Correlation

Ling250/450: Data Science for Linguistics
C.M. Downey
Spring 2025

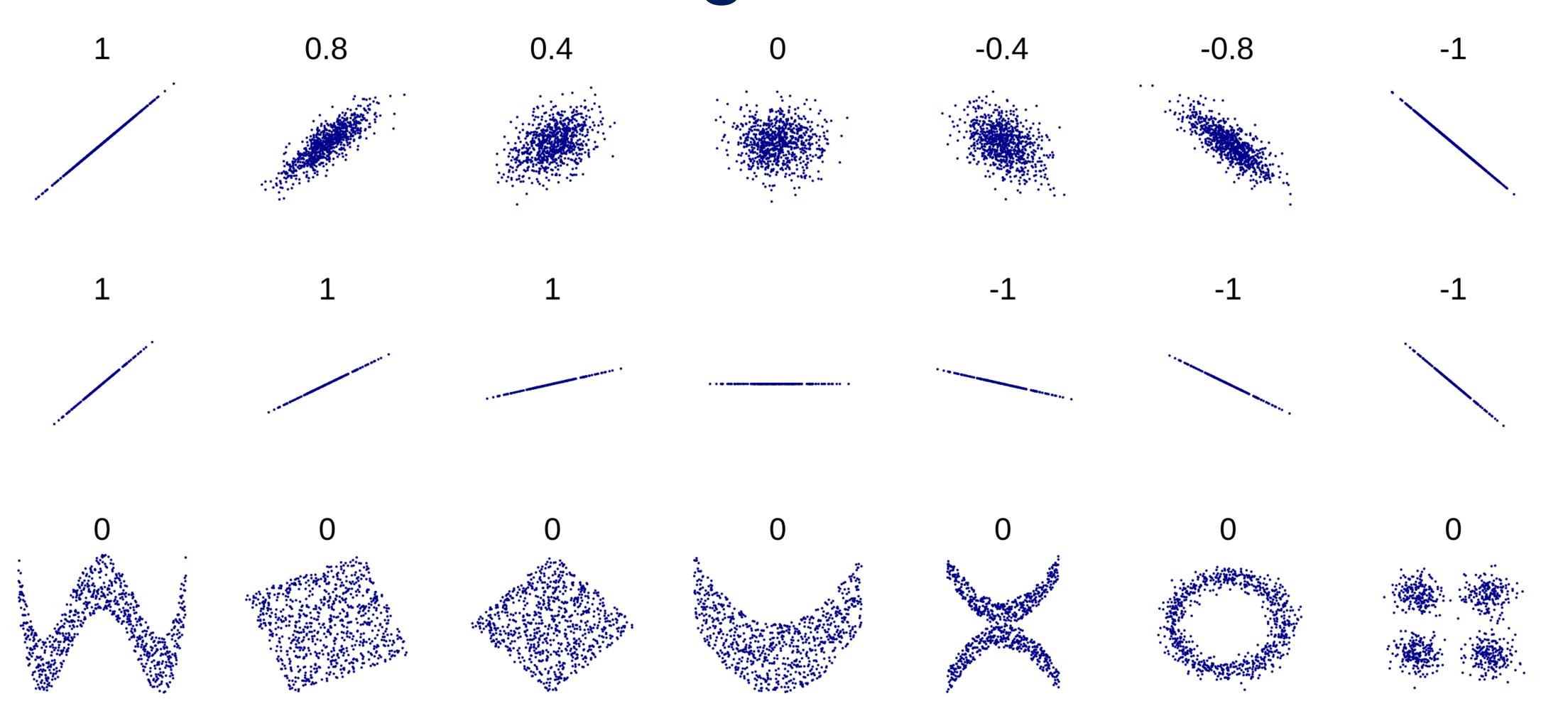


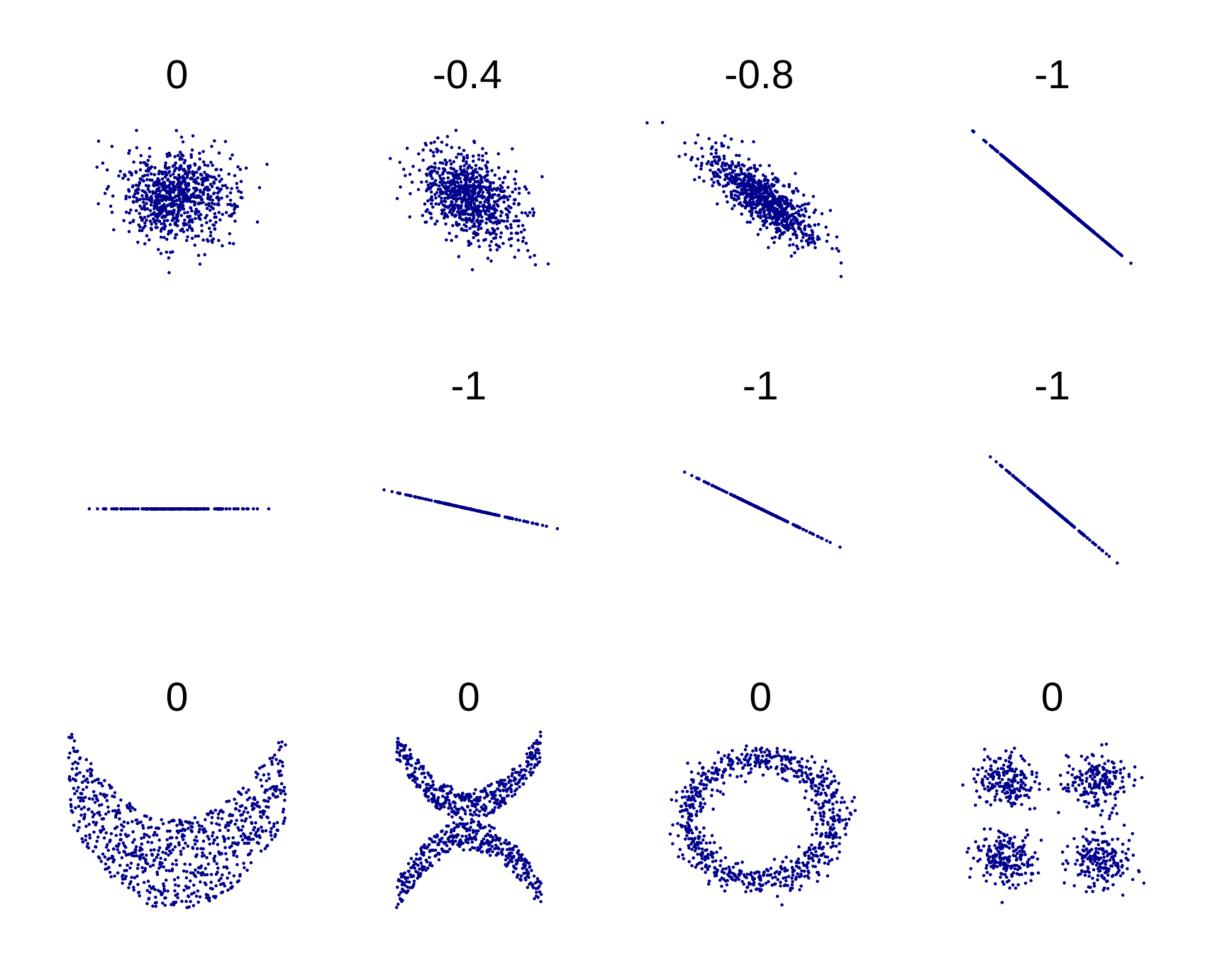
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- General intuition
 - Positive correlation: as X increases, so does Y
 - Negative correlation: as X increases, Y decreases

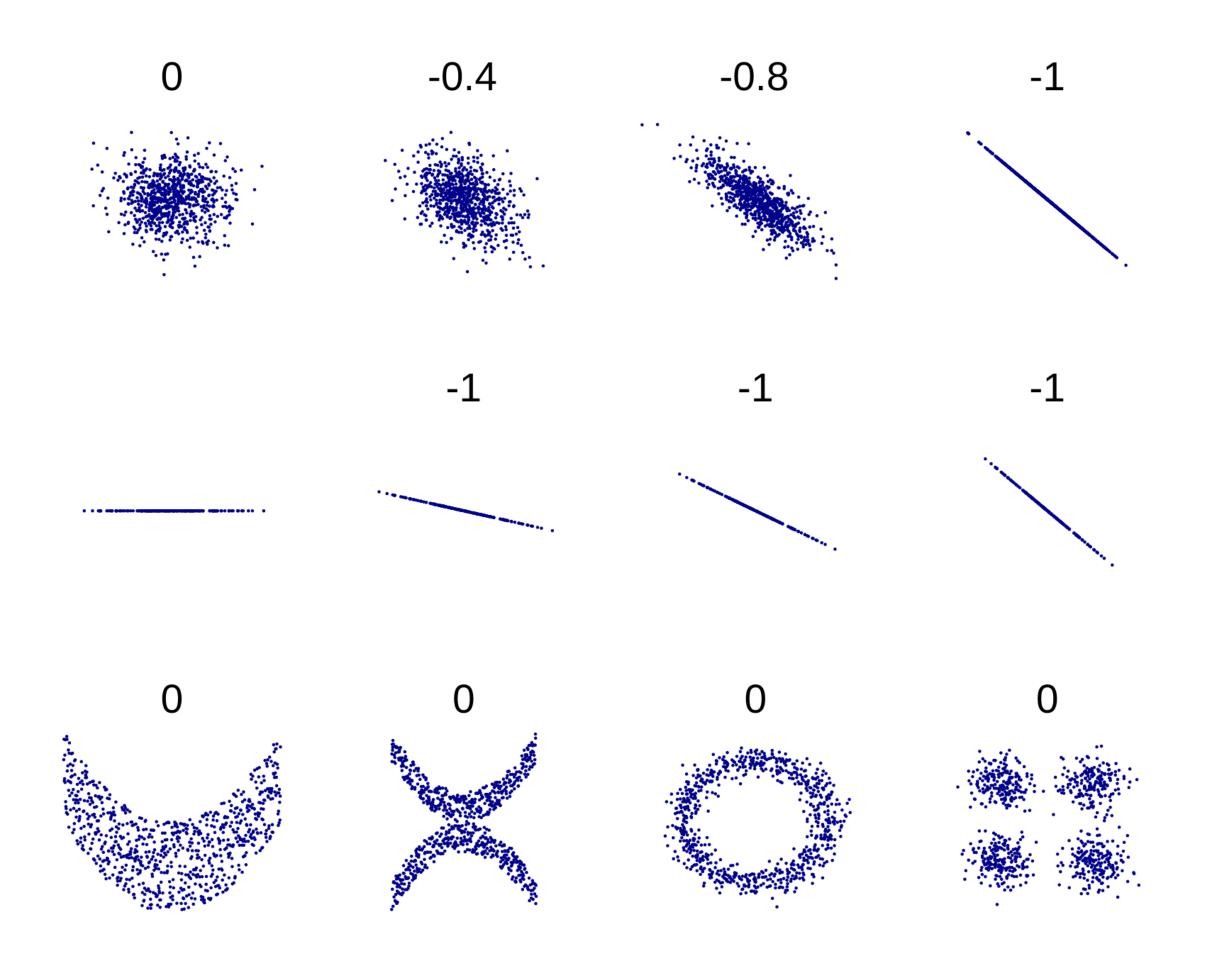
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- General intuition
 - Positive correlation: as X increases, so does Y
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- Correlation is expressed as a coefficient (a number)
 - Most commonly used is Pearson's Correlation Coefficient (written r)
 - Based on another measure called covariance

Visualizing correlation

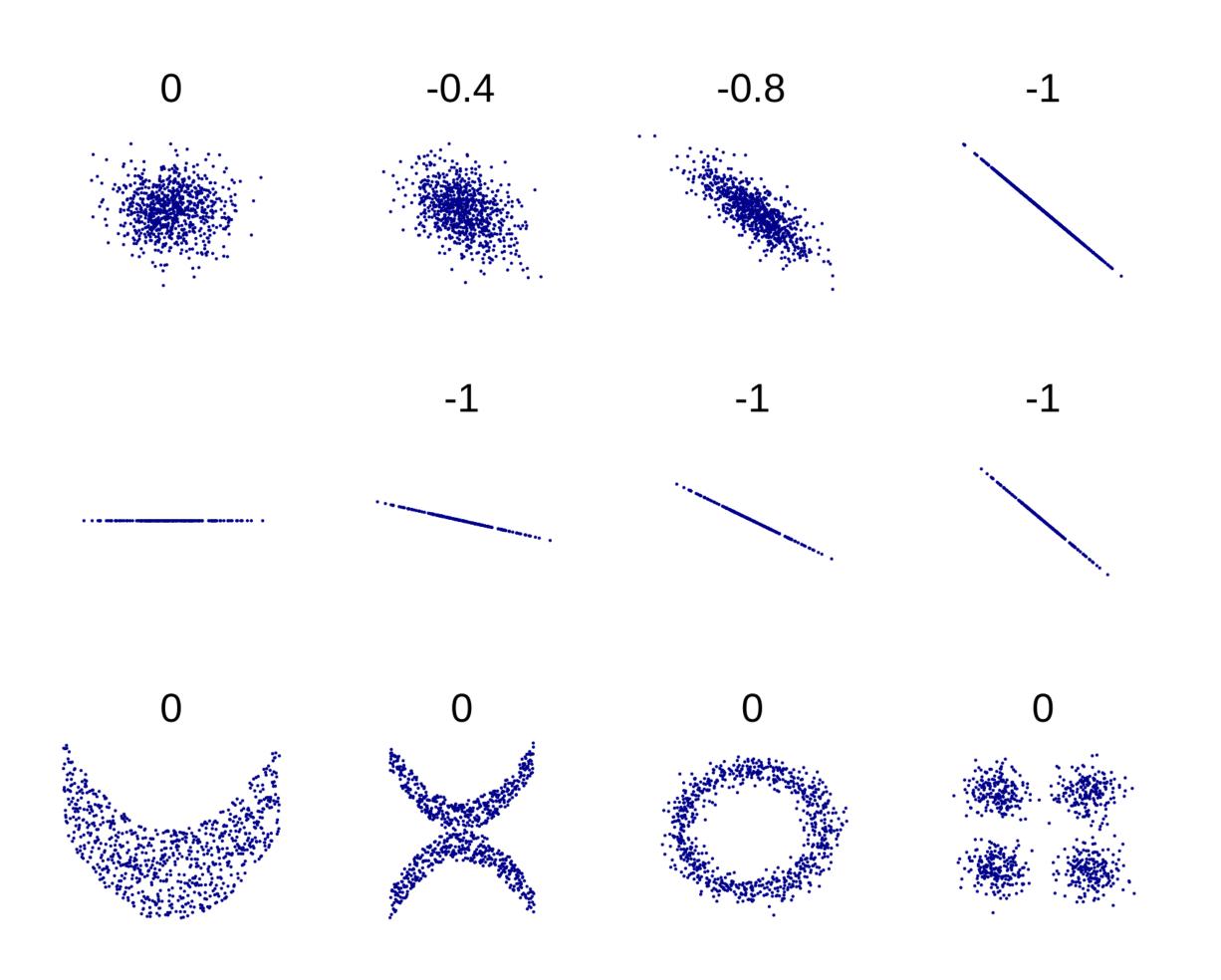




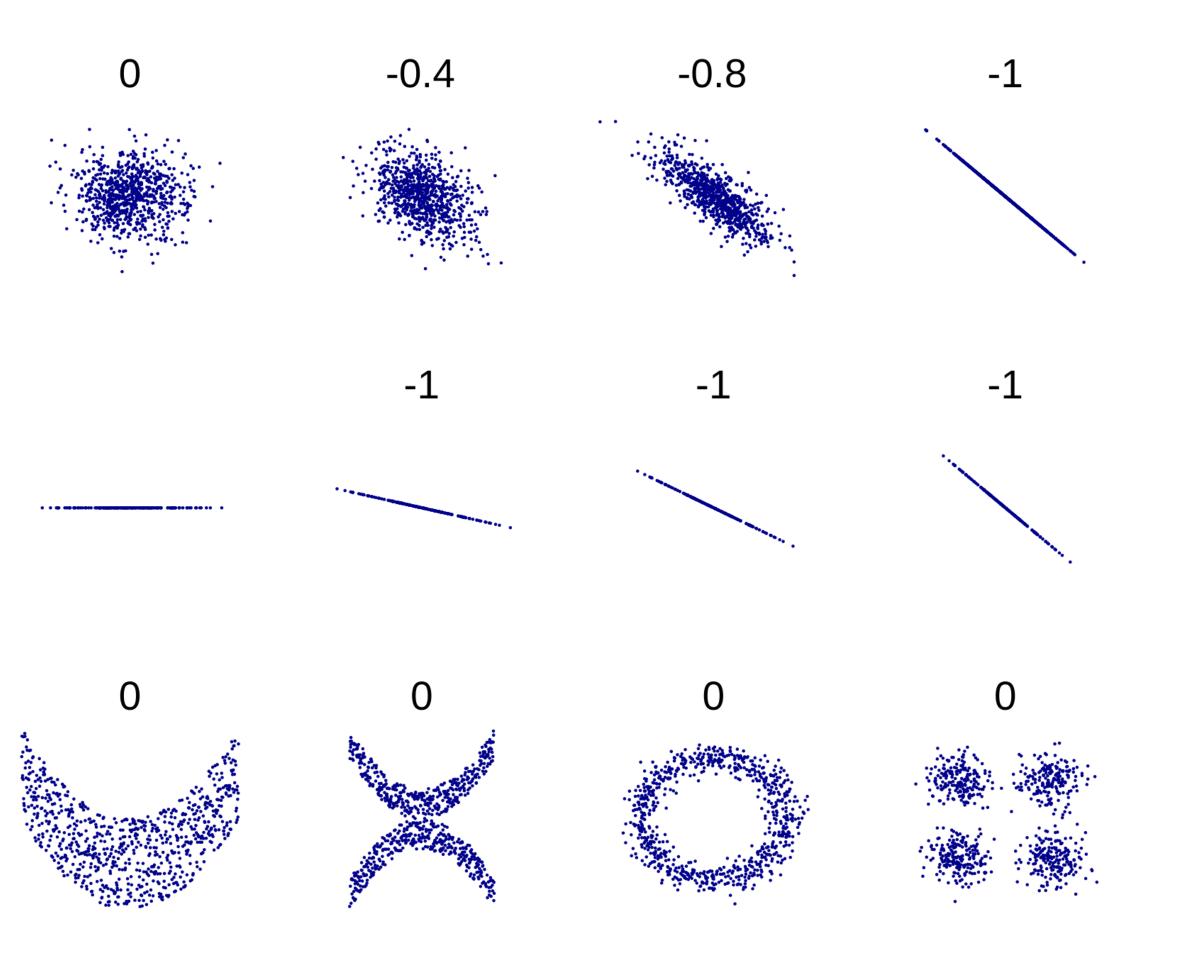
• The coefficient's range is -1 to +1



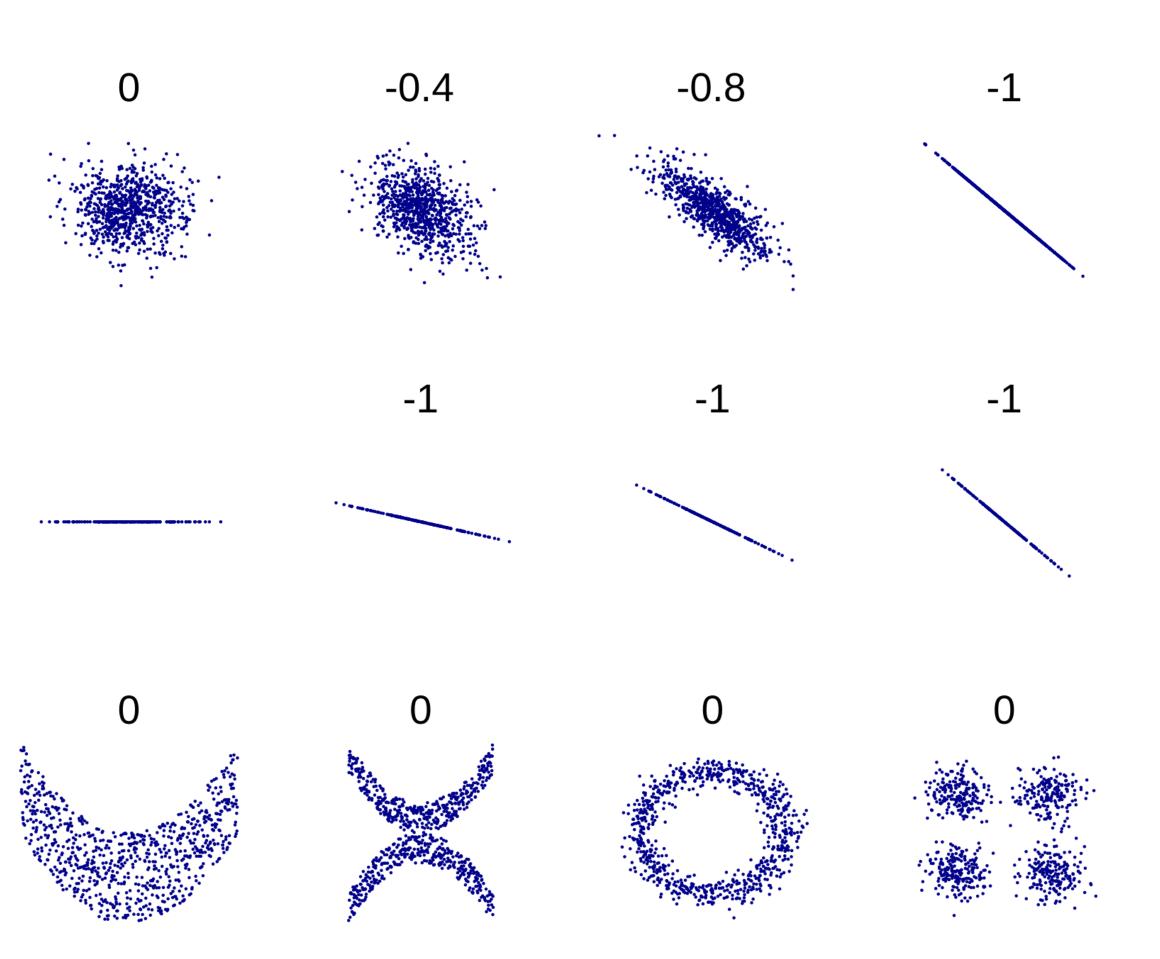
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- Data can have structure without correlation



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- The covariance has no minimum/ maximum value

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- R command: cor(x, y)

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Correlation	Strength	Direction
-1.0 to -0.9	Very strong	Negative
-0.9 to -0.7	Strong	Negative
-0.7 to -0.4	Moderate	Negative
-0.4 to -0.2	Weak	Negative
-0.2 to 0	Negligible	Negative
0 to 0.2	Negligible	Positive
0.2 to 0.4	Weak	Positive
0.4 to 0.7	Moderate	Positive
0.7 to 0.9	Strong	Positive
0.9 to 1.0	Very strong	Positive



- Interpreting a coefficient depends heavily on context
 - In some fields/situations, r < 0.95 is considered "weak"
 - In others, r > 0.3 is considered "strong"

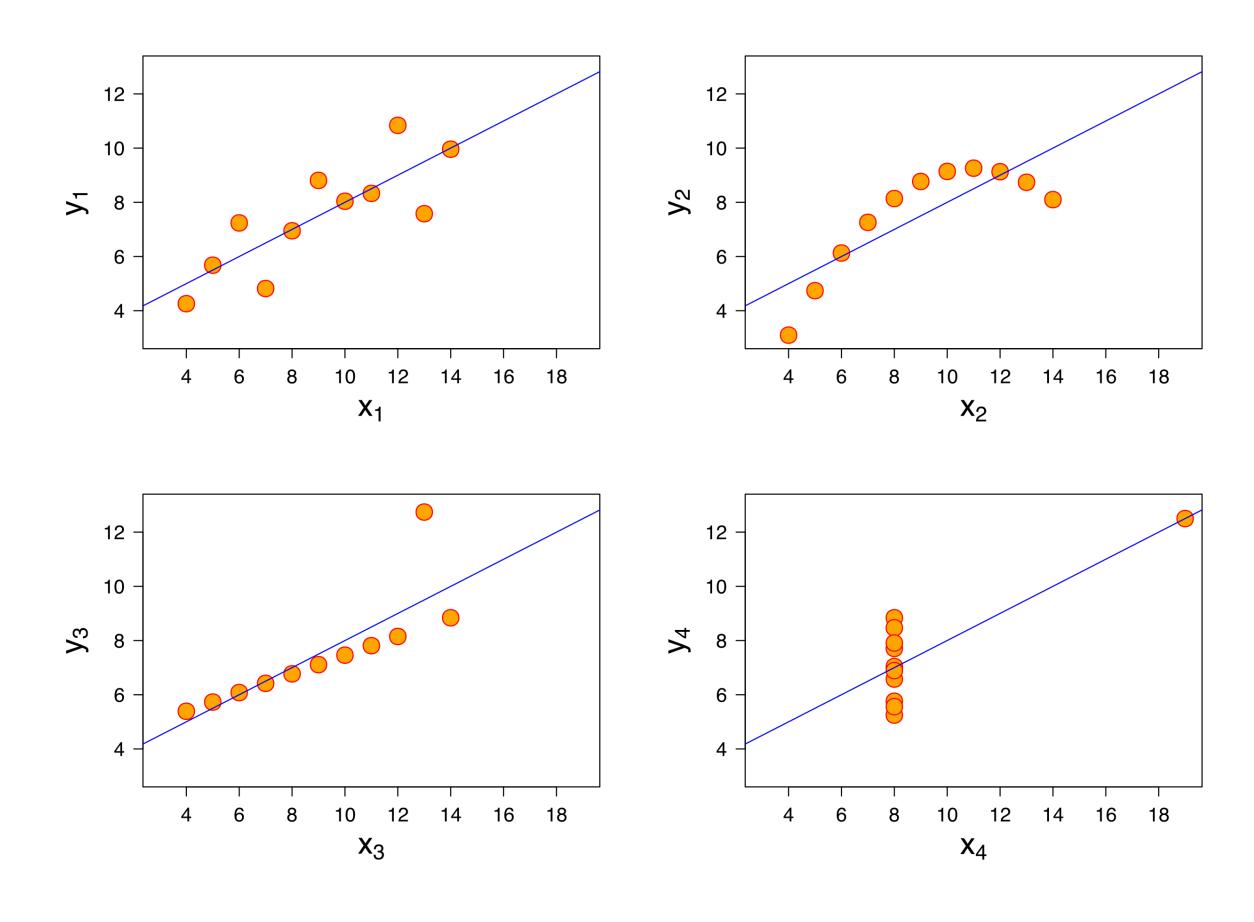
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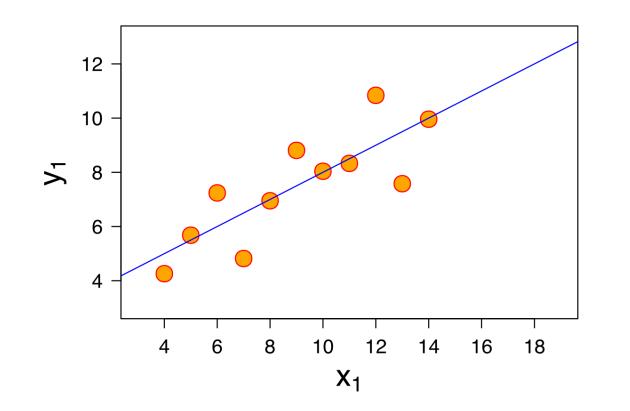
- Interpreting a coefficient depends heavily on context
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- Strong correlation does not mean "statistical significance". It's just a measurement

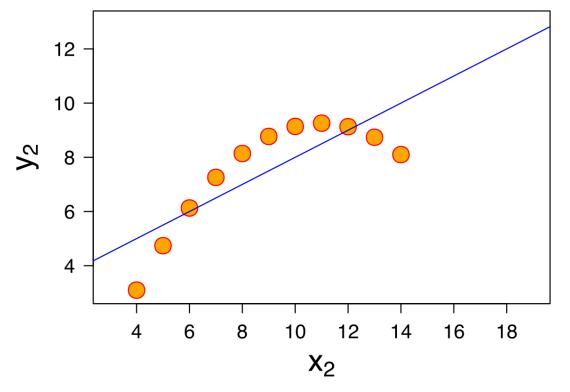
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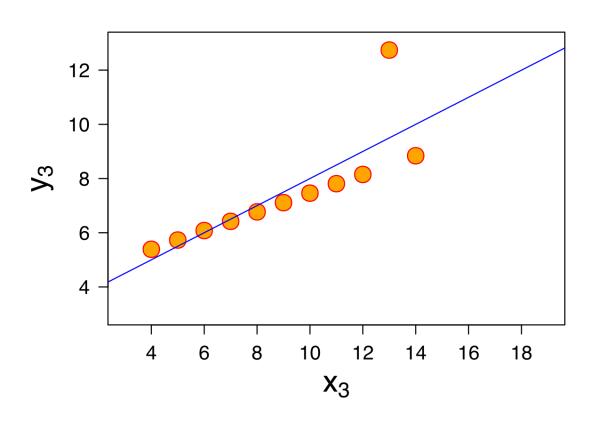


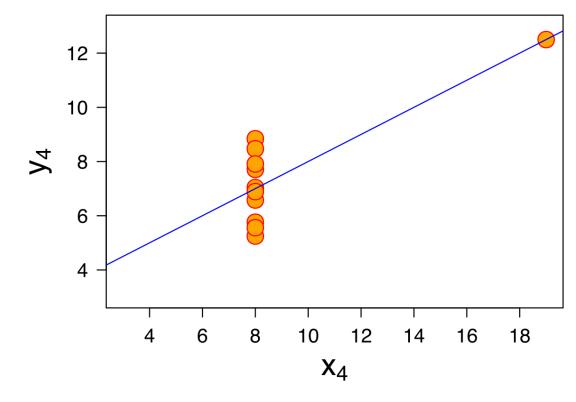


- "Anscombe's quartet" are four datasets with the exact same correlation
 - Shows that correlation does not give all important aspects of the data

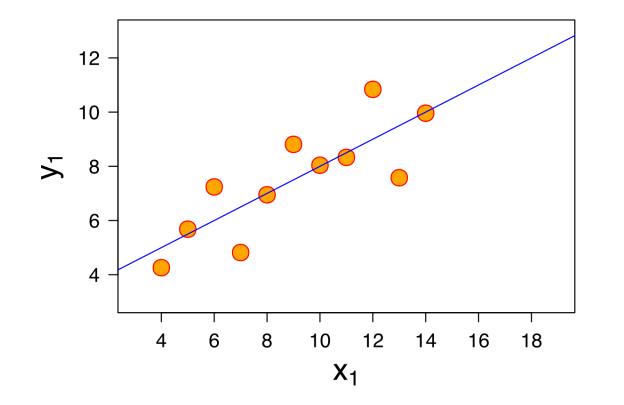


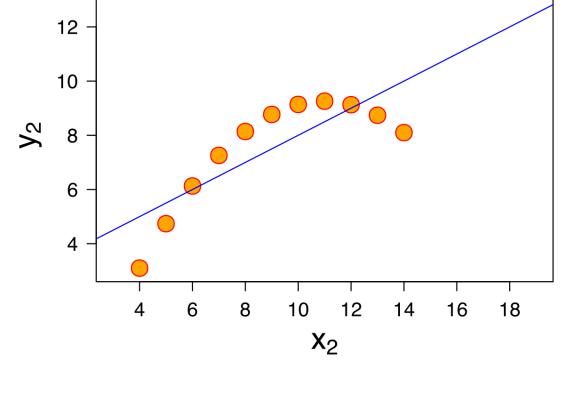


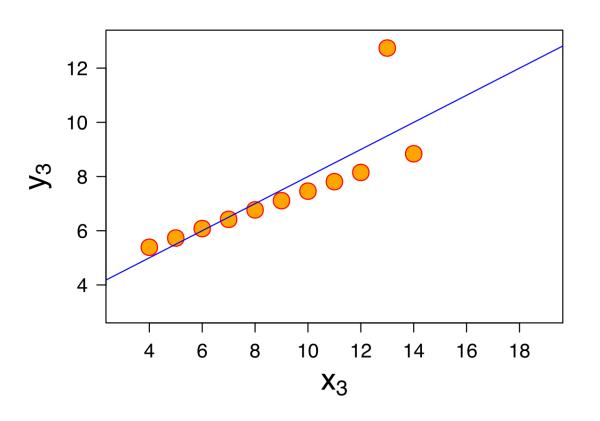


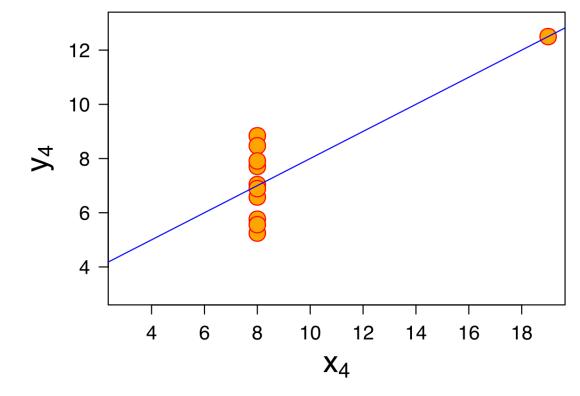


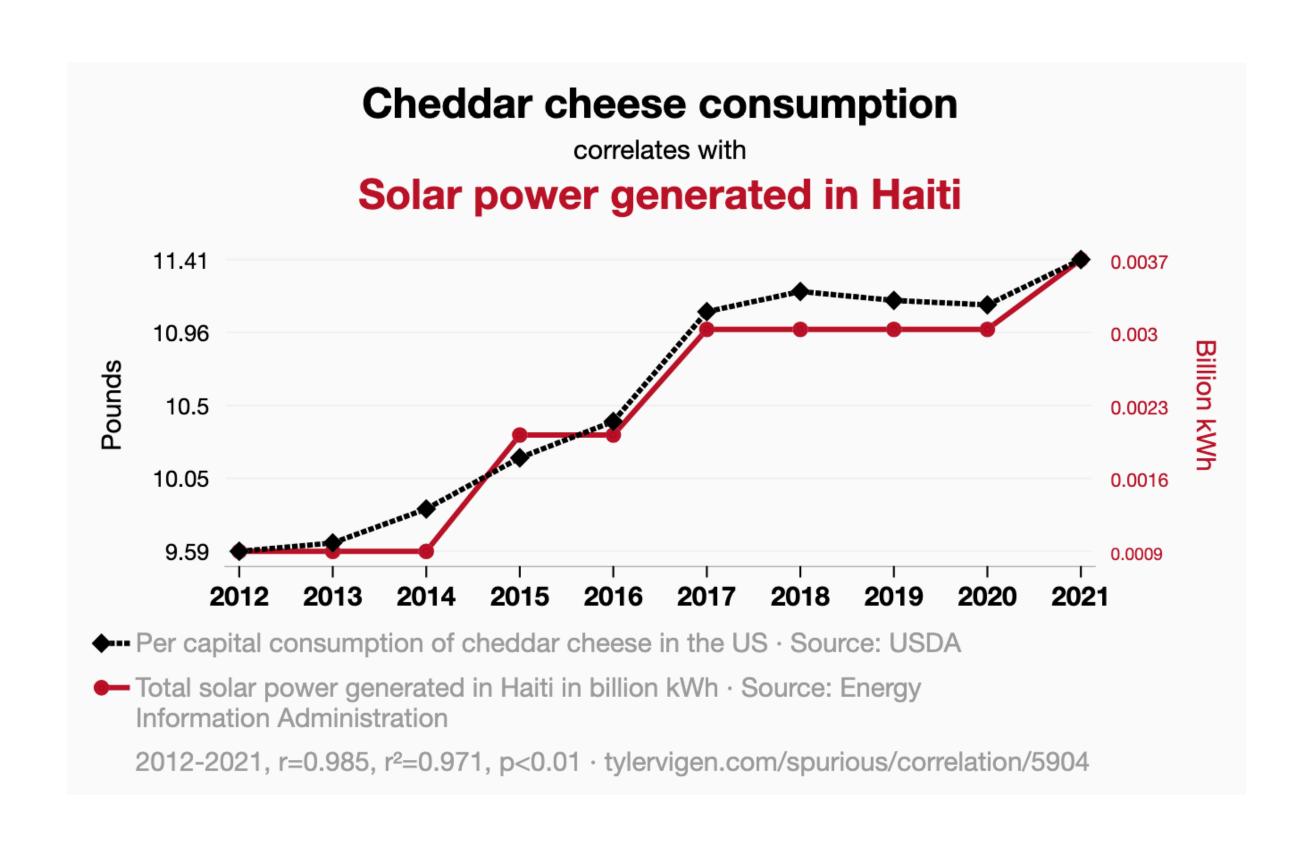
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 - Shows that correlation does not give all important aspects of the data
- Make sure to always use
 visualizations in combination with
 descriptive statistics



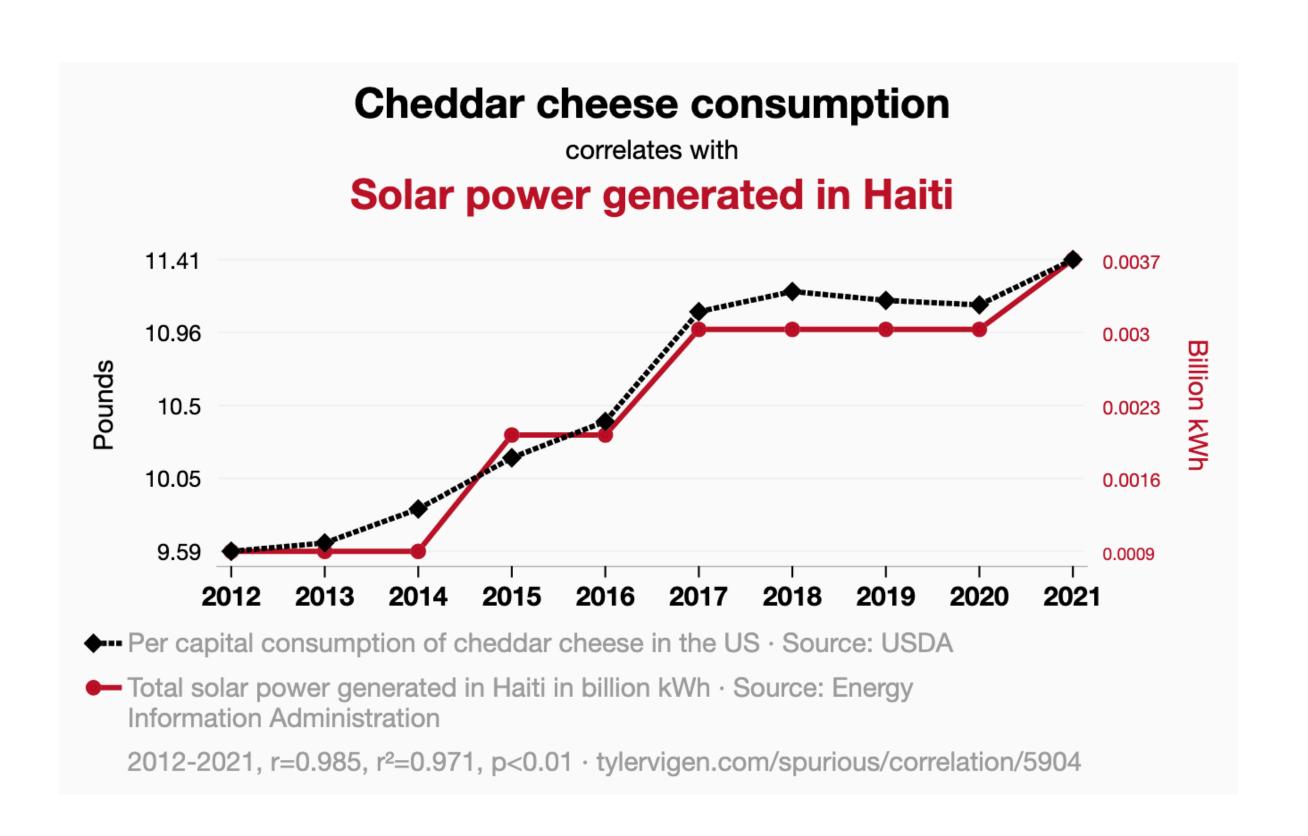




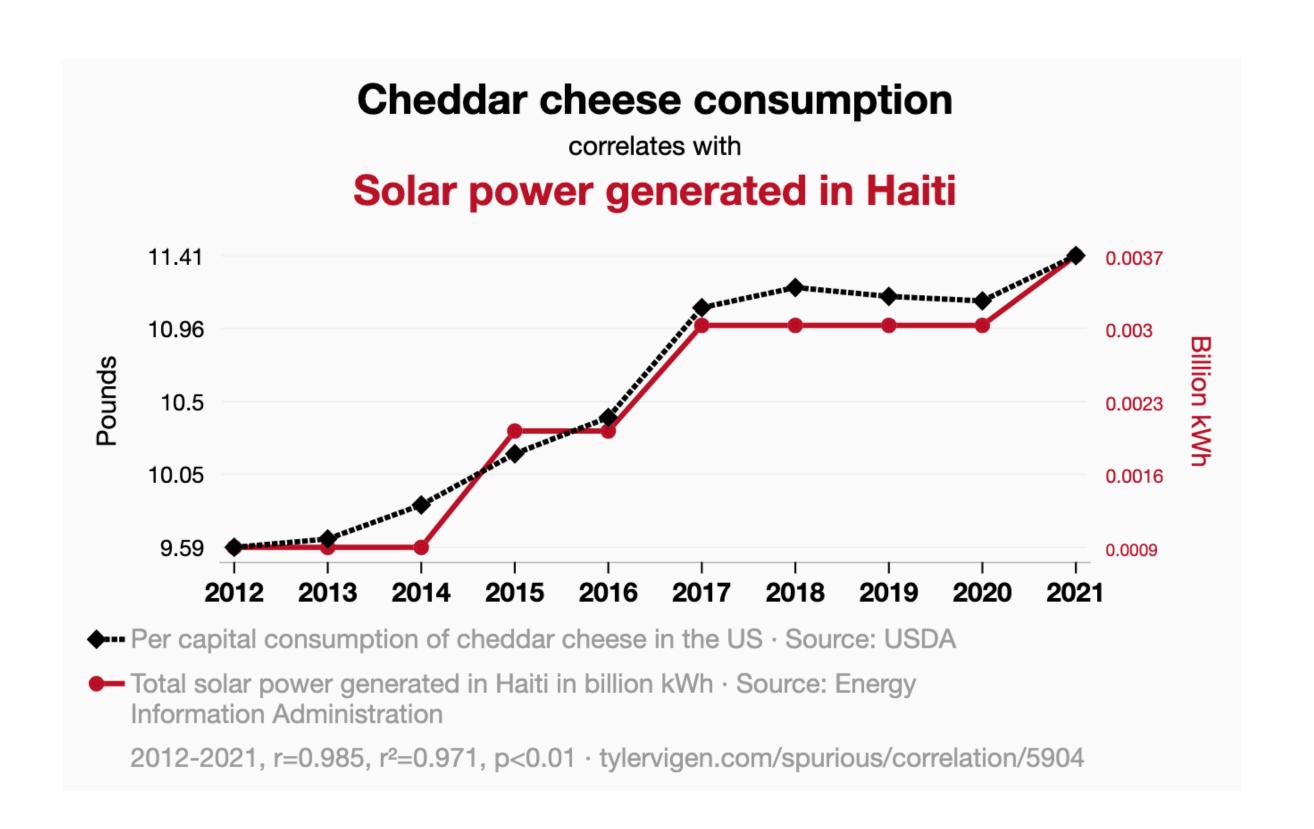




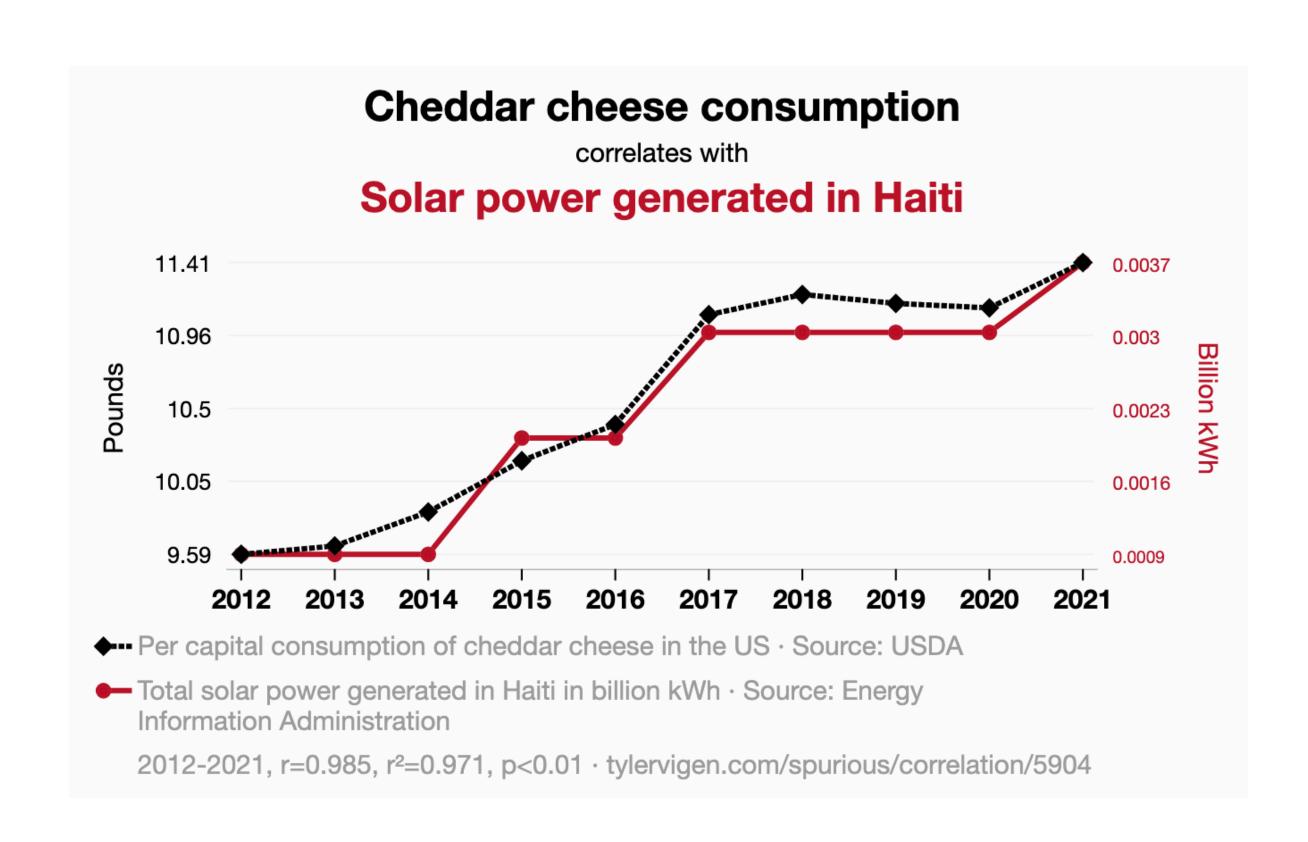
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 - X might cause Y; Y might cause X
 - or neither!
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- Lots of examples of funny spurious correlations (variables that inexplicably correlate)
- Browse some at this website



Correlation matrix

- If you call cor() on a data frame, it gives the correlation matrix
 - i.e. the correlation between all pairs of variables
 - Note that every variables is perfectly correlated with itself

```
> just_numerical_columns = data.frame(vowels$F1, v
owels$F2, vowels$HEIGHT)
> head(just_numerical_columns)
 vowels.F1 vowels.F2 vowels.HEIGHT
   848.070 1450.96
                               173
   648.318 1126.22
                              173
   259.000 1834.00
                              173
4 578.985 1715.22
                              173
   405.000 1899.00
                              173
   656.600 1414.40
                               173
> cor(just_numerical_columns)
              vowels.F1 vowels.F2 vowels.HEIGHT
vowels.F1
              1.0000000 -0.1488059
                                      -0.1996378
vowels.F2
             -0.1488059 1.0000000
                                      -0.1426429
vowels.HEIGHT -0.1996378 -0.1426429
                                      1.0000000
```