```
import pandas as pd
In [73]:
           import numpy as np
           # import ml classifiers
In [74]:
           from nltk.tokenize import sent tokenize # tokenizes sentences
           from nltk.stem import PorterStemmer
                                                       # parsing/stemmer
           from nltk.tag import pos tag
                                                       # parts-of-speech tagging
           from nltk.corpus import wordnet
                                                       # sentiment scores
           from nltk.stem import WordNetLemmatizer # stem and context
           from nltk.corpus import stopwords
                                                       # stopwords
           from nltk.util import ngrams
           df = pd.read csv('spotify.csv').iloc[:, 1:]
In [75]:
           df.head()
Out[75]:
             acousticness
                             artists danceability duration_ms energy explicit
                             ['Mamie
                0.991000
                                          0.598
                                                     168333
                                                              0.224
                                                                          0 0cS0A1fUEUd1EW3FcF8
                             Smith']
                          ['"Screamin
                0.643000
                                          0.852
                                                     150200
                                                               0.517
                                                                          0 0hbkKFlJm7705H87l9w
                                Jay
                          Hawkins"']
                             ['Mamie
          2
                0.993000
                                          0.647
                                                     163827
                                                               0.186
                                                                          0 11m7laMUgmOKql3oYzu
                             Smith']
                             ['Oscar
          3
                 0.000173
                                          0.730
                                                     422087
                                                              0.798
                                                                              19Lc5SfJJ5O1oaxY0fp
                          Velazquez']
                0.295000
                             ['Mixe']
                                          0.704
                                                     165224
                                                              0.707
                                                                          1
                                                                               2hJjbsLCytGsnAHfdsl
         5 rows × 30 columns
           df.shape
In [76]:
Out[76]: (174389, 30)
           df.describe()
In [77]:
```

| Out[77]: | | acousticness | danceability | duration_ms | energy | explicit | instrume |
|----------|-------|---------------|---------------|--------------|---------------|---------------|----------|
| | count | 174389.000000 | 174389.000000 | 1.743890e+05 | 174389.000000 | 174389.000000 | 174389 |
| | mean | 0.499228 | 0.536758 | 2.328100e+05 | 0.482721 | 0.068135 | (|
| | std | 0.379936 | 0.176025 | 1.483958e+05 | 0.272685 | 0.251978 | C |
| | min | 0.000000 | 0.000000 | 4.937000e+03 | 0.000000 | 0.000000 | С |
| | 25% | 0.087700 | 0.414000 | 1.661330e+05 | 0.249000 | 0.000000 | С |
| | 50% | 0.517000 | 0.548000 | 2.057870e+05 | 0.465000 | 0.000000 | С |
| | 75% | 0.895000 | 0.669000 | 2.657200e+05 | 0.711000 | 0.000000 | С |
| | max | 0.996000 | 0.988000 | 5.338302e+06 | 1.000000 | 1.000000 | 1 |
| | | | | | | | |

8 rows × 25 columns

| Out[92]: | | acousticness | danceability | duration_ms | energy | explicit | instrumentalness | key | liveness | I |
|----------|---|--------------|--------------|-------------|--------|----------|------------------|-----|----------|---|
| | 0 | 0.991000 | 0.598 | 168333 | 0.224 | 0 | 0.000522 | 5 | 0.3790 | |
| | 1 | 0.643000 | 0.852 | 150200 | 0.517 | 0 | 0.026400 | 5 | 0.0809 | |
| | 2 | 0.993000 | 0.647 | 163827 | 0.186 | 0 | 0.000018 | 0 | 0.5190 | |
| | 3 | 0.000173 | 0.730 | 422087 | 0.798 | 0 | 0.801000 | 2 | 0.1280 | |
| | 4 | 0.295000 | 0.704 | 165224 | 0.707 | 1 | 0.000246 | 10 | 0.4020 | |

5 rows × 27 columns

```
In [81]: # Get the dummy variables for season
df1 = pd.get_dummies(df[['season']])
df1.head()
```

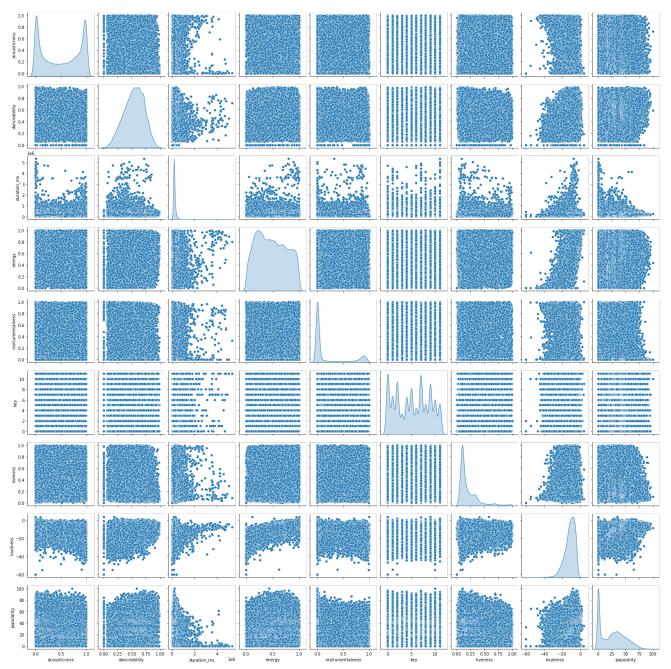
| Out[81]: | | season_Fall | season_Spring | season_Summer | season_Winter |
|----------|---|-------------|---------------|---------------|---------------|
| | 0 | 0 | 0 | 0 | 0 |
| | 1 | 0 | 0 | 0 | 1 |
| | 2 | 0 | 0 | 0 | 0 |
| | 3 | 0 | 0 | 0 | 1 |
| | 4 | 1 | 0 | 0 | 0 |

```
In [115... # Create the final dataset
    spotify = df.join(df1, how='outer')
    spotify.head()
```

| Out[115 | | acousticness | danceability | duration_ms | energy | explicit | instrumentalness | key | liveness |
|---------|---|--------------|--------------|-------------|--------|----------|------------------|-----|----------|
| | 0 | 0.991000 | 0.598 | 168333 | 0.224 | 0 | 0.000522 | 5 | 0.3790 |
| | 1 | 0.643000 | 0.852 | 150200 | 0.517 | 0 | 0.026400 | 5 | 0.0809 |
| | 2 | 0.993000 | 0.647 | 163827 | 0.186 | 0 | 0.000018 | 0 | 0.5190 |
| | 3 | 0.000173 | 0.730 | 422087 | 0.798 | 0 | 0.801000 | 2 | 0.1280 |
| | 4 | 0.295000 | 0.704 | 165224 | 0.707 | 1 | 0.000246 | 10 | 0.4020 |

5 rows × 31 columns

Out[116... <seaborn.axisgrid.PairGrid at 0x7fc7641e3700>



Building CART Model

Predict if the song is good

• A song is good if its popularity is greater than 25

```
from sklearn.model selection import train test split
In [83]:
          cols = ['acousticness', 'danceability', 'duration ms', 'energy', 'explicit',
                  'instrumentalness', 'key', 'liveness', 'loudness', 'mode', 'speechines
                 'tempo', 'valence', 'year x', 'Collaboration', 'name length', 'live',
                  'live', 'love', 'mix', 'no', 'op', 'remast', 'version', 'year_y',
                 'Season Fall', 'Season Spring', 'Season Summer', 'Season Winter']
          # re-split the dataset into training and testing data
          y = spotify['popularity']
          X = spotify[cols]
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, ran
          X_train.shape, X_test.shape
Out[83]: ((116840, 29), (57549, 29))
          y train.mean(), y train.max(), y train.min()
In [84]:
Out[84]: (25.67584731256419, 100, 0)
          import matplotlib.pyplot as plt
In [85]:
          plt.hist(y_train)
Out[85]: (array([3.9388e+04, 1.0138e+04, 1.5827e+04, 1.8036e+04, 1.4227e+04,
                 1.0394e+04, 6.0360e+03, 2.3960e+03, 3.7000e+02, 2.8000e+01]),
          array([ 0., 10., 20., 30., 40., 50., 60., 70., 80., 90., 100.]),
          <BarContainer object of 10 artists>)
          40000
         35000
         30000
         25000
         20000
         15000
         10000
          5000
                        20
                                40
                                        60
                                                80
                                                       100
```

convert the popularity to be 0 or 1

In [86]:

```
In [36]: from sklearn.tree import DecisionTreeClassifier
    from sklearn.tree import plot_tree
    from sklearn.model_selection import GridSearchCV
    from sklearn.tree import DecisionTreeClassifier

    grid_values = {'ccp_alpha': np.linspace(0, 0.1, 51)}

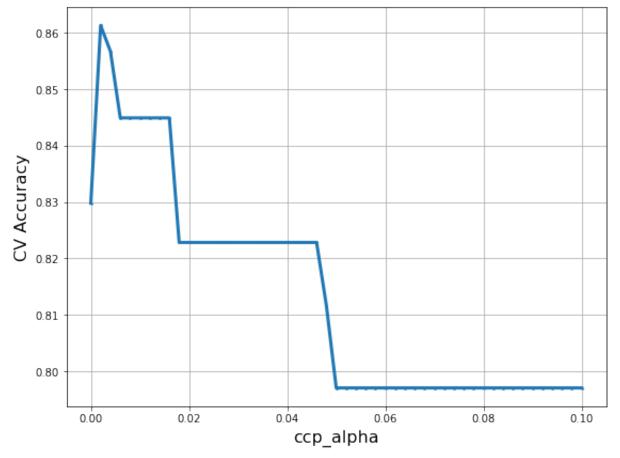
    dtc = DecisionTreeClassifier(random_state=88)
    dtc_cv = GridSearchCV(dtc, param_grid=grid_values, cv=5).fit(X_train, y_train)
```

```
In [37]: ccp_alpha = dtc_cv.cv_results_['param_ccp_alpha'].data
    ACC_scores = dtc_cv.cv_results_['mean_test_score']

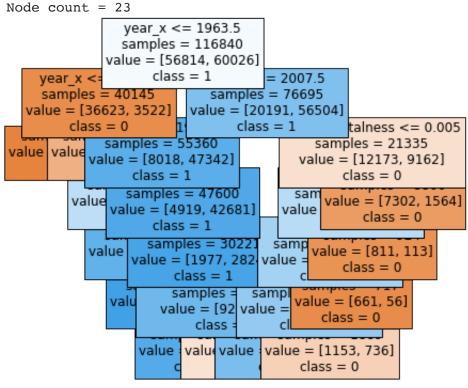
plt.figure(figsize=(8, 6))
    plt.xlabel('ccp_alpha', fontsize=16)
    plt.ylabel('CV Accuracy', fontsize=16)
    plt.scatter(ccp_alpha, ACC_scores, s=3)
    plt.plot(ccp_alpha, ACC_scores, linewidth=3)
    plt.grid(True, which='both')

plt.tight_layout()
    plt.show()

print('Best_ccp_alpha', dtc_cv.best_params_)
```



Best ccp alpha {'ccp alpha': 0.002}



```
In [39]: from sklearn.metrics import confusion_matrix
    from sklearn.metrics import accuracy_score
    from sklearn.metrics import precision_score

# Compute the performance of the training set
    y_pred = dtc_cv.predict(X_train)
    cm = confusion_matrix(y_train, y_pred)

print ("Confusion Matrix: \n", cm)
    print ("\nAccuracy:", accuracy_score(y_train, y_pred))
    print ("\nPrecision:", precision_score(y_train, y_pred))
```

```
Confusion Matrix:
    [[47044 9770]
    [ 6373 53653]]

Accuracy: 0.861836699760356

Precision: 0.8459549374832473

In [40]: # The performance of the test set
    y_pred = dtc_cv.predict(X_test)
    cm = confusion_matrix(y_test, y_pred)
    print ("Confusion Matrix: \n", cm)
    print ("\nAccuracy:", accuracy_score(y_test, y_pred))

Confusion Matrix:
    [[23145 4872]
    [ 3228 26304]]

Accuracy: 0.8592503779388
```

Random Forest

```
from sklearn.ensemble import RandomForestRegressor
In [109...
          import statsmodels.api as sm
          from sklearn.model selection import GridSearchCV
          from sklearn.tree import DecisionTreeRegressor
          from sklearn.model selection import KFold
          rf = RandomForestRegressor(max_features=5, min_samples_leaf=5,
                                     n_estimators = 500, random_state=88, verbose=2)
          rf.fit(X_train, y_train)
         [Parallel(n jobs=1)]: Using backend SequentialBackend with 1 concurrent worker
         building tree 1 of 500
         [Parallel(n_jobs=1)]: Done 1 out of 1 | elapsed: 0.4s remaining:
                                                                                     0.0
         building tree 2 of 500
         building tree 3 of 500
         building tree 4 of 500
         building tree 5 of 500
         building tree 6 of 500
         building tree 7 of 500
         building tree 8 of 500
         building tree 9 of 500
         building tree 10 of 500
         building tree 11 of 500
         building tree 12 of 500
         building tree 13 of 500
         building tree 14 of 500
         building tree 15 of 500
         building tree 16 of 500
         building tree 17 of 500
         building tree 18 of 500
```

building tree 19 of 500 building tree 20 of 500 building tree 21 of 500 building tree 22 of 500 building tree 23 of 500 building tree 24 of 500 building tree 25 of 500 building tree 26 of 500 building tree 27 of 500 building tree 28 of 500 building tree 29 of 500 building tree 30 of 500 building tree 31 of 500 building tree 32 of 500 building tree 33 of 500 building tree 34 of 500 building tree 35 of 500 building tree 36 of 500 building tree 37 of 500 building tree 38 of 500 building tree 39 of 500 building tree 40 of 500 building tree 41 of 500 building tree 42 of 500 building tree 43 of 500 building tree 44 of 500 building tree 45 of 500 building tree 46 of 500 building tree 47 of 500 building tree 48 of 500 building tree 49 of 500 building tree 50 of 500 building tree 51 of 500 building tree 52 of 500 building tree 53 of 500 building tree 54 of 500 building tree 55 of 500 building tree 56 of 500 building tree 57 of 500 building tree 58 of 500 building tree 59 of 500 building tree 60 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building tree 262 of 500 building tree 263 of 500 building tree 264 of 500 building tree 265 of 500 building tree 266 of 500 building tree 267 of 500 building tree 268 of 500 building tree 269 of 500 building tree 270 of 500 building tree 271 of 500 building tree 272 of 500 building tree 273 of 500 building tree 274 of 500 building tree 275 of 500 building tree 276 of 500 building tree 277 of 500 building tree 278 of 500 building tree 279 of 500 building tree 280 of 500 building tree 281 of 500 building tree 282 of 500 building tree 283 of 500 building tree 284 of 500 building tree 285 of 500 building tree 286 of 500 building tree 287 of 500 building tree 288 of 500 building tree 289 of 500 building tree 290 of 500 building tree 291 of 500 building tree 292 of 500 building tree 293 of 500 building tree 294 of 500 building tree 295 of 500 building tree 296 of 500 building tree 297 of 500 building tree 298 of 500 building tree 299 of 500 building tree 300 of 500 building tree 301 of 500 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         [Parallel(n jobs=1)]: Done 500 out of 500 | elapsed: 2.8min finished
Out[109... RandomForestRegressor(max_features=5, min_samples_leaf=5, n_estimators=500,
                               random state=88, verbose=2)
In [ ]:
In [112...
          # Evaluate the model performance on the testing set
          y prob = rf.predict(X test)
          y pred = pd.Series([1 if x >= 0.5 else 0 for x in y prob])
          cm = confusion matrix(y test, y pred)
          print ("Confusion Matrix : \n", cm)
          print ("\nAccuracy:", accuracy score(y test, y pred))
         [Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent worker
         [Parallel(n jobs=1)]: Done 1 out of 1 | elapsed:
                                                                 0.0s remaining:
                                                                                     0.0
         Confusion Matrix:
          [[23899 4118]
          [ 2527 27005]]
         Accuracy: 0.8845331804201637
         [Parallel(n_jobs=1)]: Done 500 out of 500 | elapsed: 9.1s finished
In [ ]:
In [ ]:
```