

## Homework Assignment: Normalizing Word Frequencies by Document Length

### Background

In Week 02, we compared word frequencies across two texts using **raw counts**. We found that “trade” appears 233 times in *The Circle of Commerce* (Text A) and 185 times in *Free Trade* (Text B). But does this mean Text A talks about trade more? Not necessarily—Text A might simply be longer!

Raw counts can be misleading when comparing documents of different lengths. To make fair comparisons, we need to **normalize** our word frequencies.

### How to Submit:

Post your code on your GitHub. On **Canvas**: post your response to the interpretive questions in the Discussion section (Week 2: Basics). Make sure to include your visualization **and a link to the code**. Finally, if Canvas, GitHub, or something similar is not functioning for this first assignment: don’t panic! We will figure it out.

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### Your Task

You will modify the Week 02 word frequency analysis to account for document length by calculating **relative frequencies** (proportions).

We will break this down into steps:

#### I. Create a diagnostics table (before stopword removal)

Add a new section **immediately after** you create texts, before word\_counts <- ....

Create a tibble called corpus\_diagnostics with one row per document that includes:

- doc\_title
- n\_chars = number of characters in each document
- n\_word\_tokens = number of word tokens *before* stopword removal
- n\_word\_types = number of unique word types *after* lowercasing, *before* stopword removal

Your diagnostics must be computed using the tidy workflow used on the website (i.e., unnest\_tokens() for word tokens; then str\_to\_lower(); then a summary that yields counts). unnest\_tokens() is the key function for converting a “one row per document” tibble into “one row per token.”

## II. Interpret the diagnostics (short prose to be shared on Canvas)

In **4–6 sentences**, answer:

- Are Text A and Text B comparable in length? (Use your diagnostics numbers.)
- If they differ substantially, what does that imply for interpreting **raw frequency** comparisons?

### Summary for you:

The corpus integrity check exists to ensure that:

- you know what your corpus contains,
- your comparisons are methodologically sound,
- and your interpretations are constrained by evidence rather than assumptions.

It is not a technical hurdle—it is an **epistemic safeguard**.

## III. Compare normalized “trade” across the texts

Using the word\_counts tibble from Week 02, calculate the total number of words (after stopword removal) in each document.

### Steps:

1. Group by doc\_title
2. Sum the word counts to get total words per document
3. Store this in a tibble called doc\_lengths

**Hint:** Use group\_by() and summarise() with sum().

Add a column to word\_counts that shows each word's frequency as a **proportion** of the total words in its document.

### Steps:

1. Join word\_counts with doc\_lengths
2. Calculate: relative\_freq = n / total\_words
3. Store the result in word\_counts\_normalized

**Hint:** Use left\_join() and mutate()

**Then,** Using your normalized frequencies:

1. Filter for the word "trade" in both texts

## 2. Compare the **raw counts** vs. **relative frequencies**

Recreate the Week 02 word frequency visualization but using **relative frequencies** instead of raw counts.

### Requirements:

- Use the same top 20 words as Week 02
- Show relative frequencies (not raw counts) on the x-axis
- Keep the side-by-side faceted layout
- Update axis labels and title to reflect normalization

**Hint:** Adapt the Week 02 plotting code, replacing n with relative\_freq and updating labels.

### Answer these questions (short prose to be shared on Canvas):

- Does Text A or Text B use "trade" more *proportionally*? And how does this compare to what the raw counts suggested?
- We normalized by dividing each word count by the total words in that document (after stopword removal). How would your results change if you normalized by the *original* document length (before stopword removal)? Would this be better or worse, and why? [This is a harder question than it would seem at first! Review the lecture notes].