

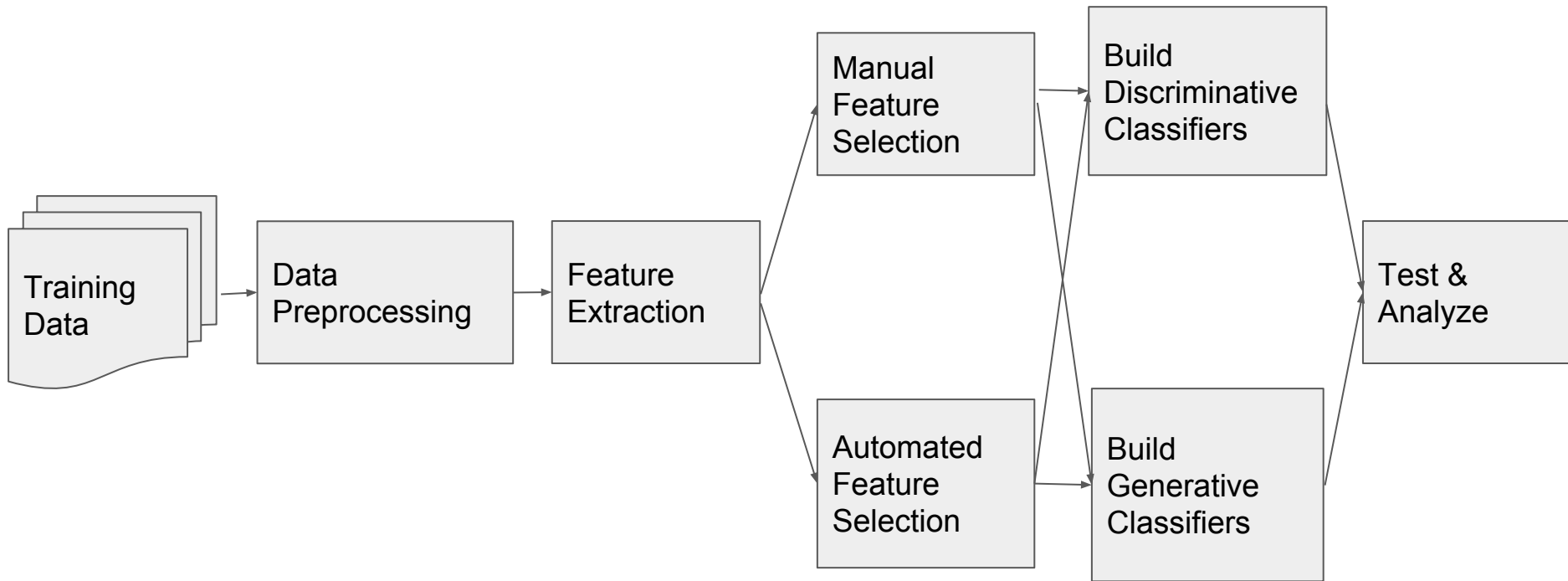
Authorship Classification

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Problem Definition

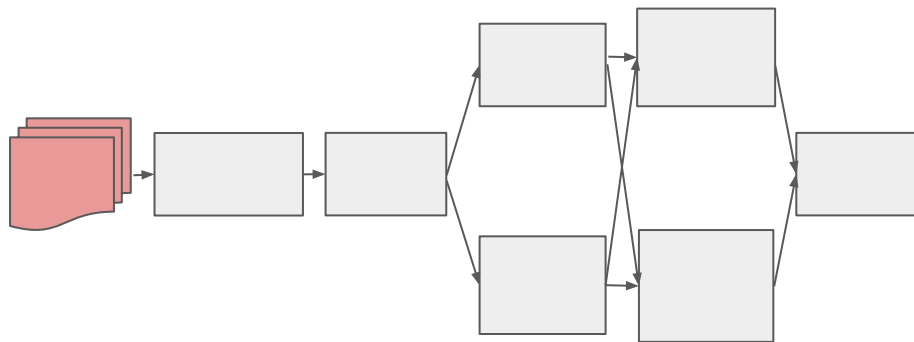
- Authorship classification has been a long studied problem in the domain of linguistics
- Existed long before computers
 - 1400s *Donation of Constantine*
 - Federalist Papers
 - Shakespeare's Plays
- Current computational approaches exist, but vary tremendously on feature selection, accuracy, and performance.
- Goal: To discover subset of features that yields best accuracy with lowest performance cost.

Solution Overview



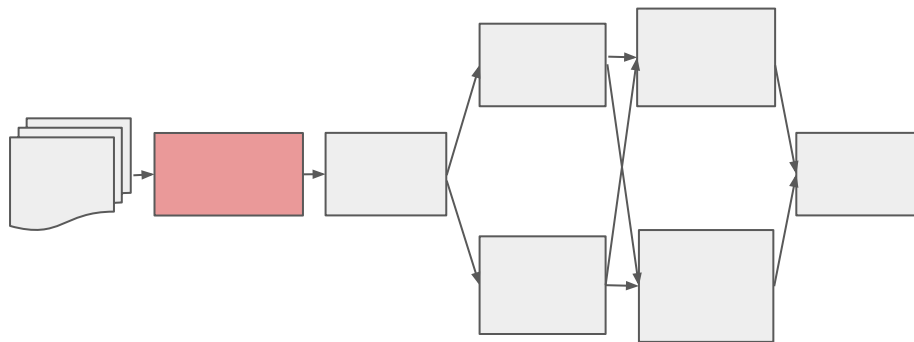
Training / Test Data

- Reuters 50-50 Dataset
- 2500 training documents, 50 authors
- 2500 test documents, 50 authors
- All documents are news articles



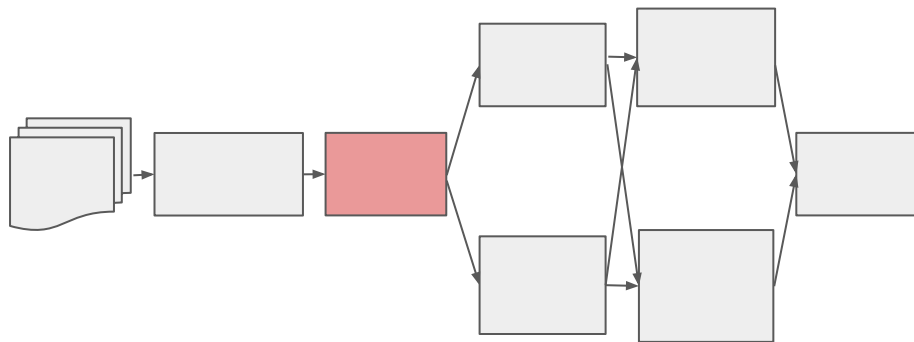
Data Preprocessing

- Separate labels from data and parse articles into proper format
- Group articles by author
- Tokenize, lemmatize, and stem each article
- Generate POS tags for data



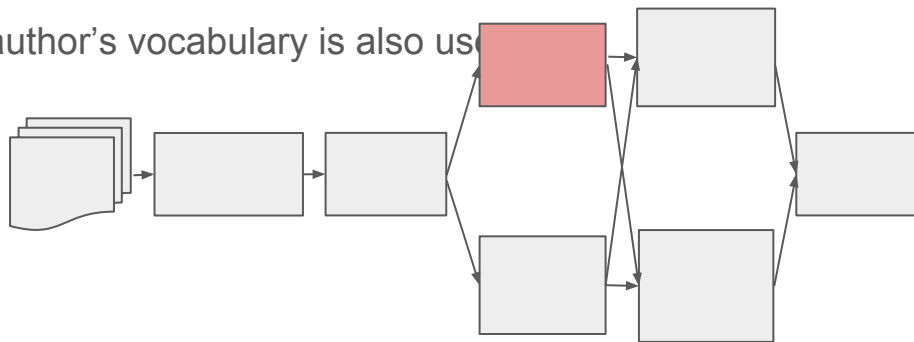
Feature Extraction

- Frequency of conjunctions
- Frequency of modals
- Frequency of determiners
- Frequency of quantifiers
- Frequency of pronouns
- Number of sentences per article
- Number of unique words per article
- Average length of sentence
- Average word length
- Number of total words per article
- Number of periods
- Number of commas
- Number of colons
- Number of semicolons
- Number of exclamation marks
- Number of question marks
- N-grams of text
 - 1,2,3
 - surface,tokens,lemmas
- Bag of words(TF-IDF)
- N-grams of parts of speech
 - 1,2,3
- POS dependencies



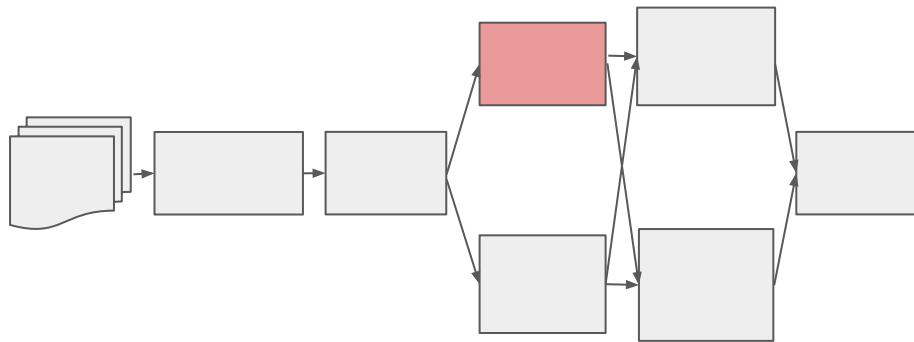
Manual Feature Selection

- Stylometry
 - The statistical analysis of style, stylometry, is based on the assumption that every author's style has certain features being accessible to conscious manipulation. Therefore they are considered to provide a reliable basis for the identification of an author.
- Bag of Words
 - The bag-of-words representation, where the document is represented with a vector of the word counts that appear in it. Depending on the classification method, the bag-of-words vector can be normalized to unity and scaled so that common words are less important than rare words
 - Measuring the “richness” or “diversity” of an author’s vocabulary is also used as a discriminating feature.
 - Use TF-IDF



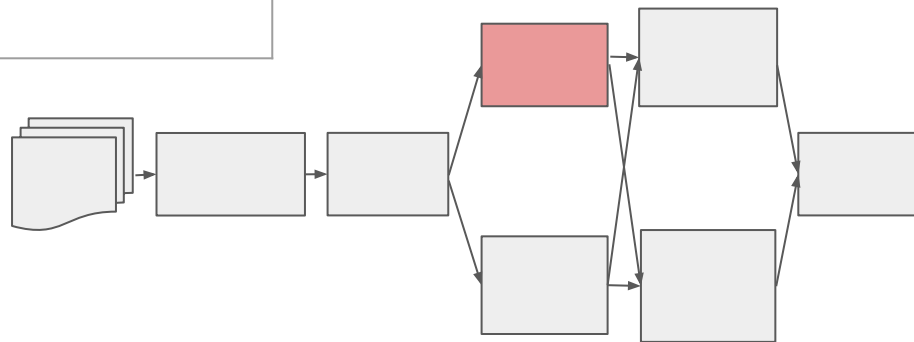
Manual Feature Selection

- Frequency of function words
 - The function words (modal, pronoun, conjunction) are used as a discriminating feature of author
- All features
 - Self explanatory



Stylometric Features

| | | |
|-----------------------------|---------------------------|-------------------------|
| Number of Sentences | Number of words | Average Sentence Length |
| Average Word Length | Number of Different Words | Number of Periods |
| Number of Commas | Number of Colons | Number of Semi colons |
| Number of Exclamation Marks | Number of Question Marks | |

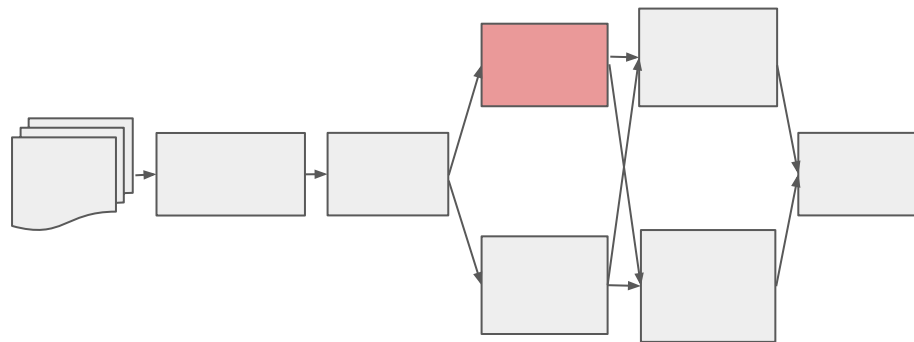


Features

- Frequency of Function Words

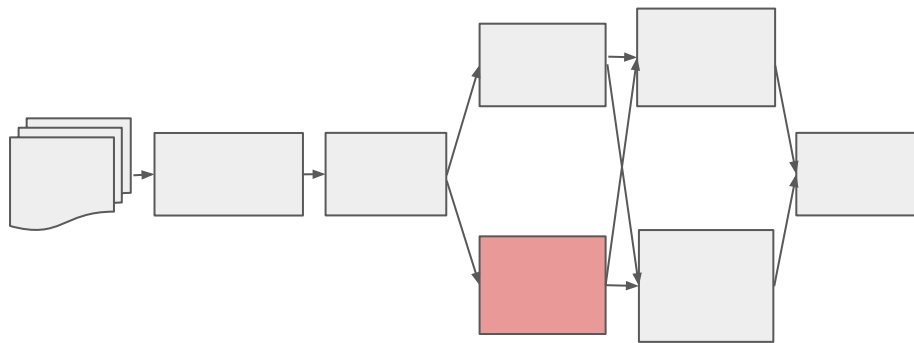
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| | | |
|-------------|---------|-------------|
| Conjunction | Modals | Determiners |
| Quantifier | Pronoun | |



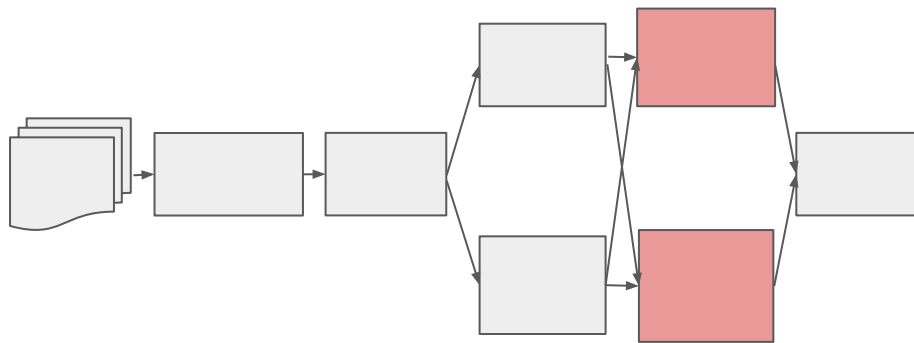
Automated Feature Selection

- Univariate Feature Selection -- SelectKBest, preprocessing
 - Recursive Feature Elimination -- prune lowest weighted features
 - L1-based feature selection -- SelectFromModel & LassoCV
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- All part of scikit



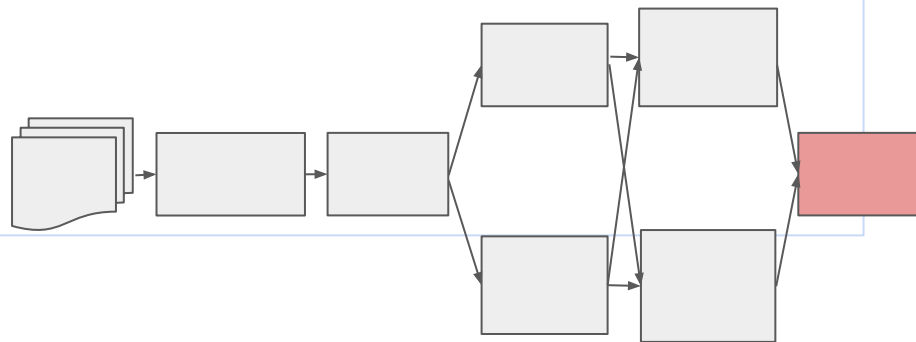
Build Classifiers

- Discriminative Classifier
 - SVM -- Linear, Polynomial, Gaussian Radial Basis Function
- Generative Classifier
 - Naive Bayes



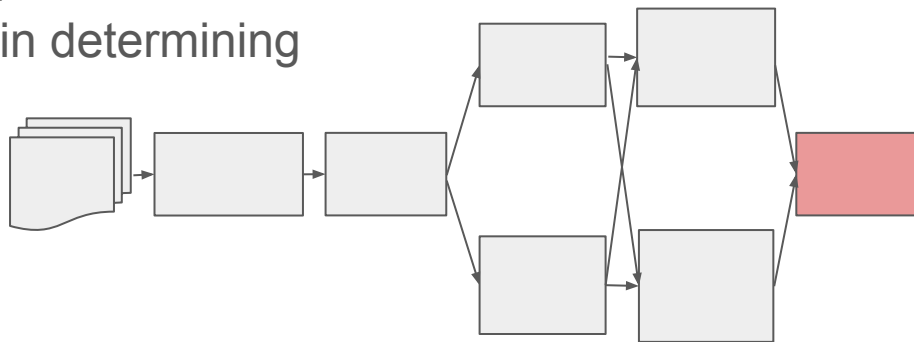
Test & Analyze

- Baseline
 - Trigram language models with smoothing
 - Provides baseline on actual data
 - Surface, token, lemma



Test & Analyze

- Given:
 - Feature subset
 - Classifier
- Output:
 - Accuracy
 - Performance
- By the end we will be able to determine the best feature subset and classifier pair that optimizes both accuracy and performance as well as see which individual features are most important in determining authorship classification



Reference

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