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# CA – Assignment 2: Argument Mining

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## Structure

```
. argument-mining-assignment

— Documentation.pdf

— README

— requirements.txt

— code

— conf_bias_evaluation.py

— model.py

— data

— essay_corpus.json

— predictions.json

— sample_prediction.json

— train-test-split.csv
```

#### Scripts

- essay\_corpus.json: Data corpus created in Data Acquisition assignment.
- model.py: The ML model that we use for generating predictions.
- conf\_bias\_evaluation.py: Script to evaluate the F1 score of the ML model.

### How to run the scripts

- On a venv install the requirements specified in requirements.txt
- Make sure you have the same directory structure as above otherwise adjust the paths in the model.py script accordingly.
- Run model.py to generate the predictions in data/ directory with name predictions.json
- Run conf\_bias\_evaluation script with the filepath to the data\.

## Model Selection

We selected SVM as the model for mainly these 3 reasons:

• In the paper Recognizing the Absence of Opposing Arguments in Persuasive Essays by Stab and Gurevych they used SVM for doing the identical task and the scores were pretty high.

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We read the paper titled: Text categorization with Support Vector Machines:
 Learning with many relevant features[1]. In the paper the author explores why SVM are suited for text-classification tasks.

- Even then we tried multiple other models: Logistic Regression, Naive Bayes, Random Forrest and the performance of SVM was the best.
- We used Linear SVM because text classification is mostly a linearly seperable problem[1] and using kernels(rbf, poly) to map the data to a higher dimensional space did not really improve the performance in this case.
- We used GridSearch for hyperparameter tuning.

[1]: https://link.springer.com/chapter/10.1007/BFb0026683

## Feature Selection

For features we used the following:

- n-grams + TF-IDF: In the range of 1-3 so that unique keywords and phrases are identified and given more importance. We used TfidfVectorizer of scikit-learn library to create n-grams.
- Adversative transitional phrases (adv): to identify conflict, contradiction, concession, and dismissal in the text which indicate presence of opposing argument. We first used all 47 adversative transitional phrases that are grouped in the following categories:
  - concession (18)
  - conflict (12)
  - dismissal (9)
  - emphasis (5)
  - replacement (3)

For each of these categories, we added features for the upper and the lower case as well as for their presence in the surrounding paragraph (introduction+conclusion or in the body)but the results were even worse than just the approach with only n-grams + TF-IDF. So we did a deeper analysis of training data by finding the occurances of phrases in both the true and false classes (using adv\_trans\_text\_analysis function in the code). As a result, we detected that the concession and conflict phrases are the best indicator for the opposing arguments in our training data. Consequently, we just used 15 phrases of these categories to get an F1 score of .754.

• We used 10-fold cross-validation on the training data for regularization.