Read in the csv file using pandas.

df = pd.read\_csv('federalist.csv')

import pandas as pd

Try logistic regression.

In [1]:

```
df[12:18]
                        #just showing myself there are rows where authors are not exactly known
Out[1]:
                             author
                                                                         text
         12
                                    FEDERALIST No. 13 Advantage of the Union in Re...
                         HAMILTON
         13
                          MADISON
                                    FEDERALIST No. 14 Objections to the Proposed C...
         14
                          HAMILTON
                                      FEDERALIST No. 15 The Insufficiency of the Pre...
                          HAMILTON FEDERALIST No. 16 The Same Subject Continued (...
         15
         16
                          HAMILTON FEDERALIST No. 17 The Same Subject Continued (...
         17 HAMILTON AND MADISON FEDERALIST No. 18 The Same Subject Continued (...
        Convert the author column to categorical data. Display the first few rows.
In [2]:
          print(df.dtypes, '\n')
                   object
         author
                   object
         text
         dtype: object
In [3]:
         df['author'] = pd.Categorical(df.author)
         print(df.dtypes, '\n')
         print(df.head())
         author
                   category
         text
                     object
         dtype: object
              author
                                                                      text
         0 HAMILTON FEDERALIST. No. 1 General Introduction For the...
                 JAY FEDERALIST No. 2 Concerning Dangers from Forei...
                 JAY FEDERALIST No. 3 The Same Subject Continued (C...
                 JAY FEDERALIST No. 4 The Same Subject Continued (C...
                 JAY FEDERALIST No. 5 The Same Subject Continued (C...
        Display the counts by author.
         #display counts by author
         s=pd.Series(df['author'])
         s.value_counts()
         HAMILTON
                                  49
Out[4]:
         MADISON
                                  15
         HAMILTON OR MADISON
                                  11
         JAY
                                   5
         HAMILTON AND MADISON
         Name: author, dtype: int64
        Divide into train and test, with 80% in train. Use random state 1234. Display the shape of train and test
In [5]:
         from sklearn.model_selection import train_test_split
         X = df.text
         y = df.author
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, train_size=0.8, random_state=1234)
         print(X_train.shape)
         print(X_test.shape)
         print(y_train.shape)
         print(y_test.shape)
         (66,)
         (17,)
         (66,)
         (17,)
        Process the text by removing stop words and performing tf-idf vectorization,
        fit to the training data only, and applied to train and test.
        Output the training set shape and the test set shape.
In [6]:
         from nltk.corpus import stopwords
         from sklearn.feature_extraction.text import TfidfVectorizer
          stopwords = set(stopwords.words('english'))
         vectorizer = TfidfVectorizer(stop_words=stopwords)
         # apply tfidf vectorizer
         X_train = vectorizer.fit_transform(X_train) # fit and transform the train data
         X_test = vectorizer.transform(X_test)
                                                        # transform only the test data
         # take a peek at the data
         print('train size:', X_train.shape)
         print('\ntest size:', X_test.shape)
         train size: (66, 7876)
         test size: (17, 7876)
        Try a Bernoulli Naïve Bayes model. What is your accuracy on the test set?
         from sklearn.naive_bayes import BernoulliNB
         BNB = BernoulliNB()
         BNB.fit(X_train, y_train)
         from sklearn.metrics import accuracy_score
         pred = BNB.predict(X_test)
         print('accuracy score: ', accuracy_score(y_test, pred))
         accuracy score: 0.5882352941176471
        Redo the vectorization with max_features option set to use only the 1000 most frequent words.
        In addition to the words, add bigrams as a feature.
        Try Naïve Bayes again on the new train/test vectors and compare your results.
In [8]:
         # re-apply tfidf vectorizer with max_features and adding bigrams.
         X = df.text
         y = df.author
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, train_size=0.8, random_state=1234)
         vectorizer = TfidfVectorizer(stop_words=stopwords, max_features=1000, ngram_range=(1,2))
         X_train = vectorizer.fit_transform(X_train)
         X test = vectorizer.transform(X test)
         BNB = BernoulliNB()
         BNB.fit(X_train, y_train)
         pred = BNB.predict(X_test)
         print('accuracy score: ', accuracy_score(y_test, pred))
         accuracy score: 0.9411764705882353
```

Adjust at least one parameter in the LogisticRegression() model to see if you can improve results over having no parameters. What are your results?

```
#logistic regression without parameters
          from sklearn.linear_model import LogisticRegression
          X = df.text
          y = df.author
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, train_size=0.8, random_state=1234)
          vectorizer = TfidfVectorizer(stop_words=stopwords, max_features=1000, ngram_range=(1,2))
          X_train = vectorizer.fit_transform(X_train)
          X_test = vectorizer.transform(X_test)
          LR = LogisticRegression()
          LR.fit(X_train, y_train)
          pred = LR.predict(X_test)
          print('accuracy score: ', accuracy_score(y_test, pred))
         accuracy score: 0.5882352941176471
In [10]:
          #logistic regression with parameters
          X = df.text
          y = df.author
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, train_size=0.8, random_state=1234)
          vectorizer = TfidfVectorizer(stop_words=stopwords, max_features=1000, ngram_range=(1,2))
          X_train = vectorizer.fit_transform(X_train)
          X_test = vectorizer.transform(X_test)
          LR = LogisticRegression(solver='newton-cg', class_weight='balanced', random_state=5, C=.2)
          LR.fit(X_train, y_train)
          pred = LR.predict(X_test)
          print('accuracy score: ', accuracy_score(y_test, pred))
         accuracy score: 0.8235294117647058
        Try a neural network. Try different topologies until you get good results. What is your final accuracy?
In [20]:
          from sklearn.neural_network import MLPClassifier
          X = df.text
          y = df.author
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, train_size=0.8, random_state=1234)
          vectorizer = TfidfVectorizer(stop_words=stopwords, max_features=1000, ngram_range=(1,2))
          X_train = vectorizer.fit_transform(X_train)
          X_test = vectorizer.transform(X_test)
          Classifier = MLPClassifier(activation='tanh', solver='adam', alpha=.0001, batch_size='auto', max_iter=1000)
          Classifier.fit(X_train, y_train)
          pred = Classifier.predict(X_test)
          print('accuracy score: ', accuracy_score(y_test, pred))
         accuracy score: 0.8823529411764706
In [24]:
          Classifier = MLPClassifier(hidden_layer_sizes=(150,5), activation='tanh', solver='adam', alpha=.0001, batch_size='auto', max_iter=1000)
          Classifier.fit(X_train, y_train)
          pred = Classifier.predict(X_test)
          print('accuracy score: ', accuracy_score(y_test, pred))
         accuracy score: 0.8823529411764706
In [36]:
          Classifier = MLPClassifier(hidden_layer_sizes=(200,150, 100, 50), activation='tanh', solver='adam', alpha=.0001, batch_size='auto', max_iter=1500)
          Classifier.fit(X_train, y_train)
          pred = Classifier.predict(X_test)
          print('accuracy score: ', accuracy_score(y_test, pred))
         accuracy score: 0.8235294117647058
 In [ ]:
```