Report When I only choose 1000 most frequent, I observed homogeneisity of the occurrence (lots of '1'), so I expanded to 5000 and introducted Laplace smoothing to counteract the zero counts for certain words to make to multiplication invalid. Also, rule out some function words which we can't get too much information from to improve the performance.

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In [1]: # import packages
         import pandas as pd
         import numpy as np
         from collections import Counter
In [2]: | # read data
         target = pd.read csv('trg.csv').to numpy()
         test = pd.read csv('tst.csv').to numpy()
In [3]:
        # Collect all words in abstracts
         abstracts = target[:,2]
         vocabulary = []
         for abstract in abstracts:
             for word in abstract.split():
                 vocabulary.append(word);
        # Eliminate words that make little sense to classification (i.e. numbers, fun
In [4]:
         vocabulary = list(filter(lambda a: a.isalpha()!=0, vocabulary))
         function words = ["a", "about", "above", "after", "again", "against", "ain", "all", "
         for function word in function words:
             if word in vocabulary:
                 vocabulary = list(filter(lambda a: a!= function word, vocabulary))
In [5]: | # Find the most frequent words
         most frequent = []
         for counter in Counter(vocabulary).most common(5000):
             most_frequent.append(counter[0])
        # We add each of the most frequent words as an separate attribute, with 1 or
In [6]:
         training = np.concatenate((target,np.zeros((len(target),len(most frequent)),d
         for rowindex in range(0,len(training)-1):
             for columnindex in range(0,len(most_frequent)-1):
                 if training[rowindex,2].find(most frequent[columnindex]):
                     training[rowindex,3+columnindex] = 1
In [7]: # Naive Bayes Algorithm
        ## Priors
In [8]:
         priors= {'A':0,'B':0,'E':0,'V':0}
         for prior in Counter(training[:,1]).most common(4):
             priors[prior[0]] = prior[1]/len(training)
        ## Conditionals with Laplace smoothing
In [9]:
         donominatorcounts= {'A':0,'B':0,'E':0,'V':0}
         for prior in Counter(training[:,1]).most_common(4):
             donominatorcounts[prior[0]] = prior[1] + len(most frequent)
         # nominatorcounts is the sum of numbers in the columns of the correponding cl
         # Then, we can apply the probabilities to test set and choose the class of Hm
```