# Notes about correlation (for Asgn 2)

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# **Overview of assignment**

Exploration of distributional similarity.

- Work with data extracted from Twitter (co-occurrence counts)
- Compare different ways to contruct context vectors and compute similarities
- Analyze and discuss differences between approaches, qualitatively and quantitatively.

Work through the lab **before** you start the assignment!

## Qualitative and quantitative analysis

Assignment asks you to do some of each.

- Examples of qualitative analysis:
  - Using visualization to illustrate/discuss examples or trends
  - Discussing one or a few examples in more detail, by looking at our dataset and/or other Tweets (e.g., use the Twitter search page).
- Examples of quantitative analysis:
  - Often: numerical comparison to a gold standard of accuracy
  - Here: consider other options, such as correlating similarity measures against word frequency.

# One kind of quantitative analysis

- Assignment spec suggests you may want to consider **correlation** between similarity measures and word frequency.
- Why?
  - A good similarity measure should measure (only) similarity.
  - So presumably *not* be correlated with frequency.
  - Unless more frequent words really are more similar to each other! (Would need to test with humans... let's assume not)

#### What is correlation?

- Intuitively: two random variables X and Y are correlated if, when the value of X increases, the value of Y also tends to increase (positive correlation) or decrease (negative correlation).
- Often, X and Y are different measurements for each data point.
  - A person's height X and weight Y
  - A word's frequency  $\boldsymbol{X}$  and length  $\boldsymbol{Y}$
- Two standard ways to measure correlation:
  - Spearman (rank) correlation: roughly as above.
  - Pearson (linear) correlation: more specific.

#### **Pearson correlation**

- Mathematically: the covariance of X and Y, normalized by the product of their individual standard deviations.
- Intuitively: if I plot X against Y, how close to a perfect linear relationship do I see?
  - Does not measure the *slope* of the line, just whether there is one. (Compare rows 1 and 2, next page.)
  - Does not tell us if there's some other *non-linear* relationship between X and Y. (See row 3, next page.)
- For data samples, the Pearson correlation coefficient is usually denoted r.

#### **Pearson correlation**

Examples datasets with Pearson r values shown:



Image source: https://commons.wikimedia.org/wiki/File:Correlation\_examples.png

## **Spearman rank correlation**

- Mathematically: compute the Pearson correlation between the rank ordering of X and Y values.
- Intuitively: how close to a perfectly monotonic relationship do X and Y have? (i.e., when X increases, Y increases)
- $\bullet$  For data samples, the Spearman rank correlation coefficient is usually denoted  $\rho$  or  $r_s.$

#### **Spearman correlation**

Data with perfect rank correlation, but not perfectly linear:



Image by Skbkekas (CC-BY-SA 3.0)

https://en.wikipedia.org/wiki/Spearman\%27s\_rank\_correlation\_coefficient

#### Which one to use?

- If correlation is roughly linear, Pearson will normally yield stronger results (larger absolute values)
  - If hypothesis testing against the possibility of no correlation, likely to have higher significance level than Spearman.
  - But if using large samples from corpora, often nearly any result is clearly "non-zero". We may care more about the actual degree of correlation.
- If correlation is non-linear, or nothing is known, use Spearman.

## But usually we do know something

Best to look at the data first! For example, word freq vs length: Seems follow to а pattern, but not strongly 12000 linear. Indeed, 10000 8000 Frequency • Spearman:  $\rho = -0.18$ 6000 • Pearson: r = -0.104000 2000 (Note: I "jittered" the data so those with same 5 10 15 25 30 20 0 (x,y) are not right on top Length

of each other.)

# Log frequency

Of course, using log frequencies is often more sensible:

We now have

- Spearman:  $\rho = -0.18$
- Pearson: r = -0.21

Notice that  $\rho$  is not affected by rescaling the data. r is higher, but still only a weak linear correlation.



## So, which one to use?

- So, Pearson can still work if there is an obvious tranformation to make the correlation roughly linear.
- But if in doubt, usually fine to use Spearman.
- As with all statistics, many subtleties if using for really careful analysis (see statistics course or online tutorials), but what I've said is probably enough for exploratory studies (i.e., your assignment).