

Natural Language Processing Project Phase 1

Human Values Behind Arguments

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1 Introduction

This is a classification task based on SemEval 2023 Task 4. Its objective is to identify human values behind arguments. To be more clear, given a text and some value categories, here we want to identify which category it falls in and also is it against or in favor of that category. You can access the project codes from here, and also all the data has uploaded to hugging face accessible via this link

2 Dataset

Initially there was a dataset of 5393 annotated arguments. There was 20 values which i chose only 3 of them:

Achievement

• Power: dominance

• Power: resources

After extracting the values which we want to classify, there remains 2188 arguments.

2.1 Data Collection

Using requests in python, I gathered needed data from the dataset provided.

2.2 Data Structure

The annotated corpus in tab-separated value format. Contains the following files for the different dataset splits: arguments-<split>.tsv: Each row corresponds to one argument Argument ID: The unique identifier for the argument Conclusion: Conclusion text of the argument Stance: Stance of the Premise towards the Conclusion; one of "in favor of", "against" Premise: Premise text of the argument labels-<split>.tsv: Each row corresponds to one argument Argument ID: The unique identifier for the argument Other: Each other column corresponds to one value category, with a 1 meaning that the argument resorts to the value category and a 0 that not

2.3 Data Filtering

In this state, I joined the arguments and labels by the Argument ID. I deleted all columns and rows which were related to other values. For each of the remaining values, there is two labels: 'against' and 'in favor of' which I showed by adding 'N' or 'P' at the end of the value name.

2.4 Data Cleaning

For cleaning data, I removed all punctuation marks except dot. The reason for that is it makes the sentence and word tokenizing easier.

2.5 Data Breaking

After cleaning the data, I broke data by its sentence. You can find the label separated files inside data/sentencebroken directory. I also broke data by its words which you can find the label separated files inside data/wordbroken directory.

3 Statistics

These are some basics statistics about the dataset.

3.1 Row Count

Number of rows for each label.

row count

Against Achievement 729

In favor of Achievement 783

Against Power: dominance 310

In favor of Power: dominance 300

Against Power: resources 319

In favor of Power: resources 306

3.2 Sentence Count

Number of sentence for each label.

sentence count

Against Achievement 970

In favor of Achievement 1222

Against Power: dominance 426

In favor of Power: dominance 497

Against Power: resources 412

In favor of Power: resources 433

3.3 Word Count

Number of words for each label.

word count

Against Achievement 15397

In favor of Achievement 19222

Against Power: dominance 6622

In favor of Power: dominance 7793

Against Power: resources 6773

In favor of Power: resources 7045

3.4 Unique Word Count

Number of unique words for each label.

unique word count

Against Achievement 2782

In favor of Achievement 3465

Against Power: dominance 1642

In favor of Power: dominance 2066

Against Power: resources 1535

In favor of Power: resources 1730

3.5 Common Unique Word Count

Number of common unique words for each label. Extra explanation: I computed the common words between all labels, then count the number of unique words for each label which are present in common words.

common unique word count

Against Achievement 416

In favor of Achievement 416

Against Power: dominance 416

In favor of Power: dominance 416

Against Power: resources 416

In favor of Power: resources 416

3.6 Uncommon Unique Word Count

Number of uncommon unique words for each label. Extra explanation: I computed the common words between all labels, then count the number of unique

words for each label which are not present in common words.

uncommon unique word count

Against Achievement 2366

In favor of Achievement 3049

Against Power: dominance 1226

In favor of Power: dominance 1650

Against Power: resources 1119

In favor of Power: resources 1314

3.7 10 Most Frequent Uncommon Words

10 most frequent uncommon words for each label. Extra explanation: I computed the common words between all labels, then count the number of unique words occurrences for each label which are not present in common words, and sort them by number of occurrences to get the 10 most frequent uncommon words.

word 1 word 2 word 3 word 4 word 5 word 6 word 7 word 8 word 9 word 10 word1 word2 word3 word4 w

Against Achievement telemarketing cloning farming jobs food cars factory child actors autonomous

In favor of Achievement students wikipedia loans stem exploration cell embryonic young learn cars

Against Power: dominance company vote deterrent democracy killing judicial voting activism war threat

In favor of Power: dominance voting compulsory policy army democracy I political vote multi-party part

In favor of Power: resources debt games prostitution expensive resources olympic unfair loans trade stude

Against Power: resources jobs company telemarketing exploration food farming factory whaling executive

3.8 Words Histogram

In this words histogram, you see 30 of the most frequent words in all labels.

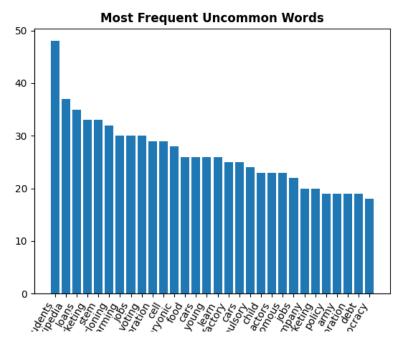


Figure 1: Most Frequent Words

4 Run Script

I wrote a python script which allows any user to run each part of this project separately.