

CA – Assignment 1: Data Acquisition

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How to reproduce the unified data file:

The unified data is saved in file: `/data/unified_data.json`. To reproduce the file, run the `data-unification.py` in the `code` folder with the following command:

```
python data-unification.py
```

How to run the preliminary statistics:

The methods to calculate the preliminary statistics are located in the file `statistics.py`. If you want to run the file, switch to the `code` folder and execute the `statistics.py` file with the following command:

```
python statistics.py
```

Otherwise we also added a jupyter notebook file `preliminary-statistics.py`. The jupyter notebook can be opened by the following command

```
jupyter notebook statistics.ipynb
```

Explanation of the method to compute the most specific words of each of the argument units:

To calculate the **most specific words** for the three different argument units, we used the **IF-IDF** score. IF-IDF stands for **Term Frequency – Inverse Document Frequency** and is often used in the information retrieval and text mining. The IF-IDF score/weight is a statistical measure used to evaluate how important a word is to a document in a collection or corpus.

The importance increases proportionally to the number of times a word appears in the document but is offset by the frequency of the word in the corpus ([TFIDF.com](https://tfidf.com)).

It also can be successfully used for stop-words filtering and other words that appears very often. In our case if a word appears very often in one argument unit but also in other argument unit(s). Therefore we calculate a score for each word based on its relevance and frequency for each of the argument units (major claims, claims and premises).

TF-IDF calculation is defined by

$$\text{IF-IDF} := \text{TF} * \text{IDF}$$

whereas TF and IDF are defined as:

$$\text{IF}(t) := \frac{\text{(number of times term } t \text{ appears in a document)}}{\text{(Total number of terms in a document)}}$$

and

$$\text{IDF}(t) := \log_e \left(\frac{\text{Total number of documents}}{\text{Number of documents with term } t \text{ in it}} \right)$$

So in our application:

$$\begin{aligned} \text{Total number of documents} \\ &= \text{total number of essays} \\ &= 322 \end{aligned}$$

And

$$\begin{aligned} \text{Number of documents with term } t \text{ in it} \\ &= \text{number of essays with term } t \text{ in it.} \end{aligned}$$

So for example if a **word** *w* appears very often in an **argument unit** *a*, it leads to a high IF score. But if it also appears very often in an other **argument unit** *a'*, the IDF score will be very low. Therefore the IF-IDF score will be very low in contrast to other words. Consequently we get only a high IF-IDF score, if the frequency of the **word** *w* is very high and the **word** *w* is not used very often in other argument units. Therefore, we have calculated the **most specific words** which are very important for each argument unit and also very specifically used for only that argument unit.