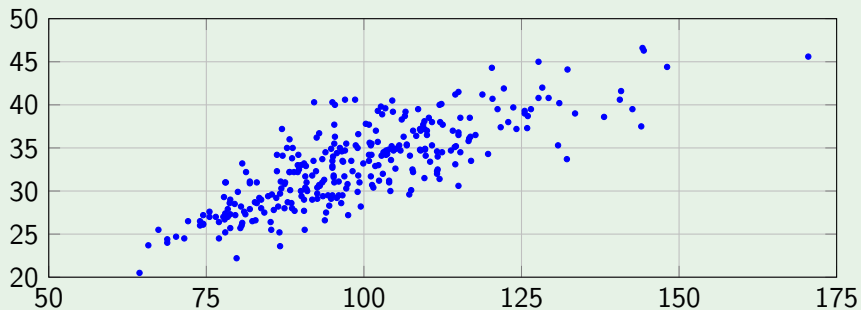
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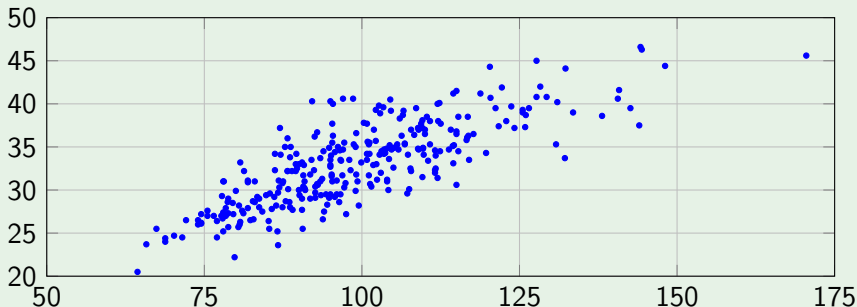
Note

Because conclusions based on visual examinations of scatterplots are largely subjective, we need more objective measures. We use the linear correlation coefficient r , which is a number that measures the strength of the linear association between the two variables.

Example 1

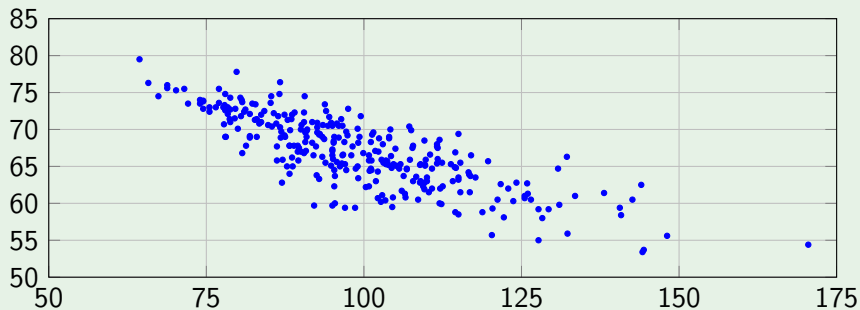


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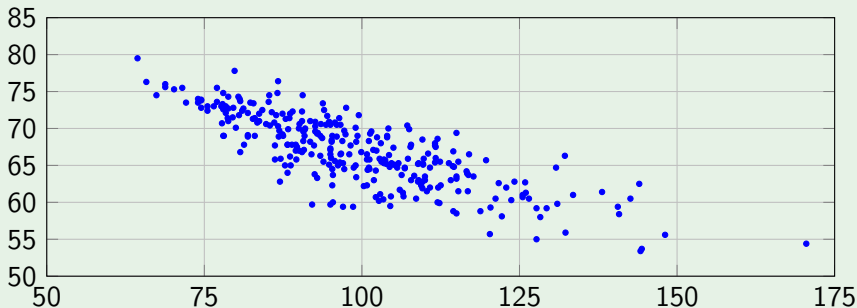


Distinct straight-line, or linear, pattern. We say that there is a positive linear correlation ($r = 0.80241$) between x and y , since as the x values increase, the corresponding y values also increase.

Example 2

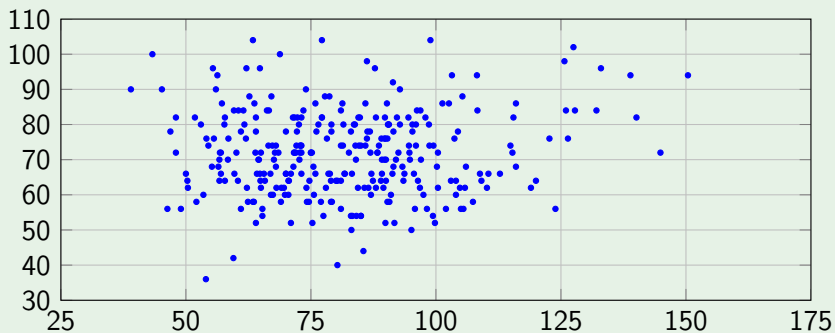


Example 2

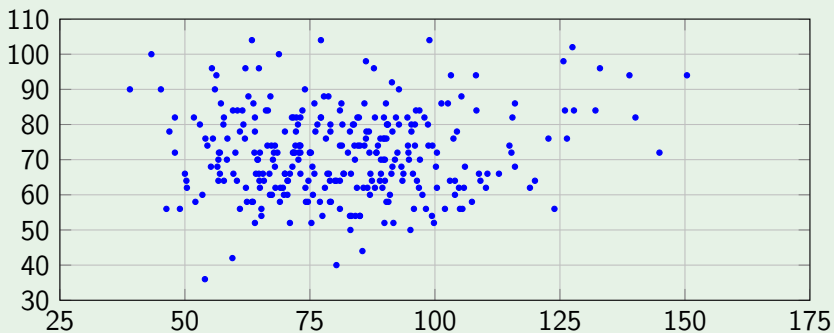


Distinct straight-line, or linear, pattern. We say that there is a positive linear correlation ($r = -0.80241$) between x and y , since as the x values increase, the corresponding y values also increase.

Example 3



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The points do not show any obvious pattern ($r = 0.08161$), and this lack of a pattern suggests that there is no relationship between the variables.

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The following should be satisfied when using the sample paired data to make a conclusion about linear correlation in the corresponding population of paired data.

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- 1 The sample of paired data is a simple random of quantitative data.
- 2 Visual examination of the scatterplot must confirm that the points approximate a straight-line pattern.
- 3 Because the results can be strongly affected by outliers, any outliers must be removed if they are known to be errors.

Caution

A linear correlation coefficient r can always be calculated, whether or not it applies.

Formula

The formula for r is:

$$r = \frac{\sum(z_x z_y)}{n - 1}$$

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Alternative Formula

A formula for r that is better for hand calculations is:

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

Properties of the Linear Correlation Coefficient

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- r measures the strength of a linear relationship. It is not designed to measure the strength of a relationship that is not linear.
- r is very sensitive to outliers in the sense that a single outlier could dramatically affect its value.

Is There a Linear Correlation?

Technology will generate a P -value along with r . If we have a significance level α , then

$P\text{-value} \leq \alpha$: Supports the claim of a linear correlation.

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Example 4

For Data Set 2 “Foot and Height,” when we use technology to calculate the linear correlation between the foot length and age of 40 randomly selected people we get:

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Because $0.02287 \leq 0.05$ we have evidence of a linear correlation.

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this implies that about 87.1% of the variation in ages cannot be explained by rates of chocolate consumption.

Do Not Assume That Correlation Implies Causality!

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Do Not Ignore The Possibility of a Nonlinear Relationship

If there is no linear correlation, there might be some correlation that is not linear.

Formal Hypothesis Test

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

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The test statistic (with $n - 2$ degrees of freedom) is

$$t = \frac{r}{\sqrt{\frac{1 - r^2}{n - 2}}}$$

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Using Data Set 16 let us conduct a formal hypothesis test of the claim that there is a linear correlation between chocolate consumption in a country and how many Nobel Laureates are from that country using a 0.05 significance.

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Because the P -value is less than the significance level, we reject H_0 .

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Claim of Negative Correlation

When testing a claim of a negative linear correlation use

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Claim of Positive Correlation

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