heartbeatclassifier

April 28, 2021

```
[37]: %%time
      # import libraries used in this project
      import os
      import fnmatch
      import numpy as np
      import pandas as pd
      import librosa
      import matplotlib.pyplot as plt
      import seaborn as sns
      from sklearn.preprocessing import OneHotEncoder
      from sklearn.preprocessing import LabelEncoder
      # configure matplot to not warn on large data set
      plt.rcParams.update({'figure.max_open_warning': 0})
      %matplotlib inline
     CPU times: user 530 μs, sys: 1 μs, total: 531 μs
     Wall time: 535 µs
 [2]: \%\time
      # method to get zero crossing, sepctral centroid, spectral rolloff, and
      \rightarrow chroma_stft features from .wav files
      def getDataFrame(inputFolders,dataFrameColumns,heartbeatclassifier):
          inputList=[]
          counter=0
          for inputFolder in inputFolders:
              for soundType in heartbeatclassifier:
                  filesOfCategory=fnmatch.filter(os.listdir("./data/"+inputFolder+"/
       →"),soundType+"*.wav")
                  if soundType == "extrastole":
```

```
moreFiles=fnmatch.filter(os.listdir("./data/"+inputFolder+"/
→"),"extrahls*.wav")
               filesOfCategory = filesOfCategory + moreFiles
           for file in filesOfCategory:
               x,sr=librosa.load("./data/"+inputFolder+"/
→"+file,duration=5,res_type='kaiser_fast')
               inputList.append([np.mean(x) for x in librosa.feature.
\rightarrowmfcc(x,sr=sr)])
               inputList[counter].append(sum(librosa.zero_crossings(x)))
               inputList[counter].append(np.mean(librosa.feature.
\hookrightarrowspectral_centroid(x)))
               inputList[counter].append(np.mean(librosa.feature.
⇔spectral_rolloff(x,sr=sr)))
               inputList[counter].append(np.mean(librosa.feature.
inputList[counter].append(soundType)
               counter+=1
   return pd.DataFrame(inputList,columns=dataFrameColumns)
```

```
CPU times: user 2 \mus, sys: 0 ns, total: 2 \mus Wall time: 4.05 \mus
```

```
[3]: %%time

# music_folders=["set_a"]
music_folders=["set_a","set_b"]

# label frequency and librosa feature column headers
inputColums1=["Freq"+str(i) for i in range(20)]
inputColums2=["zero","centroid","rolloff","chromagram","outputbeatclassifier"]
inputColums1.extend(inputColums2)

# label output classifiers
outputClassifier=["normal","artifact","murmur","extrastole"]

# process data
dataframe=getDataFrame(music_folders,inputColums1,outputClassifier)

# save data frame to csv
dataframe.to_csv(r'music_dataframe.csv', index = False)
```

/Users/vijit/opt/anaconda3/lib/python3.8/sitepackages/librosa/core/pitch.py:153: UserWarning: Trying to estimate tuning from empty frequency set. warnings.warn("Trying to estimate tuning from empty frequency set.")

```
# print data frame
     dataframe.head()
    CPU times: user 1.04 ms, sys: 29 µs, total: 1.06 ms
    Wall time: 134 µs
[4]:
             Freq0
                         Freq1
                                    Freq2
                                               Freq3
                                                          Freq4
                                                                     Freq5 \
     0 -489.621796
                     70.239189
                                61.142830
                                           48.898048
                                                      36.555328
                                                                 26.469486
     1 -406.428528
                    153.238861
                                -1.369525
                                           16.263828
                                                      10.937109
                                                                 16.939487
     2 -511.582214
                     82.091522
                                 6.478385
                                           35.782322
                                                       4.926917
                                                                 23.501289
     3 -514.132935
                     78.292191
                                65.463066
                                           49.305317
                                                      34.505836
                                                                 23.954039
     4 -371.671722 155.286530
                                35.827465
                                           19.501045 37.935871
                                                                 34.704395
                                                                   Freq16
           Freq6
                       Freq7
                                  Freq8
                                             Freq9
                                                         Freq15
       19.623278 15.663741
                                         11.869776
                                                       2.056817
     0
                             13.472218
                                                    •••
                                                                 2.049077
     1
         4.494656
                    6.633343
                               6.228123
                                          3.696192
                                                    ... -2.767086 -3.141026
     2
         2.172138 16.772097 -0.364136
                                         12.615508
                                                    ... 4.534374 -2.864163
     3 18.073490 15.343872 13.722631
                                         11.918728
                                                    ... 1.235011 1.578198
     4 17.897236
                                                    ... -3.627311 -3.266012
                    9.181622 10.555705
                                         10.437612
          Freq17
                    Freq18
                              Freq19
                                       zero
                                                centroid
                                                              rolloff
                                                                       chromagram
       2.306891
                  2.535395
                            2.566249
                                        605
                                              157,260560
                                                            208.852132
                                                                          0.743970
     1 -3.952870 -3.527147 -4.081270
                                       7624
                                             1046.537575
                                                          1888.940430
                                                                          0.663114
     2 2.395745 -2.710135 1.320220
                                      22103
                                             2243.817234
                                                         4613.987223
                                                                         0.754454
     3 2.168477
                  2.570371
                            2.564390
                                        660
                                                           213.587443
                                              167.595804
                                                                         0.703607
                                       1834
     4 0.071095 0.160198 -1.439477
                                              513.218302
                                                           749.435967
                                                                         0.634785
        outputbeatclassifier
     0
                      normal
     1
                      normal
     2
                      normal
     3
                      normal
     4
                      normal
     [5 rows x 25 columns]
[5]: %%time
     # statistically describe librosa features
     columnsToDescribe=["zero","centroid","rolloff","chromagram"]
     dataframe[columnsToDescribe].describe()
    CPU times: user 92.7 ms, sys: 4.07 ms, total: 96.8 ms
```

CPU times: user 16min 43s, sys: 15.4 s, total: 16min 59s

Wall time: 2min 33s

[4]: %%time

Wall time: 12 ms

```
[5]:
                             centroid
                                           rolloff
                                                     chromagram
                    zero
     count
              585.000000
                           585.000000
                                        585.000000
                                                     585.000000
    mean
             4062.259829
                           593.459736 1163.797054
                                                       0.663512
    std
             8735.425918
                           840.006987
                                       1452.608685
                                                       0.079439
    min
              161.000000
                           101.834424
                                          74.468994
                                                       0.115923
    25%
              958.000000
                           266.161344
                                        543.713379
                                                       0.643089
     50%
             1529.000000
                           323.542573
                                        676.552012
                                                       0.681797
    75%
             2502.000000
                           415.151187
                                        860.391899
                                                       0.707757
            75315.000000
    max
                          6769.973347 8923.169963
                                                       0.806292
```

[6]: %%time # statistically describe frequencies freqColums=["Freq"+str(i) for i in range(20)] dataframe[freqColums].describe()

CPU times: user 279 ms, sys: 5.65 ms, total: 284 ms

Wall time: 35.5 ms

[6]:		Freq0	Freq1	Freq2	Freq3	Freq4	\	
	count	585.000000	585.000000	585.000000	585.000000	585.000000		
	mean	-405.757946	170.079428	54.052007	-1.498439	1.345889		
	std	93.353767	48.520015	26.297832	23.266973	12.645158		
	min	-1081.043335	-29.527250	-72.262596	-53.103825	-28.883732		
	25%	-443.315155	155.420853	52.296009	-17.343857	-5.253887		
	50%	-403.216949	177.656403	62.453419	-4.244802	0.300408		
	75%	-365.044312	199.055923	68.766457	8.346755	5.557584		
	max	130.121475	271.676331	91.316147	62.930393	48.075516		
		Freq5	Freq6	Freq7	Freq8	Freq9	Freq10	\
	count	585.000000	585.000000	585.000000	585.000000	585.000000	585.000000	
	mean	27.155482	20.358246	2.652989	-3.246019	10.722109	13.194057	
	std	9.634136	12.516750	9.144572	8.266007	6.241936	10.146853	
	min	-31.063078	-44.440163	-43.664692	-49.969437	-38.816525	-31.041985	
	25%	22.754147	19.233067	-1.925813	-7.692395	8.333854	12.741400	
	50%	27.771646	23.946444	2.696302	-2.829227	11.513292	16.187937	
	75%	33.162655	27.599564	7.811433	1.094737	14.004535	18.846331	
	max	50.836891	48.271702	38.122517	20.969564	24.214865	27.890099	
		Freq11	Freq12	Freq13	Freq14	Freq15	Freq16	\
	count	585.000000	585.000000	585.000000	585.000000	585.000000	585.000000	
	mean	4.294947	-5.005561	0.597928	5.872527	4.242858	-2.246271	
	std	4.716163	4.521289	3.401845	6.309628	3.699300	4.221884	
	min	-35.024479	-22.014156	-18.243839	-20.437279	-14.023951	-13.795611	
	25%	2.292238	-7.601873	-1.298143	5.311940	2.965489	-4.529719	
	50%	5.078257	-5.150558	0.380347	7.603932	5.057027	-2.086371	

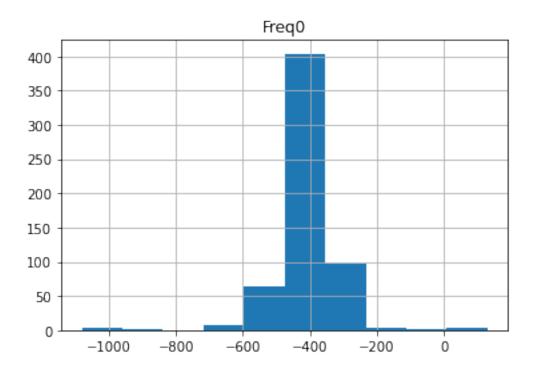
```
75%
         7.108112
                    -2.797366
                                  1.936331
                                               9.408846
                                                           6.316283
                                                                       -0.114404
        23.331343
                     23.732828
                                 29.896132
                                             43.768089
                                                          27.439344
                                                                      47.039963
max
           Freq17
                        Freq18
                                    Freq19
count
       585.000000
                   585.000000
                                585.000000
        -0.726721
                      3.581544
                                  4.159655
mean
                      4.734485
std
         3.813255
                                  3.515664
min
       -25.276119 -16.287870
                                -11.521476
25%
        -2.209787
                      2.712182
                                  3.098541
50%
        -0.976935
                      4.631459
                                  5.032918
75%
         0.263281
                      5.920696
                                  6.155967
max
        50.584553
                     39.139015
                                 32.866734
```

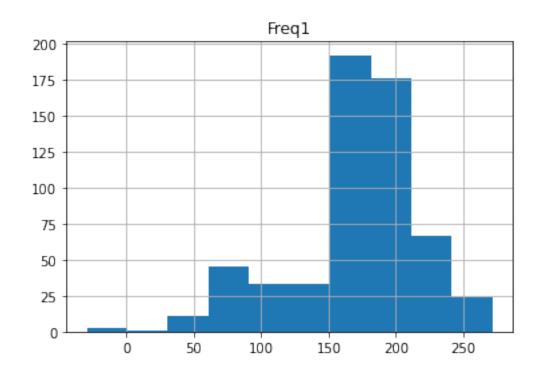
[7]: %%time

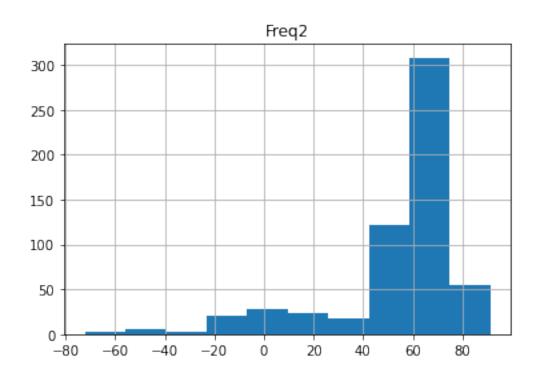
```
# graph data frame
columnsToRemove=["filename","outputbeatclassifier"]
for col in dataframe.columns:
   if(col not in columnsToRemove):
        dataframe.hist(column=col)
```

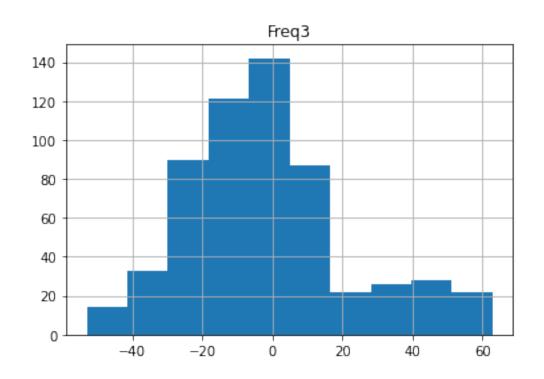
CPU times: user 995 ms, sys: 11.7 ms, total: 1.01 s

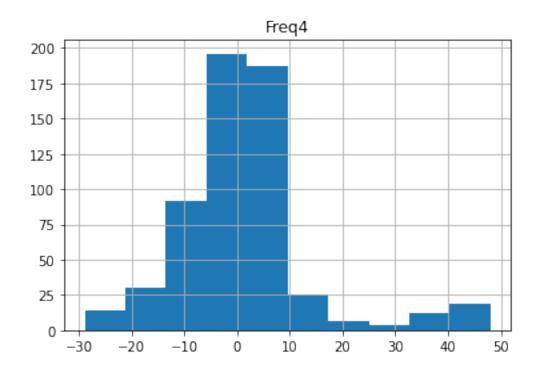
Wall time: 594 ms

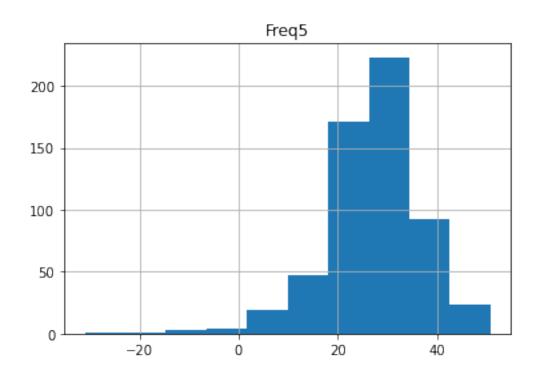


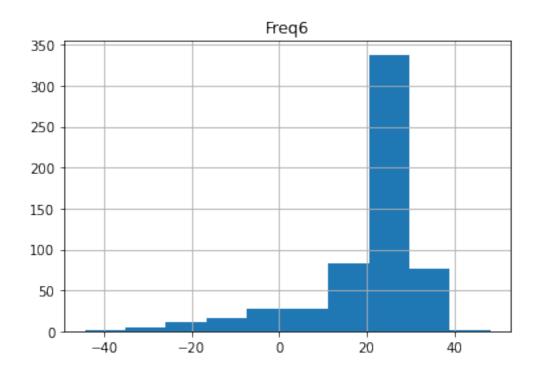


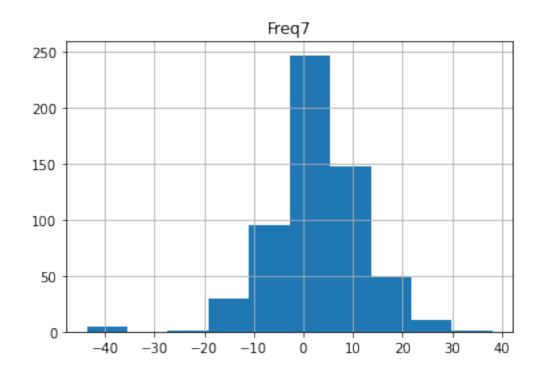


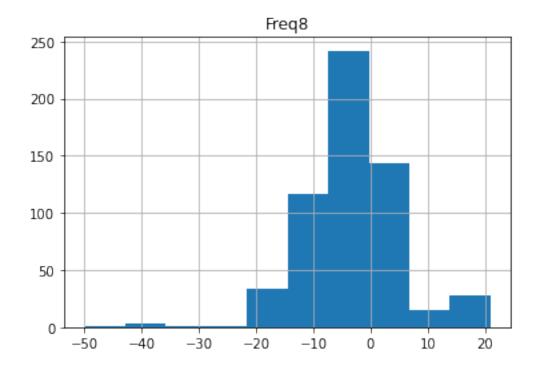


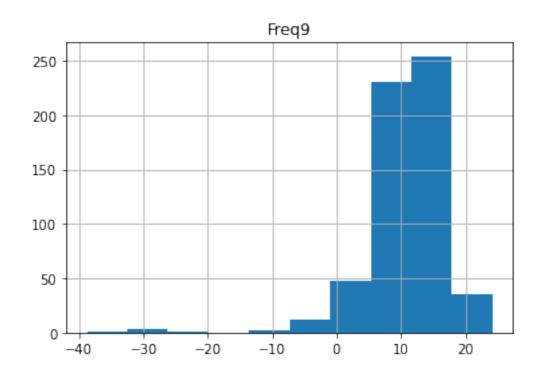


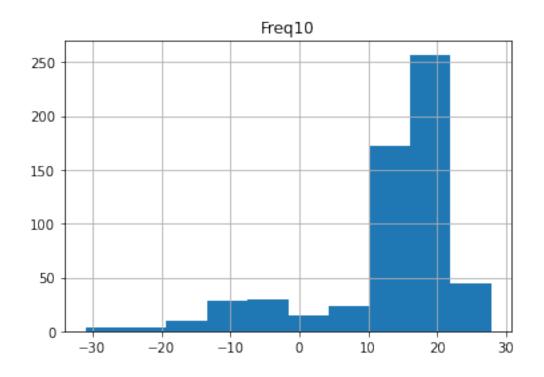


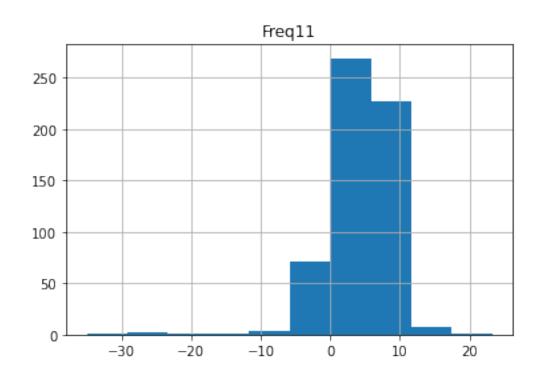


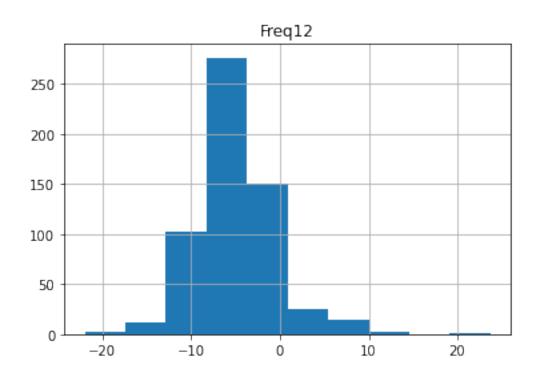


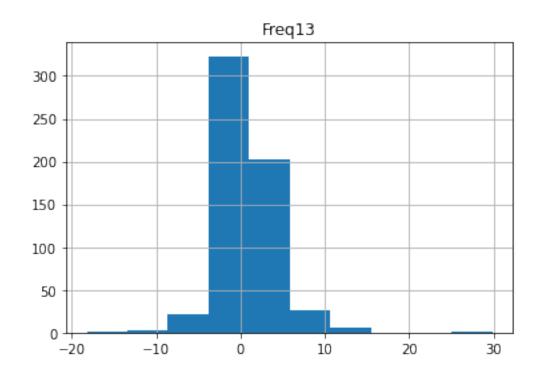


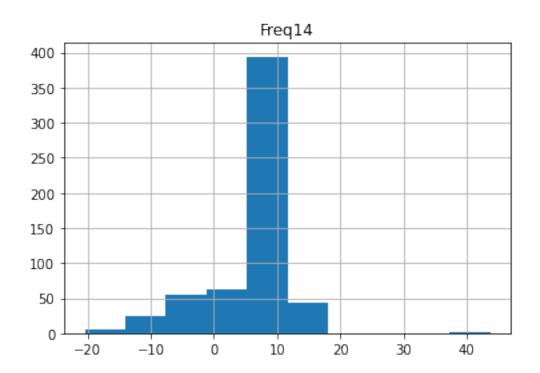


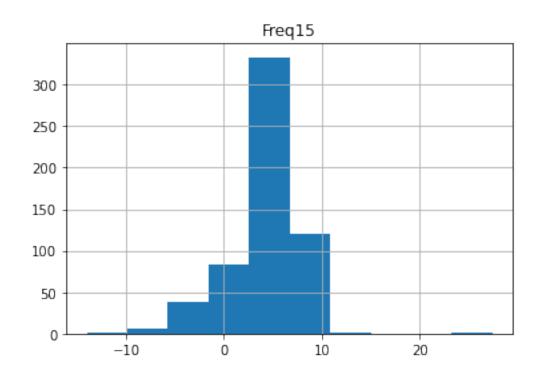


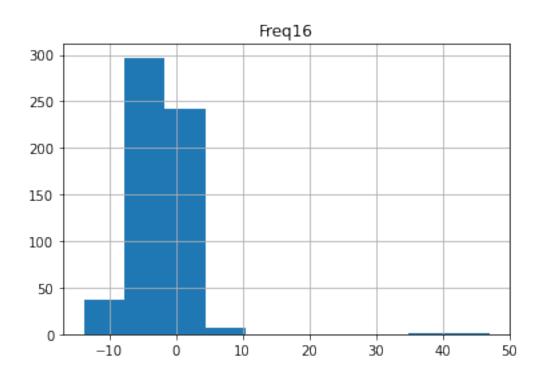


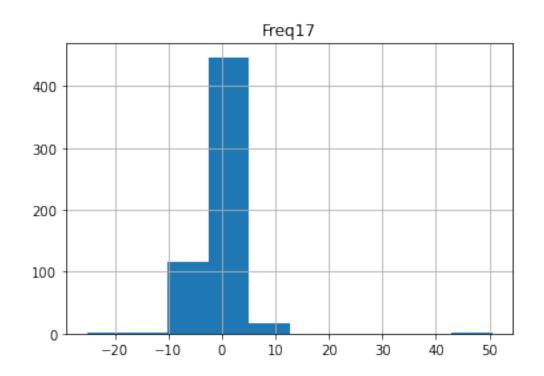


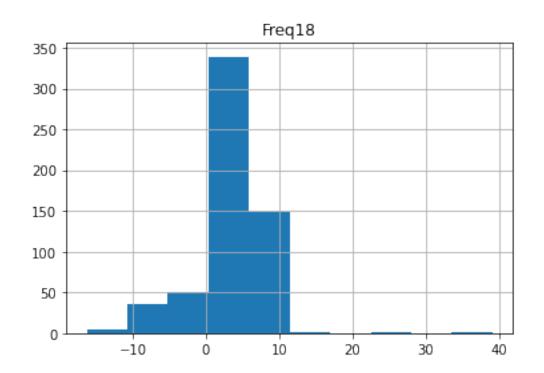


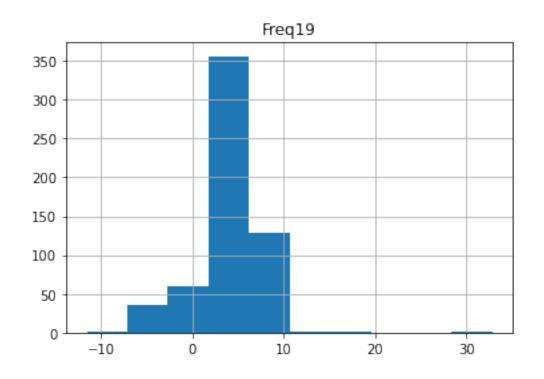


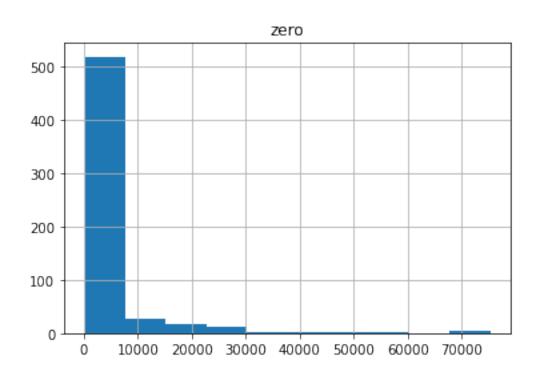


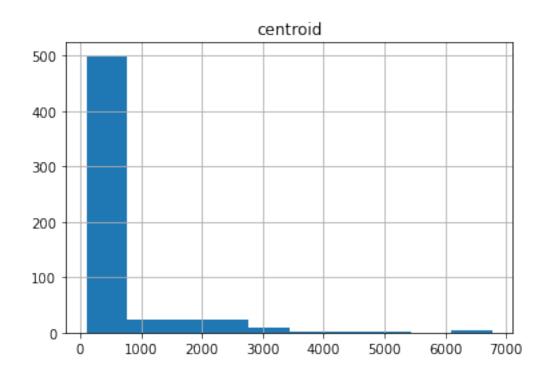


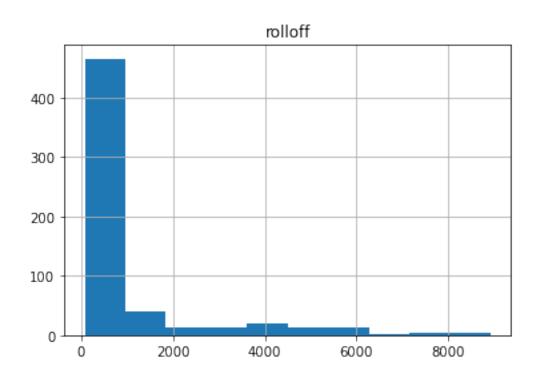


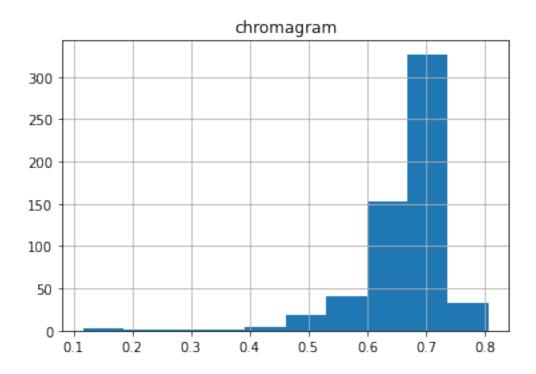












CPU times: user 77.2 ms, sys: 8.48 ms, total: 85.7 ms

Wall time: 83.9 ms

[8]: <AxesSubplot:>



```
[47]: %%time
      \# classification of each file in dataframe
      x=dataframe.iloc[:, 0]
      y=dataframe.iloc[:,-1]
      values = np.array(y)
      enc = OneHotEncoder(handle_unknown='ignore')
      # enc.fit(y)
      # print(enc.transform(y))
      # integer encode
      label_encoder = LabelEncoder()
      integer_encoded = label_encoder.fit_transform(values)
      # print(integer_encoded)
      # binary encode
      onehot_encoder = OneHotEncoder(sparse=False)
      integer_encoded = integer_encoded.reshape(len(integer_encoded), 1)
      onehot_encoded = onehot_encoder.fit_transform(integer_encoded)
      print(onehot_encoded)
```

```
correlation= np.corrcoef(x,onehot_encoded, rowvar=False)
print(correlation)
```

```
[[0. 0. 0. 1.]
 [0. 0. 0. 1.]
 [0. 0. 0. 1.]
 [0. 1. 0. 0.]
 [0. 1. 0. 0.]
 [0. 1. 0. 0.]]
[[ 1.
               0.04853167 -0.08556399 0.04915338 -0.01171136]
 [ 0.04853167 1.
                         -0.09578263 -0.14409343 -0.33180076]
 [-0.08556399 -0.09578263 1.
                                     -0.18804745 -0.4330127 ]
 [ 0.04915338 -0.14409343 -0.18804745 1.
                                                  -0.65141546]
 [-0.01171136 -0.33180076 -0.4330127 -0.65141546 1.
                                                             ]]
CPU times: user 9.73 ms, sys: 1.75 ms, total: 11.5 ms
Wall time: 3.56 ms
```