

Ch. 10: Two Quantitative Variables

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10.1: Two Quantitative Variables

Scatterplots and Correlation

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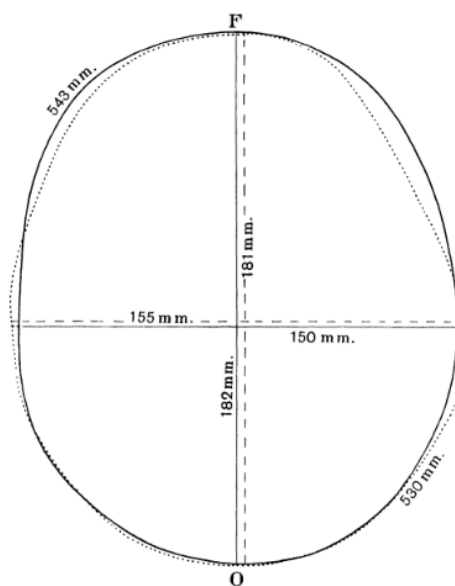
Graphical Summaries of Quantitative Variables

R.J. Gladstone (1905). "A Study of the Relations of the Brain to the Size of the Head", *Biometrika*, Vol. 4, p 105-123.

Data collected during 237 autopsies at Middlesex Hospital in London, excluding cases "in which the brain showed a distinctly pathological condition which would have obviously affected its weight"

Variables:

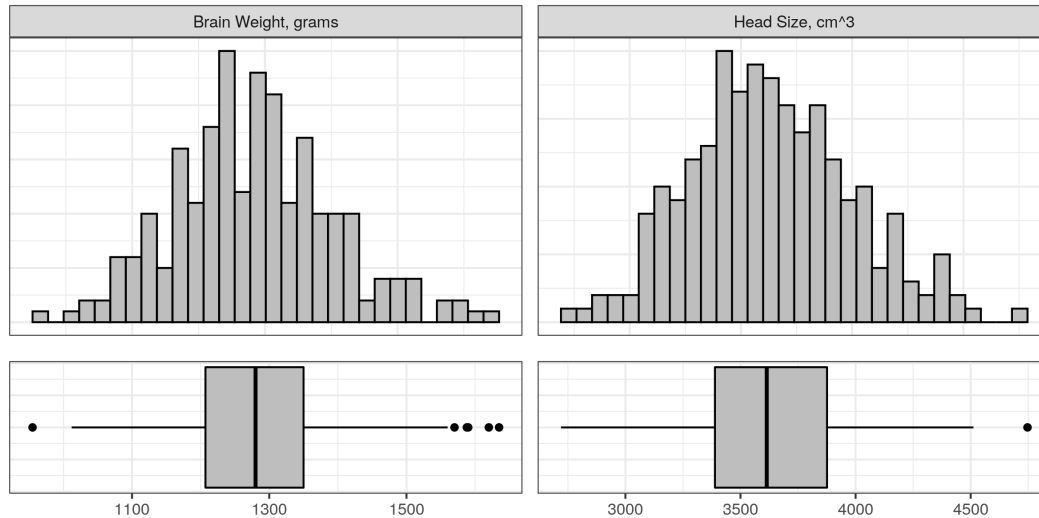
- Gender
- Age (20 - 45 or 46+)
- Brain Weight (g)
- Head Size (cubic cm) the smallest rectangular block which could contain the head



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Graphical Summaries of Quantitative Variables

A single quantitative variable can be summarized visually using a histogram or a bar chart:

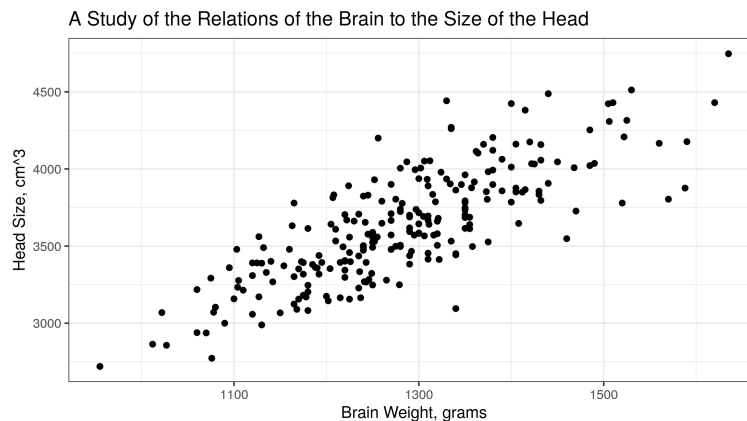


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Graphical Summaries of Quantitative Variables

But, summarizing each variable separately doesn't tell us how the two variables might be related.

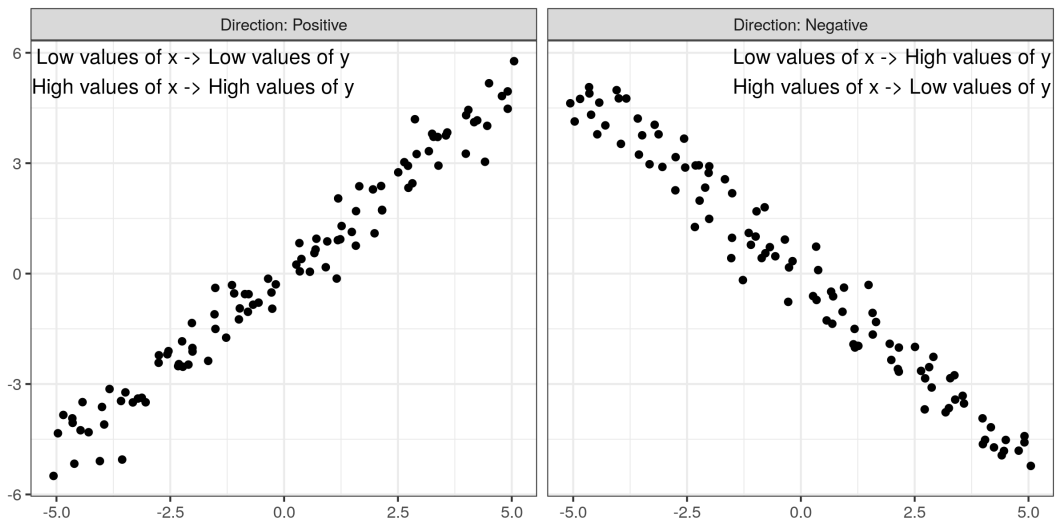
Is there a relationship between brain weight and head size? How do you know?



A **scatterplot** is a plot with the explanatory variable on the x-axis, and the response variable on the y-axis. Observations are shown as points corresponding to a set of quantitative measurements.

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Describing Variable Relationships: Direction

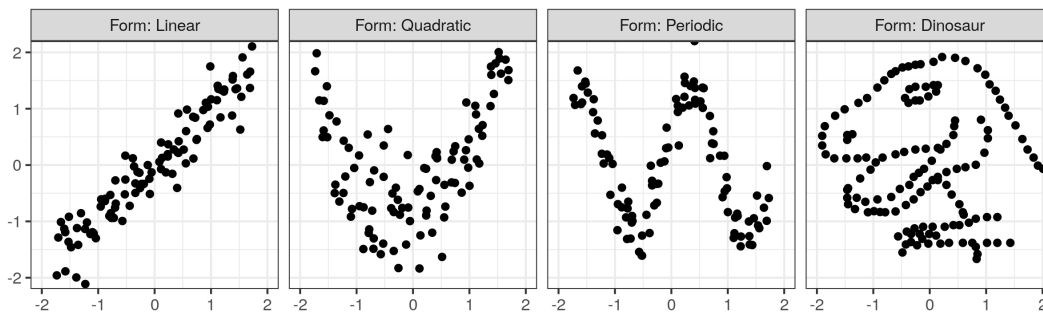


Positive slope: as x increases, y increases too.

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Describing Variable Relationships: Form

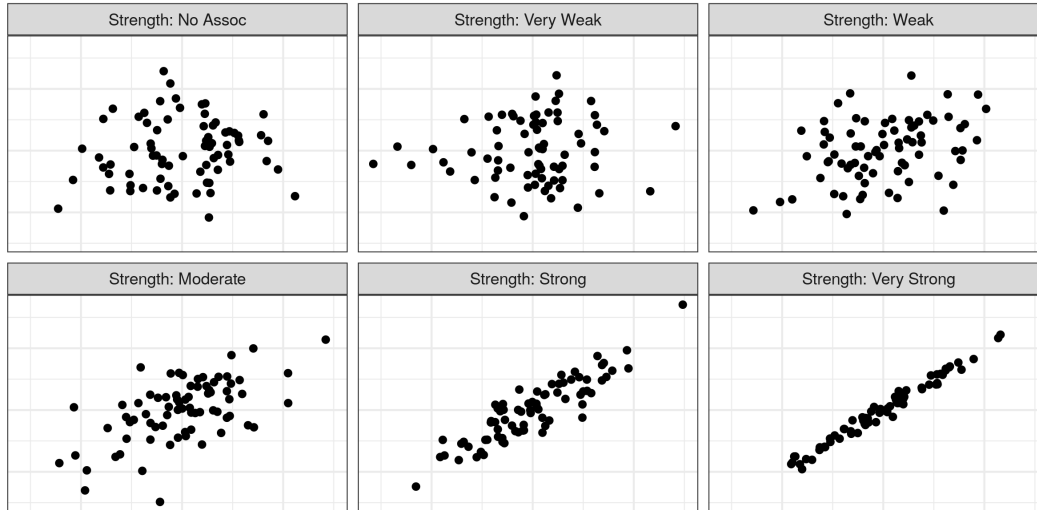
The **form** of an association is whether it follows a linear pattern, or some sort of more complicated pattern - periodic, polynomial (quadratic, cubic, etc.)



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Describing Variable Relationships: Strength

The **strength** of an association indicates how well the value of one variable can be predicted if you know the value of the other variable.

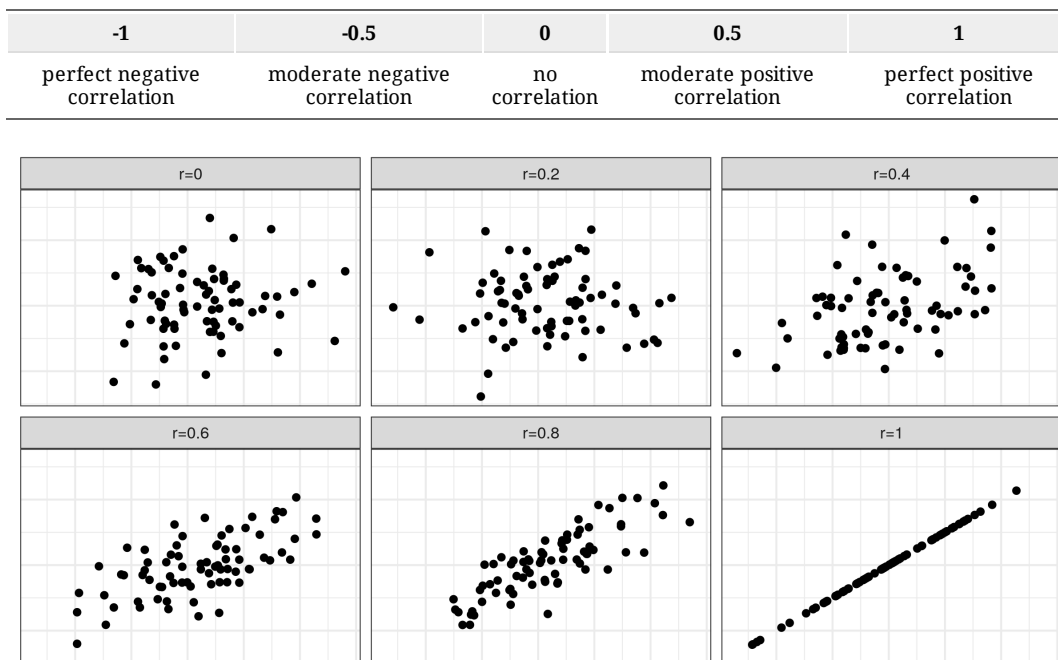


We can describe the strength and direction of a *linear* relationship using the **correlation coefficient**

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Correlation Coefficient

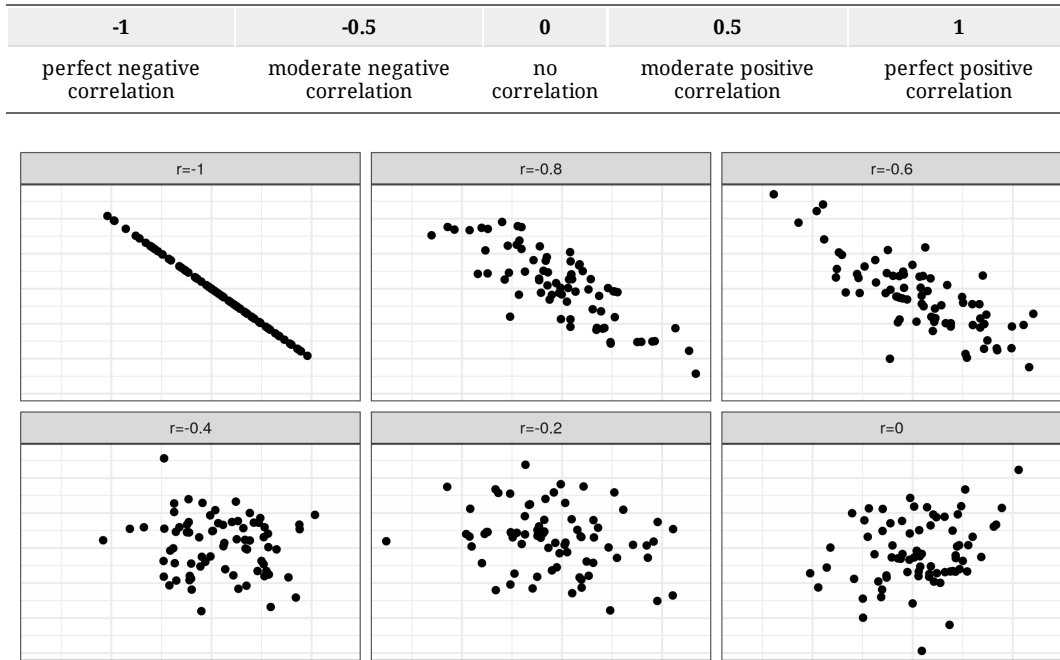
The **correlation coefficient**, r , is always between -1 and 1.



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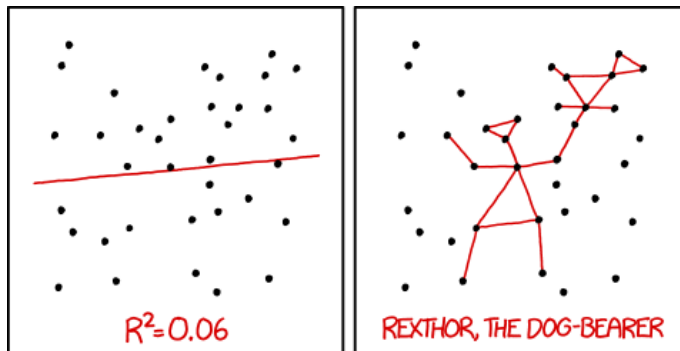
Correlation Coefficient

The **correlation coefficient**, r , is always between -1 and 1.



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Correlation Coefficient

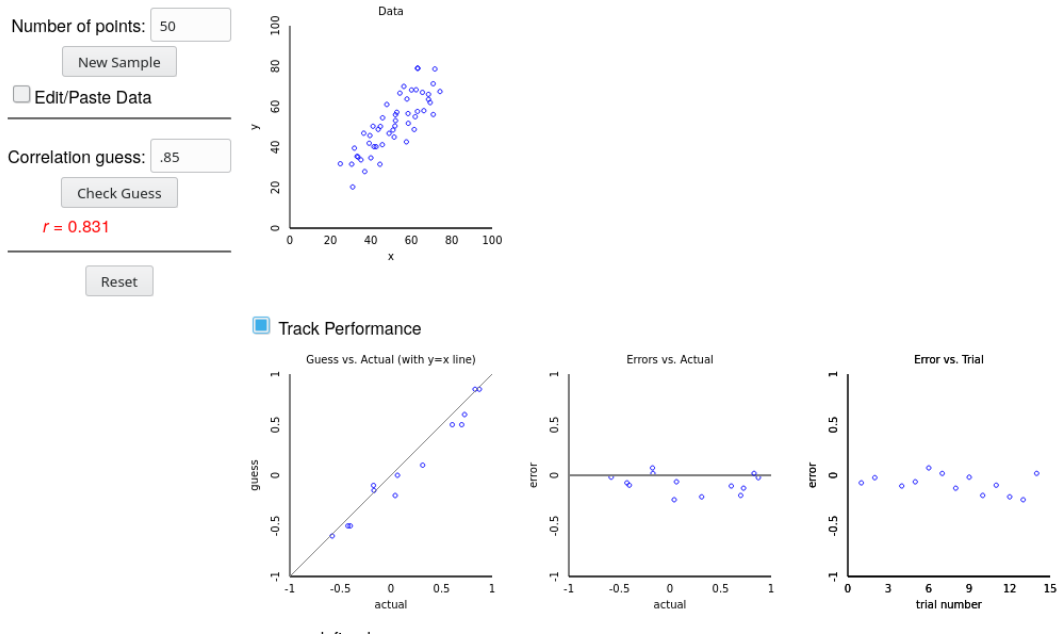


I DON'T TRUST LINEAR REGRESSIONS WHEN IT'S HARDER TO GUESS THE DIRECTION OF THE CORRELATION FROM THE SCATTER PLOT THAN TO FIND NEW CONSTELLATIONS ON IT.

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Correlation Coefficient

Get a feel for it by [playing the correlation guessing game!](#)

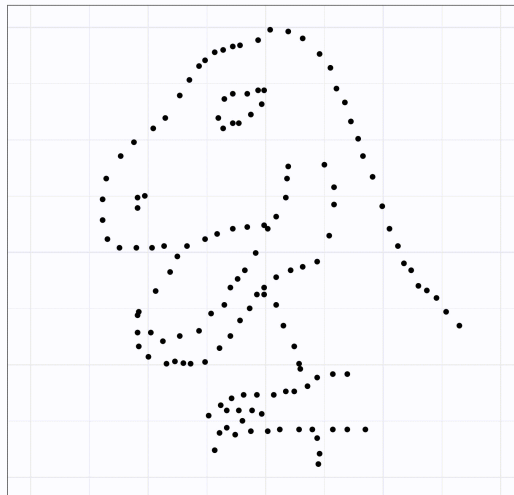


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Correlation Coefficient

The correlation coefficient is only useful for showing the strength of linear relationships.

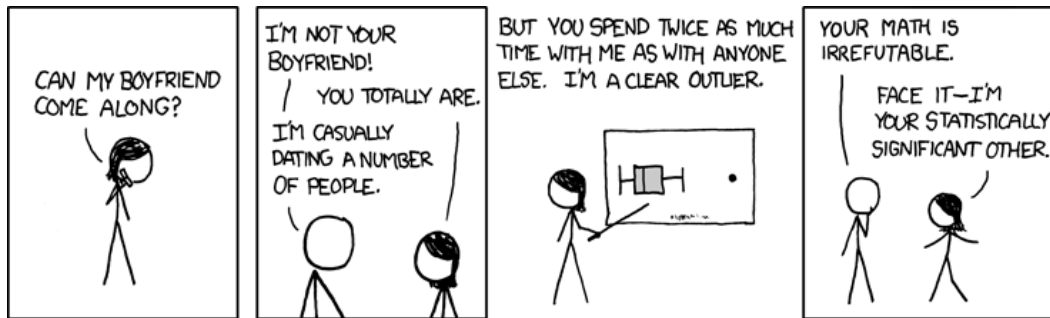
X: 47.26 (SD = 16.77), Y: 47.83 (SD = 26.94), $r = -0.06$



All of these plots have essentially the same correlation coefficient, but in some cases there are very clear associations between x and y

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Outliers and Influential Observations



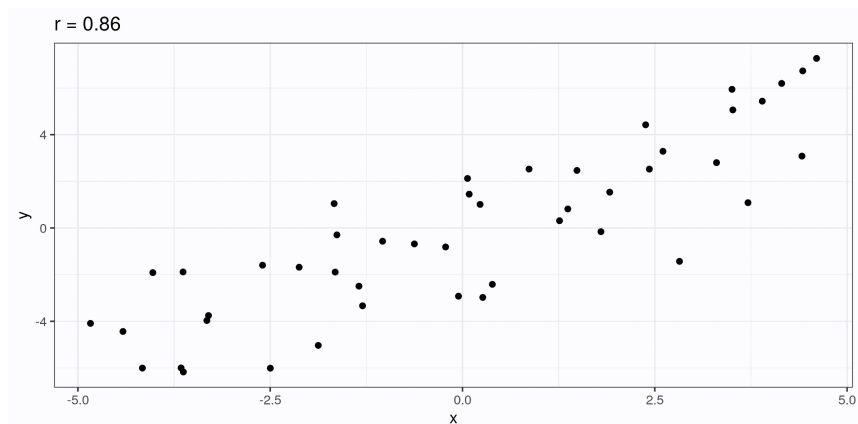
With one variable, outliers are fairly easy to spot

When there are two variables, we don't just have to worry about outliers in one dimension; we also have to worry about **influential observations**

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Outliers and Influential Observations

Influential observations are observations which, if included, change our understanding of the relationship between two variables.



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