```
In [1]: import pandas as pd
         import numpy as np
         import os
         import seaborn as sns
         import matplotlib.pyplot as plt
         import librosa
         import librosa.display
         from IPython.display import Audio
         import warnings
         warnings.filterwarnings('ignore')
 In [3]: paths = []
         labels = []
         for dirname, _, filenames in os.walk('Asthma Detection Dataset Version 2'):
              for filename in filenames:
                  paths.append(os.path.join(dirname, filename))
                  label = filename.split(' ')[0]
                  label = label.split('.')[0]
                  labels.append(label.lower())
              if len(paths) == 1211:
                  break
         print('Dataset is Loaded')
        Dataset is Loaded
 In [4]: len(paths)
 Out[4]: 1211
 In [5]: paths[:5]
 Out[5]: ['Asthma Detection Dataset Version 2\\asthma\\Asthma RS 20 (1).wav'
           'Asthma Detection Dataset Version 2\\asthma\\Asthma_RS_20 (10).wav',
           'Asthma Detection Dataset Version 2\\asthma\\Asthma RS 20 (100).wav',
           'Asthma Detection Dataset Version 2\\Asthma_RS_20 (101).wav',
           'Asthma Detection Dataset Version 2\\asthma\\Asthma RS 20 (102).wav']
 In [6]: labels[:5]
 Out[6]: ['asthma', 'asthma', 'asthma', 'asthma']
 In [7]: ## Create a dataframe
         df = pd.DataFrame()
         df['sound'] = paths
         df['label'] = labels
 In [8]: df.head()
 Out[8]:
                                               sound
                                                        label
         0 Asthma Detection Dataset Version 2\asthma\Asth... asthma
          1 Asthma Detection Dataset Version 2\asthma\Asth... asthma
         2 Asthma Detection Dataset Version 2\asthma\Asth... asthma
         3 Asthma Detection Dataset Version 2\asthma\Asth... asthma
          4 Asthma Detection Dataset Version 2\asthma\Asth... asthma
 In [9]: df.tail()
 Out[9]:
                                                   sound
                                                               label
         1206 Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
         1207 Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
          1208 Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
          1209 Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
          1210 Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
In [10]: df.sample(20)
```

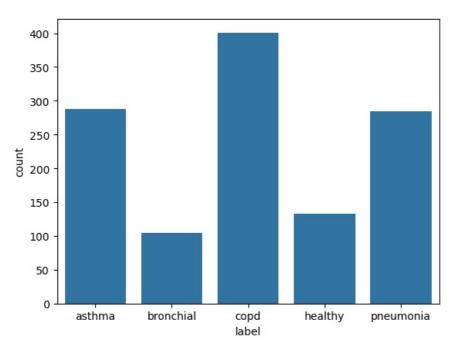
	sound	label
932	Asthma Detection Dataset Version 2\pneumonia\P	pneumonia
482	Asthma Detection Dataset Version 2\copd\COPD_M	copd
890	Asthma Detection Dataset Version 2\healthy\Hea	healthy
1108	Asthma Detection Dataset Version 2\pneumonia\P	pneumonia
825	Asthma Detection Dataset Version 2\healthy\Hea	healthy
588	Asthma Detection Dataset Version 2\copd\COPD_M	copd
502	$Asthma\ Detection\ Dataset\ Version\ 2\\ \ COPD_M$	copd
372	Asthma Detection Dataset Version 2\Bronchial\B	bronchial
799	Asthma Detection Dataset Version 2\healthy\Hea	healthy
652	Asthma Detection Dataset Version 2\copd\COPD_M	copd
828	Asthma Detection Dataset Version 2\healthy\Hea	healthy
317	Asthma Detection Dataset Version 2\Bronchial\B	bronchial
820	Asthma Detection Dataset Version 2\healthy\Hea	healthy
431	Asthma Detection Dataset Version 2\copd\COPD_M	copd
246	Asthma Detection Dataset Version 2\asthma\Asth	asthma
374	Asthma Detection Dataset Version 2\Bronchial\B	bronchial
1099	Asthma Detection Dataset Version 2\pneumonia\P	pneumonia
217	Asthma Detection Dataset Version 2\asthma\Asth	asthma
34	Asthma Detection Dataset Version 2\asthma\Asth	asthma
165	Asthma Detection Dataset Version 2\asthma\Asth	asthma

In [11]: df.sample(30)

Out[10]:

```
label
                                                  sound
 326
          Asthma Detection Dataset Version 2\Bronchial\B...
                                                            bronchial
1155
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
  41
         Asthma Detection Dataset Version 2\asthma\Asth...
                                                              asthma
 504
      Asthma Detection Dataset Version 2\copd\COPD_M...
                                                                copd
  71
         Asthma Detection Dataset Version 2\asthma\Asth...
                                                              asthma
  29
         Asthma Detection Dataset Version 2\asthma\Asth...
                                                              asthma
 110
         Asthma Detection Dataset Version 2\asthma\Asth...
                                                              asthma
1176
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
 949
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
 354
          Asthma Detection Dataset Version 2\Bronchial\B...
                                                            bronchial
 231
         Asthma Detection Dataset Version 2\asthma\Asth...
                                                              asthma
 630
       Asthma Detection Dataset Version 2\copd\COPD_M...
                                                                copd
1143
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
 137
         Asthma Detection Dataset Version 2\asthma\Asth...
                                                              asthma
 305
          Asthma Detection Dataset Version 2\Bronchial\B...
                                                            bronchial
 525
      Asthma Detection Dataset Version 2\copd\COPD_M...
                                                                copd
 193
         Asthma Detection Dataset Version 2\asthma\Asth
                                                              asthma
 985
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
 956
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
 764
       Asthma Detection Dataset Version 2\copd\COPD_M...
 730
      Asthma Detection Dataset Version 2\copd\COPD_M...
                                                                copd
1033
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
1140
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
 734
      Asthma Detection Dataset Version 2\copd\COPD_M...
 507
      Asthma Detection Dataset Version 2\copd\COPD_M...
                                                                copd
1091
        Asthma Detection Dataset Version 2\pneumonia\P... pneumonia
 859
         Asthma Detection Dataset Version 2\healthy\Hea...
                                                              healthy
 617
      Asthma Detection Dataset Version 2\copd\COPD_M...
                                                                copd
 535
      Asthma Detection Dataset Version 2\copd\COPD_M...
                                                                copd
  51
         Asthma Detection Dataset Version 2\asthma\Asth...
                                                              asthma
```

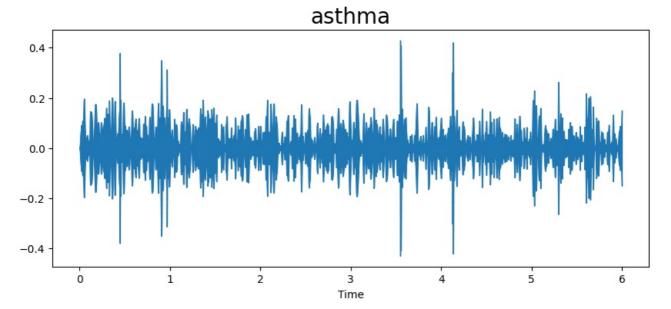
Out[11]:



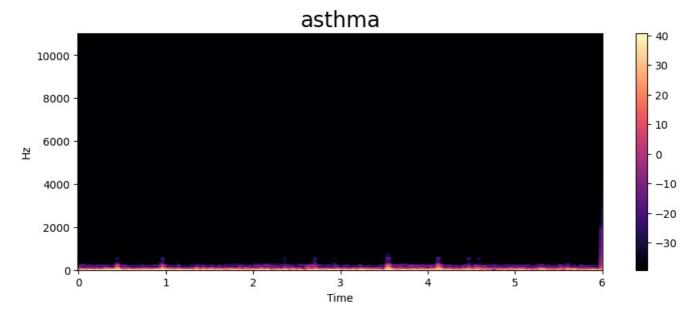
spectogram(data, sampling_rate, audio)

Audio(path)

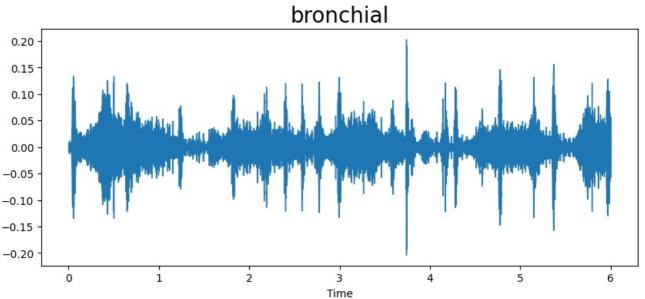
```
In [14]: def waveplot(data, sr, audio):
             plt.figure(figsize=(10,4))
             plt.title(audio, size=20)
             librosa.display.waveshow(data, sr=sr)
             plt.show()
         def spectogram(data, sr, audio):
             x = librosa.stft(data)
             xdb = librosa.amplitude to db(abs(x))
             plt.figure(figsize=(11,4))
             plt.title(audio, size=20)
             librosa.display.specshow(xdb, sr=sr, x_axis='time', y_axis='hz')
             plt.colorbar()
In [15]: audio = 'asthma'
         path = np.array(df['sound'][df['label']==audio])[0]
         data, sampling_rate = librosa.load(path)
         waveplot(data, sampling_rate, audio)
```



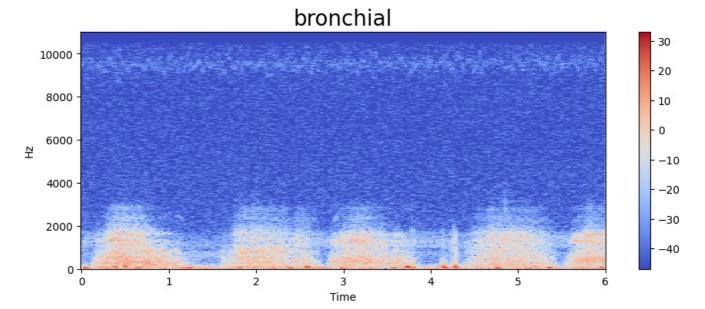
Out[15]: Your browser does not support the audio element.



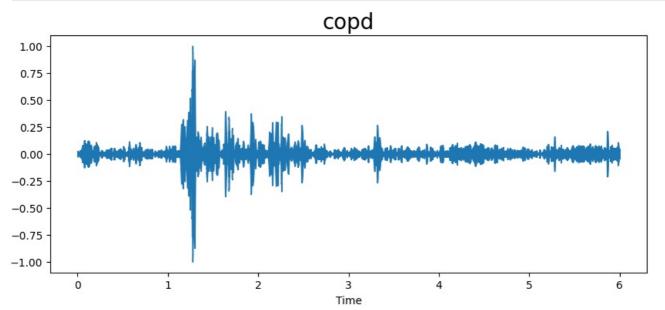




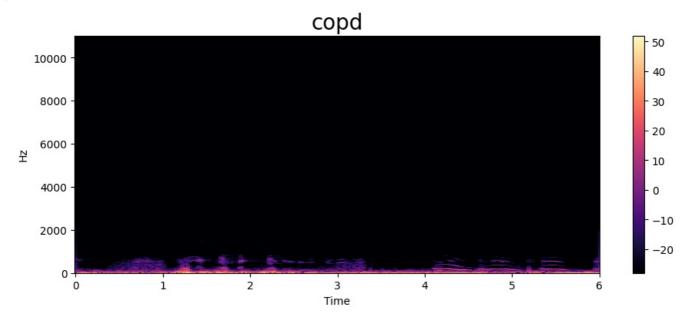
Out[16]: Your browser does not support the audio element.



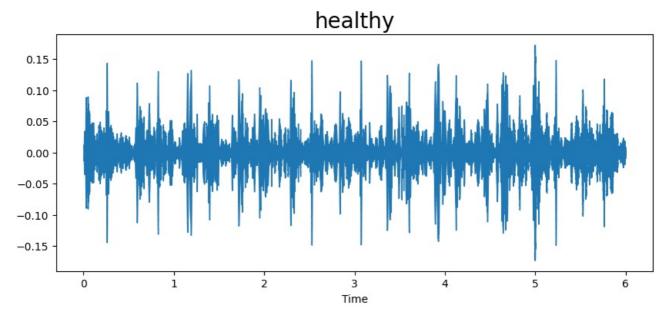
```
In [17]:
    audio = 'copd'
    path = np.array(df['sound'][df['label']==audio])[0]
    data, sampling_rate = librosa.load(path)
    waveplot(data, sampling_rate, audio)
    spectogram(data, sampling_rate, audio)
    Audio(path)
```



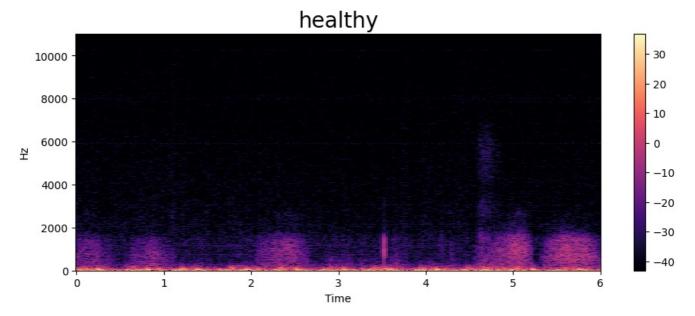
Out [17] : Your browser does not support the audio element.



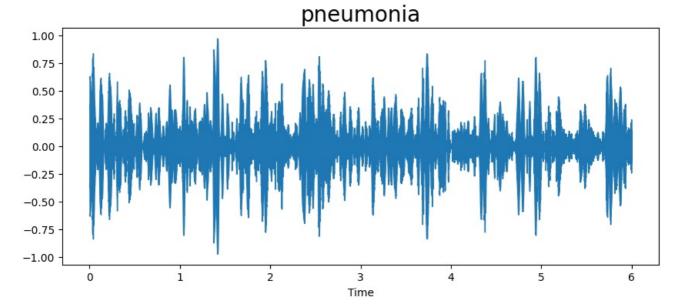
```
In [18]:
    audio = 'healthy'
    path = np.array(df['sound'][df['label']==audio])[0]
    data, sampling_rate = librosa.load(path)
    waveplot(data, sampling_rate, audio)
    spectogram(data, sampling_rate, audio)
    Audio(path)
```



Out[18]: Your browser does not support the audio element.

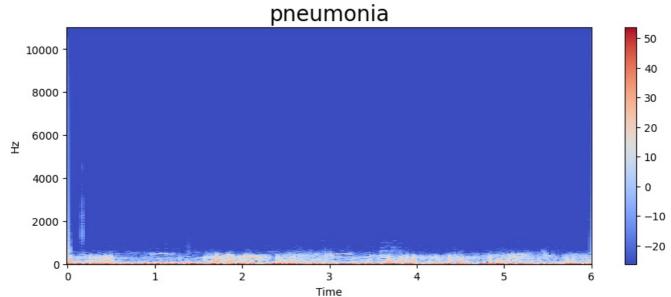


```
In [19]: audio = 'pneumonia'
  path = np.array(df['sound'][df['label']==audio])[0]
  data, sampling_rate = librosa.load(path)
  waveplot(data, sampling_rate, audio)
  spectogram(data, sampling_rate, audio)
  Audio(path)
```



Out[19]: Your browser does not support the audio element.

In [23]: X mfcc



```
In [20]: def extract_mfcc(filename):
              y, sr = librosa.load(filename, duration=3, offset=0.5)
              mfcc = np.mean(librosa.feature.mfcc(y=y, sr=sr, n_mfcc=40).T, axis=0)
              return mfcc
In [21]: extract_mfcc(df['sound'][0])
                                  50.119514 ,
                                                               43.384907 ,
Out[21]: array([-548.83496
                                                45.78449
                   40.99531
                                  37.92376
                                                34.31658
                                                              30.903397
                   27.728077 ,
                                  24.546072 ,
                                                21.37009
                                                               18.457767 ,
                   15.855426 ,
                                  13.384048 ,
                                                11.001728 ,
                                                               8.931421
                    7.295938 ,
                                  5.979423 ,
                                                 4.9392314,
                                                               4.3285704,
                    4.1767793,
                                   4.285747 ,
                                                 4.5153503,
                                                                4.891299 ,
                                   5.767992 ,
                    5.3693304,
                                                 5.9868765,
                                                               6.0878544,
                    6.109491 ,
                                  5.9857993,
                                                 5.6918344,
                                                               5.316192 ,
                    4.9219117,
                                   4.4640946,
                                                 3.9046216,
                                                               3.3001502,
                    2.7046251,
                                   2.0866601,
                                                               0.7722891],
                                                 1.4203374,
                dtype=float32)
In [22]: X mfcc = df['sound'].apply(lambda x: extract mfcc(x))
```

```
Out[23]: 0
                 [-548.83496, 50.119514, 45.78449, 43.384907, 4...
                  [-510.55096, 64.57206, 58.473934, 54.521935, 5...
                  [-491.56403, 122.547005, 97.057144, 68.30173, ...
          2
                  [-410.38647, 96.17459, 71.43918, 44.839233, 25...
          4
                  [-584.2018, 149.43059, 109.23822, 64.08953, 26...
                  [-472.8342, 57.394924, 52.172047, 48.309334, 4...
          1206
          1207
                  [-514.3326, 67.60713, 61.357025, 55.430042, 48...
          1208
                  [-408.67758, 100.07234, 81.82103, 62.381306, 4...
          1209
                  [-451.33817, 125.4286, 52.841415, 16.206491, 1...
                  [-385.91092, 113.33983, 90.27464, 67.36163, 44...
          1210
          Name: sound, Length: 1211, dtype: object
In [24]: X = [x for x in X_mfcc]
          X = np.array(X)
         X.shape
Out[24]: (1211, 40)
In [25]: y=np.array(df['label'].tolist())
In [26]: y.shape
Out[26]: (1211,)
In [27]: X
                                   50.119514
Out[27]: array([[-548.83496 ,
                                                 45.78449 , ...,
                                                                      2.0866601,
                     1.4203374,
                                   0.7722891],
                 [-510.55096 ,
                                   64.57206
                                                 58.473934 , ...,
                                                                      2.840531 ,
                     2.3669758,
                                   1.8308487],
                 [-491.56403 , 122.547005 ,
                                                                      4.366303 ,
                                                 97.057144 , ...,
                     3.3435152,
                                  2.3243194],
                 [-408.67758
                                 100.07234
                                                 81.82103 , ...,
                                                                      2.2515595,
                                  2.8444352],
                     2.4373362,
                 [-451.33817 ,
                                 125.4286
                                                 52.841415 , ...,
                                                                      1.9024341,
                     2.301741 ,
                                  1.5902761],
                 [-385.91092 ,
                                 113.33983 ,
                                                 90.27464 , ...,
                                                                      1.5085611,
                     1.7577705.
                                  2.1316829]], dtype=float32)
In [28]: y
Out[28]: array(['asthma', 'asthma', 'asthma', ..., 'pneumonia', 'pneumonia',
                 'pneumonia'], dtype='<U9')
In [29]: from imblearn.over_sampling import SMOTE
          smote = SMOTE(random state=42)
          X_resampled, y_resampled = smote.fit_resample(X, y)
In [30]: X resampled.shape
Out[30]: (2005, 40)
In [31]: y resampled.shape
Out[31]: (2005,)
In [32]: X resampled
Out[32]: array([[-5.48834961e+02, 5.01195145e+01, 4.57844887e+01, ...,
                   2.08666015e+00, 1.42033744e+00, 7.72289097e-01],
                 [-5.10550964e+02, 6.45720596e+01, 5.84739342e+01, \ldots,
                 2.84053111e+00, 2.36697578e+00, 1.83084869e+00],
[-4.91564026e+02, 1.22547005e+02, 9.70571442e+01, ...,
                   4.36630297e+00, 3.34351516e+00, 2.32431936e+00],
                 [-4.79628448e+02, 1.23179596e+02, 5.41606674e+01, ...,
                  -4.48403448e-01, 2.29700789e-01,
                                                      6.30438566e-011.
                 [-4.96546539e+02, 1.02657547e+02, 5.94003410e+01, ...,
                 9.22891736e-01, 1.54411340e+00, 1.38727260e+00], [-4.69754822e+02, 9.63708801e+01, 6.30505295e+01,
                                                      6.30505295e+01, ...,
                   4.35085803e-01, 7.85743356e-01, 8.84941578e-01]], dtype=float32)
In [33]: y_resampled
Out[33]: array(['asthma', 'asthma', 'asthma', ..., 'pneumonia', 'pneumonia',
                 'pneumonia'], dtype='<U9')
In [34]: np.unique(y, return counts=True)
```

```
Out[34]: (array(['asthma', 'bronchial', 'copd', 'healthy', 'pneumonia'], dtype='<U9'),</pre>
           array([288, 104, 401, 133, 285], dtype=int64))
In [35]: np.unique(y resampled, return counts=True)
Out[35]: (array(['asthma', 'bronchial', 'copd', 'healthy', 'pneumonia'], dtype='<U9'),</pre>
           array([401, 401, 401, 401, 401], dtype=int64))
In [36]: fig, ax=plt.subplots(nrows=1, ncols=2, figsize=(30,6))
          sns.scatterplot(x=X_resampled[:,0], y=X_resampled[:,1], hue=y_resampled, ax=ax[0])
          idx, c=np.unique(y_resampled, return_counts=True)
          sns.barplot(x=idx, y=c, ax=ax[1])
          plt.show()
In [37]: from tensorflow.keras.utils import to_categorical
          from sklearn.preprocessing import LabelEncoder
          labelencoder=LabelEncoder()
          y_resampled=to_categorical(labelencoder.fit_transform(y_resampled))
In [38]: y_resampled.shape
Out[38]: (2005, 5)
In [39]: X_resampled
Out[39]: array([[-5.48834961e+02, 5.01195145e+01, 4.57844887e+01, ...,
                    {\tt 2.08666015e+00, \quad 1.42033744e+00, \quad 7.72289097e-01],}
                  [-5.10550964e+02, 6.45720596e+01, 5.84739342e+01, 2.84053111e+00, 2.36697578e+00, 1.83084869e+00],
                                                         5.84739342e+01, ...,
                  [-4.91564026e+02, 1.22547005e+02, 9.70571442e+01, ...,
                    4.36630297e+00, 3.34351516e+00, 2.32431936e+00],
                  [-4.79628448e+02, 1.23179596e+02, 5.41606674e+01, ...,
                   -4.48403448e-01, 2.29700789e-01, 6.30438566e-01],
                  [-4.96546539e+02, 1.02657547e+02, 9.22891736e-01, 1.54411340e+00,
                                                        5.94003410e+01, ...,
                                                        1.38727260e+00],
                  [-4.69754822e+02, 9.63708801e+01,
                                                        6.30505295e+01, ...
                    4.35085803e-01, 7.85743356e-01, 8.84941578e-01]], dtype=float32)
In [40]: y_resampled
Out[40]: array([[1., 0., 0., 0., 0.],
                  [1., 0., 0., 0., 0.],
[1., 0., 0., 0., 0.],
                  [0., 0., 0., 0., 1.],
                  [0., 0., 0., 0., 1.],
[0., 0., 0., 0., 1.]])
In [41]: X_resampled.shape
Out[41]: (2005, 40)
In [42]: y_resampled.shape
Out[42]: (2005, 5)
In [43]: from sklearn.model selection import train test split
          X train,X test,y train,y test=train test split(X resampled,y resampled,test size=0.2,random state=0)
In [44]: X_train
```

```
1.82643723e+00, 2.44722754e-01, -3.70863616e-01], [-4.77032867e+02, 1.20868042e+02, 5.64808655e+01, ..., 6.10798039e-02, 4.33665454e-01, 2.71079212e-01],
                    [-4.84113190e+02, 1.14587006e+02, 4.15470352e+01, ..., 1.10013235e+00, 3.49691749e-01, 1.18099129e+00], [-3.99894287e+02, 1.16806595e+02, 5.40612411e+01, ..., 8.57568979e-01, 2.63239563e-01, -6.58950865e-01],
                    [-4.40016235e+02, 1.02298325e+02, 6.68499146e+01, ..., 2.66842890e+00, 1.67758262e+00, 6.89723551e-01]], dtype=float32)
In [45]: print(X_train.shape)
           print(X_test.shape)
           print(y_train.shape)
           print(y test.shape)
          (1604, 40)
          (401, 40)
          (1604, 5)
          (401, 5)
In [46]: import tensorflow as tf
In [47]: from tensorflow.keras.models import Sequential
           from tensorflow.keras.layers import Dense, Activation, Dropout, BatchNormalization, Flatten
           from tensorflow.keras.optimizers import Adam
           from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
           from datetime import datetime
           from sklearn import metrics
In [54]: model=Sequential()
           ###first layer
           model.add(Dense(100,input_shape=(40,)))
           model.add(Activation('relu'))
           model.add(Dropout(0.3))
           ###second layer
           model.add(Dense(200))
           model.add(Activation('relu'))
           model.add(Dropout(0.3))
           ###third layer
           model.add(Dense(100))
           model.add(Activation('relu'))
           model.add(Dropout(0.3))
           ###final layer
           model.add(Dense(5))
           model.add(Activation('softmax'))
In [55]: model.summary()
```

Model: "sequential 1"

Layer (type)	Output Shape	Param #
dense_4 (Dense)	(None, 100)	4,100
activation_4 (Activation)	(None, 100)	0
dropout_3 (Dropout)	(None, 100)	0
dense_5 (Dense)	(None, 200)	20,200
activation_5 (Activation)	(None, 200)	0
dropout_4 (Dropout)	(None, 200)	0
dense_6 (Dense)	(None, 100)	20,100
activation_6 (Activation)	(None, 100)	0
dropout_5 (Dropout)	(None, 100)	0
dense_7 (Dense)	(None, 5)	505
activation_7 (Activation)	(None, 5)	0

Total params: 44,905 (175.41 KB)
Trainable params: 44,905 (175.41 KB)
Non-trainable params: 0 (0.00 B)

```
In [56]: model.compile(loss='categorical crossentropy',metrics=['accuracy'],optimizer='adam')
In [57]: ## Trianing my model
                  from tensorflow.keras.callbacks import ModelCheckpoint
                  from datetime import datetime
                  num_epochs = 150
                  num batch size = 32
                  checkpointer = ModelCheckpoint(filepath='saved models/Asthma audioclassification.keras',
                                                                             verbose=1, save_best_only=True)
                  start = datetime.now()
                  \label{eq:history} \textbf{history} = \textbf{model.fit}(\textbf{X\_train}, \ \textbf{y\_train}, \ \textbf{batch\_size=num\_batch\_size}, \ \textbf{epochs=num\_epochs}, \ \textbf{validation\_data=}(\textbf{X\_test}, \ \textbf{y\_test}, \ \textbf{y\_test},
                  duration = datetime.now() - start
                  print("Training time", duration)
                Epoch 1/150
                                                                   - 0s 1ms/step - accuracy: 0.1919 - loss: 33.8253
                Epoch 1: val_loss improved from inf to 1.90047, saving model to saved_models/Asthma_audioclassification.keras
                                                                   - 1s 5ms/step - accuracy: 0.1979 - loss: 29.9292 - val accuracy: 0.2045 - val loss: 1.9
                51/51
                005
                Epoch 2/150
                                                                  - 0s 2ms/step - accuracy: 0.2280 - loss: 7.1544
                34/51
                Epoch 2: val loss improved from 1.90047 to 1.57572, saving model to saved models/Asthma audioclassification.kera
                                                                 — 0s 3ms/step - accuracy: 0.2259 - loss: 6.6593 - val accuracy: 0.2594 - val loss: 1.57
                51/51
                57
                Epoch 3/150
                40/51
                                                                  - 0s 1ms/step - accuracy: 0.2595 - loss: 2.9680
                Epoch 3: val loss did not improve from 1.57572
                                                                   - 0s 2ms/step - accuracy: 0.2558 - loss: 2.9096 - val_accuracy: 0.1995 - val_loss: 1.61
                51/51
                04
                Epoch 4/150
                36/51
                                                                  - 0s 1ms/step - accuracy: 0.2016 - loss: 2.0635
                Epoch 4: val loss did not improve from 1.57572
                                                                   - 0s 2ms/step - accuracy: 0.2081 - loss: 2.0532 - val accuracy: 0.1970 - val loss: 1.61
                51/51
                14
                Epoch 5/150
                39/51 •
                                                                  0s 1ms/step - accuracy: 0.2535 - loss: 1.7592
                Epoch 5: val loss did not improve from 1.57572
                51/51
                                                                   - 0s 2ms/step - accuracy: 0.2485 - loss: 1.7707 - val accuracy: 0.2045 - val loss: 1.61
                02
                Epoch 6/150
                33/51
                                                                  - 0s 2ms/step - accuracy: 0.2104 - loss: 1.7303
                Epoch 6: val loss did not improve from 1.57572
                                                                   - 0s 2ms/step - accuracy: 0.2168 - loss: 1.7351 - val_accuracy: 0.1895 - val_loss: 1.61
                51/51
                17
                Epoch 7/150
                                                                  - 0s 1ms/step - accuracy: 0.2187 - loss: 1.7020
                37/51 -
                Epoch 7: val_loss did not improve from 1.57572
                                                                   - 0s 2ms/step - accuracy: 0.2229 - loss: 1.6982 - val_accuracy: 0.1895 - val loss: 1.61
                51/51
                18
                Fnoch 8/150
                47/51 -
                                                                  - 0s 2ms/step - accuracy: 0.2486 - loss: 1.6435
                Epoch 8: val loss did not improve from 1.57572
                51/51 •
                                                                   - 0s 3ms/step - accuracy: 0.2482 - loss: 1.6437 - val accuracy: 0.1895 - val loss: 1.61
                18
                Epoch 9/150
                36/51
                                                                  - 0s 1ms/step - accuracy: 0.2280 - loss: 1.6209
                Epoch 9: val_loss did not improve from 1.57572
                                                                   - 0s 2ms/step - accuracy: 0.2315 - loss: 1.6245 - val accuracy: 0.1970 - val loss: 1.61
                51/51
                04
                Epoch 10/150
                                                                  - 0s 1ms/step - accuracy: 0.2241 - loss: 1.5912
                35/51
                Epoch 10: val loss did not improve from 1.57572
                51/51
                                                                   - 0s 2ms/step - accuracy: 0.2292 - loss: 1.5958 - val_accuracy: 0.2195 - val_loss: 1.60
                28
                Epoch 11/150
                                                                  - 0s 1ms/step - accuracy: 0.2510 - loss: 1.6315
                37/51
                Epoch 11: val loss improved from 1.57572 to 1.54338, saving model to saved models/Asthma audioclassification.ker
                as
                                                                  - 0s 3ms/step - accuracy: 0.2510 - loss: 1.6224 - val accuracy: 0.2843 - val loss: 1.54
                51/51
                34
                Epoch 12/150
                36/51
                                                                — 0s 1ms/step - accuracy: 0.2553 - loss: 1.5894
```

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Epoch 12: val loss did not improve from 1.54338
                         — 0s 2ms/step - accuracy: 0.2565 - loss: 1.5924 - val accuracy: 0.2768 - val loss: 1.54
51/51 -
69
Epoch 13/150
37/51 -
                         — 0s 1ms/step - accuracy: 0.2779 - loss: 1.5540
Epoch 13: val loss improved from 1.54338 to 1.53995, saving model to saved models/Asthma audioclassification.ker
as
                        — 0s 4ms/step - accuracy: 0.2777 - loss: 1.5581 - val accuracy: 0.2868 - val loss: 1.54
51/51
00
Epoch 14/150
38/51
                         — 0s 1ms/step - accuracy: 0.2937 - loss: 1.5649
Epoch 14: val_loss improved from 1.53995 to 1.50121, saving model to saved_models/Asthma_audioclassification.ker
                         — 0s 3ms/step - accuracy: 0.2932 - loss: 1.5653 - val accuracy: 0.3192 - val loss: 1.50
51/51
12
Epoch 15/150
                         — 0s 1ms/step - accuracy: 0.3054 - loss: 1.5310
37/51 •
Epoch \ 15: \ val\_loss \ improved \ from \ 1.50121 \ to \ 1.45567, \ saving \ model \ to \ saved\_models/Asthma\_audioclassification.ker
                         — 0s 3ms/step - accuracy: 0.3047 - loss: 1.5318 - val_accuracy: 0.3516 - val_loss: 1.45
51/51 -
57
Epoch 16/150
34/51 -
                   • Os 2ms/step - accuracy: 0.3296 - loss: 1.4904
Epoch 16: val loss improved from 1.45567 to 1.43787, saving model to saved_models/Asthma_audioclassification.ker
51/51
                         - 0s 3ms/step - accuracy: 0.3304 - loss: 1.4949 - val accuracy: 0.3616 - val loss: 1.43
79
Epoch 17/150
                        — 0s 2ms/step - accuracy: 0.2881 - loss: 1.4970
32/51 -
Epoch 17: val_loss did not improve from 1.43787
                         — 0s 2ms/step - accuracy: 0.3007 - loss: 1.4893 - val accuracy: 0.3566 - val loss: 1.44
51/51
08
Epoch 18/150
                         — 0s 1ms/step - accuracy: 0.3275 - loss: 1.4975
36/51
Epoch 18: val loss improved from 1.43787 to 1.41187, saving model to saved models/Asthma audioclassification.ker
as
51/51
                         — 0s 3ms/step - accuracy: 0.3292 - loss: 1.4962 - val accuracy: 0.3641 - val loss: 1.41
19
Epoch 19/150
30/51 -
                         — 0s 4ms/step - accuracy: 0.3426 - loss: 1.4598
Epoch 19: val loss improved from 1.41187 to 1.35668, saving model to saved models/Asthma audioclassification.ker
as
51/51 •
                         — 0s 4ms/step - accuracy: 0.3491 - loss: 1.4493 - val accuracy: 0.4314 - val loss: 1.35
67
Epoch 20/150
                         — 0s 2ms/step - accuracy: 0.3798 - loss: 1.4249
34/51 -
Epoch 20: val loss improved from 1.35668 to 1.34865, saving model to saved models/Asthma audioclassification.ker
51/51 -
                         — 0s 3ms/step - accuracy: 0.3736 - loss: 1.4277 - val accuracy: 0.4289 - val loss: 1.34
86
Epoch 21/150
                         - 0s 1ms/step - accuracy: 0.3855 - loss: 1.4279
Epoch 21: val loss improved from 1.34865 to 1.25639, saving model to saved models/Asthma audioclassification.ker
as
51/51
                      ——— 0s 3ms/step - accuracy: 0.3852 - loss: 1.4213 - val accuracy: 0.4663 - val loss: 1.25
64
Epoch 22/150
33/51
                         - 0s 2ms/step - accuracy: 0.3726 - loss: 1.3987
Epoch 22: val loss improved from 1.25639 to 1.23032, saving model to saved models/Asthma audioclassification.ker
51/51 -
                        — 0s 3ms/step - accuracy: 0.3703 - loss: 1.3985 - val accuracy: 0.4464 - val loss: 1.23
03
Epoch 23/150
                         - 0s 2ms/step - accuracy: 0.4124 - loss: 1.3430
Epoch 23: val_loss did not improve from 1.23032
51/51 -
                         - 0s 2ms/step - accuracy: 0.4071 - loss: 1.3457 - val accuracy: 0.4863 - val loss: 1.23
71
Epoch 24/150
37/51 -
                      Os 1ms/step - accuracy: 0.4044 - loss: 1.3354
Epoch 24: val loss improved from 1.23032 to 1.20787, saving model to saved models/Asthma audioclassification.ker
51/51 •
                         — 0s 3ms/step - accuracy: 0.4024 - loss: 1.3306 - val accuracy: 0.4613 - val loss: 1.20
79
Epoch 25/150
                         — 0s 1ms/step - accuracy: 0.4002 - loss: 1.3454
Epoch 25: val loss improved from 1.20787 to 1.19131, saving model to saved models/Asthma audioclassification.ker
as
51/51
                      —— 0s 3ms/step - accuracy: 0.4060 - loss: 1.3346 - val accuracy: 0.4589 - val loss: 1.19
13
Epoch 26/150
                         — 0s 2ms/step - accuracy: 0.4328 - loss: 1.2383
27/51
Epoch 26: val loss improved from 1.19131 to 1.17584, saving model to saved models/Asthma audioclassification.ker
```

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51/51 -
                     ——— 0s 3ms/step - accuracy: 0.4300 - loss: 1.2512 - val accuracy: 0.5112 - val loss: 1.17
58
Epoch 27/150
30/51
                       Os 2ms/step - accuracy: 0.4569 - loss: 1.2735
Epoch 27: val loss improved from 1.17584 to 1.13784, saving model to saved models/Asthma audioclassification.ker
51/51
                   ———— 0s 3ms/step - accuracy: 0.4488 - loss: 1.2752 - val accuracy: 0.5312 - val loss: 1.13
78
Epoch 28/150
35/51 -
                         — 0s 1ms/step - accuracy: 0.4806 - loss: 1.2675
Epoch 28: val loss improved from 1.13784 to 1.12395, saving model to saved models/Asthma audioclassification.ker
51/51
                        — 0s 3ms/step - accuracy: 0.4772 - loss: 1.2651 - val accuracy: 0.5461 - val loss: 1.12
40
Epoch 29/150
                      Os 3ms/step - accuracy: 0.4664 - loss: 1.2388
38/51 -
Epoch 29: val loss improved from 1.12395 to 1.08826, saving model to saved models/Asthma audioclassification.ker
as
51/51
                         – 0s 4ms/step - accuracy: 0.4658 - loss: 1.2370 - val accuracy: 0.5436 - val loss: 1.08
83
Epoch 30/150
                        — 0s 1ms/step - accuracy: 0.4941 - loss: 1.2220
39/51
Epoch 30: val loss improved from 1.08826 to 1.04719, saving model to saved models/Asthma audioclassification.ker
51/51
                         — 0s 3ms/step - accuracy: 0.4945 - loss: 1.2091 - val accuracy: 0.6010 - val loss: 1.04
72
Epoch 31/150
                        — 0s 1ms/step - accuracy: 0.4887 - loss: 1.1980
37/51
Epoch 31: val loss improved from 1.04719 to 0.98053, saving model to saved models/Asthma audioclassification.ker
51/51
                        — 0s 3ms/step - accuracy: 0.4950 - loss: 1.1862 - val accuracy: 0.6384 - val loss: 0.98
05
Epoch 32/150
                         - 0s 1ms/step - accuracy: 0.5403 - loss: 1.0857
43/51
Epoch 32: val loss improved from 0.98053 to 0.92926, saving model to saved models/Asthma audioclassification.ker
as
51/51
                         — 0s 2ms/step - accuracy: 0.5428 - loss: 1.0859 - val accuracy: 0.6733 - val loss: 0.92
93
Epoch 33/150
29/51 -
                         — 0s 4ms/step - accuracy: 0.5685 - loss: 1.0854
Epoch 33: val loss improved from 0.92926 to 0.88784, saving model to saved models/Asthma audioclassification.ker
as
51/51 •
                         — 0s 4ms/step - accuracy: 0.5813 - loss: 1.0626 - val accuracy: 0.7307 - val loss: 0.88
78
Epoch 34/150
                         — 0s 2ms/step - accuracy: 0.6350 - loss: 0.9883
27/51 -
Epoch 34: val loss improved from 0.88784 to 0.81525, saving model to saved models/Asthma audioclassification.ker
51/51 -
                         — 0s 3ms/step - accuracy: 0.6268 - loss: 0.9948 - val accuracy: 0.7631 - val loss: 0.81
53
Epoch 35/150
                         - 0s 2ms/step - accuracy: 0.6090 - loss: 0.9937
Epoch 35: val loss improved from 0.81525 to 0.73927, saving model to saved models/Asthma audioclassification.ker
as
51/51
                      ——— 0s 3ms/step - accuracy: 0.6153 - loss: 0.9816 - val accuracy: 0.7681 - val loss: 0.73
93
Epoch 36/150
42/51
                         - 0s 1ms/step - accuracy: 0.6365 - loss: 0.9046
Epoch 36: val loss improved from 0.73927 to 0.72723, saving model to saved models/Asthma audioclassification.ker
51/51 -
                        — 0s 3ms/step - accuracy: 0.6370 - loss: 0.9105 - val accuracy: 0.7756 - val loss: 0.72
72
Epoch 37/150
                         - 0s 1ms/step - accuracy: 0.6764 - loss: 0.8272
Epoch \ 37: \ val\_loss \ improved \ from \ 0.72723 \ to \ 0.67774, \ saving \ model \ to \ saved\_models/Asthma\_audioclassification.ker
as
51/51
                         — 0s 3ms/step - accuracy: 0.6739 - loss: 0.8416 - val_accuracy: 0.7955 - val_loss: 0.67
77
Epoch 38/150
                         — 0s 1ms/step - accuracy: 0.6873 - loss: 0.8505
Epoch 38: val loss improved from 0.67774 to 0.62924, saving model to saved_models/Asthma_audioclassification.ker
                         – 0s 2ms/step - accuracy: 0.6894 - loss: 0.8483 - val accuracy: 0.7805 - val loss: 0.62
51/51 -
92
Epoch 39/150
30/51 -
                         — 0s 2ms/step - accuracy: 0.6784 - loss: 0.8470
Epoch 39: val_loss did not improve from 0.62924
51/51
                         — 0s 2ms/step - accuracy: 0.6826 - loss: 0.8466 - val accuracy: 0.8055 - val loss: 0.65
69
Epoch 40/150
35/51 -
                         — 0s 2ms/step - accuracy: 0.7189 - loss: 0.7647
Epoch 40: val loss did not improve from 0.62924
```

- 0s 2ms/step - accuracy: 0.7173 - loss: 0.7733 - val_accuracy: 0.7805 - val_loss: 0.63

51/51 •

```
Epoch 41/150
42/51
                Os 1ms/step - accuracy: 0.7245 - loss: 0.7611
Epoch 41: val loss improved from 0.62924 to 0.61173, saving model to saved models/Asthma audioclassification.ker
51/51 -
                        — 0s 3ms/step - accuracy: 0.7259 - loss: 0.7609 - val accuracy: 0.8005 - val loss: 0.61
17
Epoch 42/150
                      —— 0s 1ms/step - accuracy: 0.7449 - loss: 0.7524
Epoch 42: val_loss improved from 0.61173 to 0.58295, saving model to saved_models/Asthma_audioclassification.ker
as
51/51
                        — 0s 2ms/step - accuracy: 0.7433 - loss: 0.7501 - val_accuracy: 0.8130 - val_loss: 0.58
29
Epoch 43/150
31/51 -
                       — 0s 2ms/step - accuracy: 0.7424 - loss: 0.6936
Epoch 43: val_loss improved from 0.58295 to 0.58018, saving model to saved_models/Asthma_audioclassification.ker
                        – 0s 3ms/step - accuracy: 0.7410 - loss: 0.7067 - val accuracy: 0.7980 - val loss: 0.58
51/51 -
02
Epoch 44/150
39/51 -
                        — 0s 1ms/step - accuracy: 0.7454 - loss: 0.6948
Epoch 44: val_loss did not improve from 0.58018
                        — 0s 3ms/step - accuracy: 0.7465 - loss: 0.6982 - val accuracy: 0.7805 - val loss: 0.63
03
Epoch 45/150
38/51 -
                        — 0s 1ms/step - accuracy: 0.7476 - loss: 0.6838
Epoch 45: val loss improved from 0.58018 to 0.55470, saving model to saved models/Asthma audioclassification.ker
                        — 0s 3ms/step - accuracy: 0.7463 - loss: 0.6894 - val accuracy: 0.8180 - val loss: 0.55
51/51
47
Epoch 46/150
28/51 -
                        — 0s 2ms/step - accuracy: 0.7671 - loss: 0.6079
Epoch 46: val loss did not improve from 0.55470
                        – 0s 2ms/step - accuracy: 0.7559 - loss: 0.6367 - val accuracy: 0.8030 - val loss: 0.57
51/51 -
12
Epoch 47/150
                    • Os 3ms/step - accuracy: 0.7527 - loss: 0.6890
40/51 -
Epoch 47: val loss did not improve from 0.55470
                      Epoch 48/150
36/51 -
                        — 0s 1ms/step - accuracy: 0.7479 - loss: 0.6640
Epoch 48: val loss improved from 0.55470 to 0.53316, saving model to saved models/Asthma audioclassification.ker
51/51 -
                        — 0s 3ms/step - accuracy: 0.7497 - loss: 0.6576 - val accuracy: 0.8030 - val loss: 0.53
32
Epoch 49/150
                        — 0s 2ms/step - accuracy: 0.7161 - loss: 0.7330
33/51 -
Epoch 49: val loss improved from 0.53316 to 0.51362, saving model to saved models/Asthma audioclassification.ker
51/51 -
                        — 0s 3ms/step - accuracy: 0.7343 - loss: 0.6966 - val accuracy: 0.8229 - val loss: 0.51
36
Epoch 50/150
37/51
                  -- 0s 3ms/step - accuracy: 0.7803 - loss: 0.6262
Epoch 50: val loss did not improve from 0.51362
                        – 0s 3ms/step - accuracy: 0.7801 - loss: 0.6250 - val accuracy: 0.8329 - val loss: 0.52
51/51 -
17
Epoch 51/150
                       — 0s 1ms/step - accuracy: 0.7704 - loss: 0.5902
Epoch 51: val loss improved from 0.51362 to 0.50181, saving model to saved_models/Asthma_audioclassification.ker
51/51
                    ——— 0s 3ms/step - accuracy: 0.7698 - loss: 0.5968 - val accuracy: 0.8229 - val loss: 0.50
18
Epoch 52/150
30/51 -
                        — 0s 2ms/step - accuracy: 0.7707 - loss: 0.6720
Epoch 52: val loss did not improve from 0.50181
                         - 0s 2ms/step - accuracy: 0.7721 - loss: 0.6606 - val_accuracy: 0.8254 - val_loss: 0.50
51/51 -
44
Epoch 53/150
                        — 0s 1ms/step - accuracy: 0.7915 - loss: 0.5850
Epoch 53: val_loss improved from 0.50181 to 0.49019, saving model to saved_models/Asthma_audioclassification.ker
                        – 0s 3ms/step - accuracy: 0.7904 - loss: 0.5873 - val accuracy: 0.8304 - val loss: 0.49
51/51 -
02
Epoch 54/150
                        — 0s 1ms/step - accuracy: 0.7705 - loss: 0.6323
Epoch \ 54: \ val\_loss \ improved \ from \ 0.49019 \ to \ 0.48464, \ saving \ model \ to \ saved\_models/Asthma\_audioclassification.ker
                        — 0s 3ms/step - accuracy: 0.7733 - loss: 0.6237 - val_accuracy: 0.8229 - val_loss: 0.48
51/51 -
46
Epoch 55/150
                    ---- 0s 3ms/step - accuracy: 0.7745 - loss: 0.5730
```

Epoch 55: val_loss did not improve from 0.48464

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51/51 -
                    ——— 0s 3ms/step - accuracy: 0.7766 - loss: 0.5767 - val accuracy: 0.8229 - val loss: 0.48
96
Epoch 56/150
35/51
                   Os 1ms/step - accuracy: 0.7827 - loss: 0.5989
Epoch 56: val loss improved from 0.48464 to 0.45922, saving model to saved models/Asthma audioclassification.ker
51/51
                  ——— 0s 3ms/step - accuracy: 0.7863 - loss: 0.5844 - val accuracy: 0.8429 - val loss: 0.45
92
Epoch 57/150
32/51 -
                        — 0s 2ms/step - accuracy: 0.7702 - loss: 0.5618
Epoch 57: val loss improved from 0.45922 to 0.45917, saving model to saved models/Asthma audioclassification.ker
51/51
                        — 0s 4ms/step - accuracy: 0.7699 - loss: 0.5757 - val accuracy: 0.8404 - val loss: 0.45
92
Epoch 58/150
                       — 0s 2ms/step - accuracy: 0.7973 - loss: 0.5737
34/51 -
Epoch 58: val loss did not improve from 0.45917
                         - 0s 2ms/step - accuracy: 0.7972 - loss: 0.5686 - val_accuracy: 0.8454 - val_loss: 0.48
51/51 -
41
Epoch 59/150
34/51 -
                        — 0s 2ms/step - accuracy: 0.7944 - loss: 0.5688
Epoch 59: val_loss did not improve from 0.45917
                        – 0s 2ms/step - accuracy: 0.7941 - loss: 0.5694 - val accuracy: 0.8229 - val loss: 0.46
53
Epoch 60/150
                        — 0s 3ms/step - accuracy: 0.7844 - loss: 0.5714
41/51 -
Epoch 60: val loss did not improve from 0.45917
                         - 0s 4ms/step - accuracy: 0.7879 - loss: 0.5652 - val_accuracy: 0.8579 - val_loss: 0.47
51/51 -
24
Epoch 61/150
                   ----- 0s 1ms/step - accuracy: 0.7807 - loss: 0.5584
37/51 -
Epoch 61: val loss did not improve from 0.45917
                         – 0s 2ms/step - accuracy: 0.7841 - loss: 0.5547 - val accuracy: 0.8329 - val loss: 0.47
15
Epoch 62/150
                        — 0s 2ms/step - accuracy: 0.7808 - loss: 0.5710
35/51 -
Epoch 62: val loss improved from 0.45917 to 0.42703, saving model to saved models/Asthma audioclassification.ker
as
51/51
                     70
Epoch 63/150
30/51 -
                        - 0s 3ms/step - accuracy: 0.7816 - loss: 0.5696
Epoch 63: val loss did not improve from 0.42703
                        – 0s 3ms/step - accuracy: 0.7903 - loss: 0.5512 - val accuracy: 0.8404 - val loss: 0.43
51/51 -
69
Epoch 64/150
34/51 -
                        — 0s 2ms/step - accuracy: 0.7895 - loss: 0.5514
Epoch 64: val loss improved from 0.42703 to 0.42140, saving model to saved models/Asthma audioclassification.ker
51/51 -
                        - 0s 3ms/step - accuracy: 0.7976 - loss: 0.5365 - val_accuracy: 0.8504 - val_loss: 0.42
14
Epoch 65/150
35/51 -
                       — 0s 2ms/step - accuracy: 0.8622 - loss: 0.4169
Epoch 65: val loss did not improve from 0.42140
                      —— 0s 3ms/step - accuracy: 0.8520 - loss: 0.4380 - val accuracy: 0.8678 - val loss: 0.44
51/51
Epoch 66/150
35/51 -
                        - 0s 1ms/step - accuracy: 0.8035 - loss: 0.5364
Epoch 66: val loss did not improve from 0.42140
                         - 0s 2ms/step - accuracy: 0.8073 - loss: 0.5300 - val accuracy: 0.8504 - val loss: 0.47
51/51 -
64
Epoch 67/150
                        — 0s 1ms/step - accuracy: 0.8190 - loss: 0.4781
36/51 -
Epoch 67: val loss did not improve from 0.42140
51/51 -
                         - 0s 2ms/step - accuracy: 0.8213 - loss: 0.4751 - val_accuracy: 0.8678 - val loss: 0.44
80
Epoch 68/150
32/51 •
                        — 0s 3ms/step - accuracy: 0.8395 - loss: 0.4856
Epoch 68: val_loss did not improve from 0.42140
                        – 0s 3ms/step - accuracy: 0.8268 - loss: 0.5005 - val accuracy: 0.8529 - val loss: 0.44
51/51 -
23
Epoch 69/150
37/51 -
                        — 0s 1ms/step - accuracy: 0.8546 - loss: 0.4383
Epoch 69: val loss did not improve from 0.42140
                        - 0s 2ms/step - accuracy: 0.8495 - loss: 0.4464 - val_accuracy: 0.8728 - val_loss: 0.42
51/51 -
50
Epoch 70/150
39/51 -
                      —— 0s 1ms/step - accuracy: 0.8156 - loss: 0.4841
Epoch 70: val loss did not improve from 0.42140
                         - 0s 3ms/step - accuracy: 0.8138 - loss: 0.4871 - val accuracy: 0.8628 - val loss: 0.44
46
Epoch 71/150
```

— 0s 2ms/step - accuracy: 0.8361 - loss: 0.4683

29/51 -

```
Epoch 71: val_loss did not improve from 0.42140
                         — 0s 2ms/step - accuracy: 0.8330 - loss: 0.4797 - val accuracy: 0.8304 - val loss: 0.46
51/51 -
46
Epoch 72/150
39/51 -
                         — 0s 1ms/step - accuracy: 0.8201 - loss: 0.4616
Epoch 72: val loss did not improve from 0.42140
51/51
                         – 0s 2ms/step - accuracy: 0.8205 - loss: 0.4641 - val accuracy: 0.8678 - val loss: 0.43
00
Epoch 73/150
                         — 0s 2ms/step - accuracy: 0.8415 - loss: 0.4466
25/51 -
Epoch 73: val loss did not improve from 0.42140
                          - 0s 2ms/step - accuracy: 0.8349 - loss: 0.4520 - val_accuracy: 0.8703 - val_loss: 0.43
27
Epoch 74/150
37/51 -
                        — 0s 1ms/step - accuracy: 0.8225 - loss: 0.4778
Epoch 74: val loss did not improve from 0.42140
                         - 0s 2ms/step - accuracy: 0.8284 - loss: 0.4662 - val accuracy: 0.8728 - val loss: 0.42
51/51
58
Epoch 75/150
49/51 -
                         - 0s 3ms/step - accuracy: 0.8252 - loss: 0.4454
Epoch 75: val loss did not improve from 0.42140
                         - 0s 3ms/step - accuracy: 0.8249 - loss: 0.4463 - val_accuracy: 0.8853 - val_loss: 0.42
51/51 -
83
Epoch 76/150
                         — 0s 1ms/step - accuracy: 0.8232 - loss: 0.4792
Epoch 76: val loss improved from 0.42140 to 0.39945, saving model to saved models/Asthma audioclassification.ker
51/51 -
                         — 0s 3ms/step - accuracy: 0.8265 - loss: 0.4770 - val accuracy: 0.8828 - val loss: 0.39
94
Epoch 77/150
                      --- 0s 3ms/step - accuracy: 0.8575 - loss: 0.4283
Epoch 77: val loss improved from 0.39945 to 0.37900, saving model to saved models/Asthma audioclassification.ker
51/51 -
                         - 0s 4ms/step - accuracy: 0.8547 - loss: 0.4325 - val accuracy: 0.8703 - val loss: 0.37
90
Epoch 78/150
                        — 0s 1ms/step - accuracy: 0.8286 - loss: 0.4660
Epoch 78: val loss did not improve from 0.37900
                       —— 0s 2ms/step - accuracy: 0.8273 - loss: 0.4674 - val_accuracy: 0.8678 - val_loss: 0.40
51/51
23
Epoch 79/150
40/51 -
                         - 0s 3ms/step - accuracy: 0.8465 - loss: 0.4173
Epoch 79: val loss did not improve from 0.37900
                         – 0s 3ms/step - accuracy: 0.8445 - loss: 0.4214 - val accuracy: 0.8778 - val loss: 0.37
51/51 -
92
Epoch 80/150
38/51 -
                       — 0s 1ms/step - accuracy: 0.8544 - loss: 0.4005
Epoch 80: val loss did not improve from 0.37900
51/51 -
                         - 0s 2ms/step - accuracy: 0.8509 - loss: 0.4128 - val accuracy: 0.8678 - val loss: 0.41
99
Epoch 81/150
40/51 -
                         - 0s 1ms/step - accuracy: 0.8440 - loss: 0.4376
Epoch 81: val loss did not improve from 0.37900
51/51 -
                         - 0s 3ms/step - accuracy: 0.8433 - loss: 0.4371 - val accuracy: 0.8579 - val loss: 0.42
49
Epoch 82/150
37/51 -
                         — 0s 1ms/step - accuracy: 0.8117 - loss: 0.5421
Epoch 82: val loss did not improve from 0.37900
51/51 -
                         - 0s 2ms/step - accuracy: 0.8162 - loss: 0.5275 - val accuracy: 0.8678 - val loss: 0.39
03
Epoch 83/150
29/51
                         — 0s 2ms/step - accuracy: 0.8454 - loss: 0.3988
Epoch 83: val_loss did not improve from 0.37900
                         — 0s 3ms/step - accuracy: 0.8437 - loss: 0.4117 - val accuracy: 0.8678 - val loss: 0.42
51/51
59
Epoch 84/150
                        — 0s 1ms/step - accuracy: 0.8397 - loss: 0.4738
36/51 -
Epoch 84: val loss improved from 0.37900 to 0.36641, saving model to saved models/Asthma audioclassification.ker
51/51 -
                      —— 0s 3ms/step - accuracy: 0.8417 - loss: 0.4606 - val accuracy: 0.8728 - val loss: 0.36
64
Epoch 85/150
42/51
                         0s 1ms/step - accuracy: 0.8420 - loss: 0.4718
Epoch 85: val loss did not improve from 0.36641
                         - 0s 2ms/step - accuracy: 0.8418 - loss: 0.4673 - val_accuracy: 0.8728 - val_loss: 0.40
51/51 -
51
Epoch 86/150
34/51
                        — 0s 3ms/step - accuracy: 0.8219 - loss: 0.4897
Epoch 86: val_loss did not improve from 0.36641
                         - 0s 4ms/step - accuracy: 0.8256 - loss: 0.4724 - val accuracy: 0.8653 - val loss: 0.38
46
Epoch 87/150
```

— 0s 1ms/step - accuracy: 0.8493 - loss: 0.4500

36/51 -

```
Epoch 87: val_loss did not improve from 0.36641
                         — 0s 2ms/step - accuracy: 0.8471 - loss: 0.4456 - val accuracy: 0.8678 - val loss: 0.39
51/51 -
51
Epoch 88/150
27/51 •
                         — 0s 2ms/step - accuracy: 0.8411 - loss: 0.3807
Epoch 88: val loss did not improve from 0.36641
51/51
                         – 0s 3ms/step - accuracy: 0.8418 - loss: 0.3927 - val accuracy: 0.8603 - val loss: 0.41
52
Epoch 89/150
                         — 0s 1ms/step - accuracy: 0.8249 - loss: 0.5041
41/51 -
Epoch 89: val loss did not improve from 0.36641
51/51 •
                          - 0s 2ms/step - accuracy: 0.8249 - loss: 0.5024 - val_accuracy: 0.8753 - val_loss: 0.40
08
Epoch 90/150
43/51 -
                        — 0s 1ms/step - accuracy: 0.8550 - loss: 0.4381
Epoch 90: val loss did not improve from 0.36641
                          - 0s 2ms/step - accuracy: 0.8514 - loss: 0.4424 - val accuracy: 0.8628 - val loss: 0.41
51/51 •
87
Epoch 91/150
34/51 -
                         — 0s 2ms/step - accuracy: 0.8257 - loss: 0.4922
Epoch 91: val loss did not improve from 0.36641
                          - 0s 2ms/step - accuracy: 0.8274 - loss: 0.4720 - val_accuracy: 0.8728 - val_loss: 0.37
51/51 -
87
Epoch 92/150
                         - 0s 2ms/step - accuracy: 0.8263 - loss: 0.4589
Epoch 92: val loss did not improve from 0.36641
51/51 -
                          - 0s 2ms/step - accuracy: 0.8348 - loss: 0.4390 - val accuracy: 0.8703 - val loss: 0.36
89
Epoch 93/150
41/51 -
                       — 0s 1ms/step - accuracy: 0.8495 - loss: 0.4298
Epoch 93: val_loss did not improve from 0.36641
                        — 0s 2ms/step - accuracy: 0.8504 - loss: 0.4262 - val accuracy: 0.8753 - val loss: 0.39
51/51
55
Fnoch 94/150
47/51 -
                        — 0s 2ms/step - accuracy: 0.8352 - loss: 0.4406
Epoch 94: val loss did not improve from 0.36641
                         – 0s 3ms/step - accuracy: 0.8350 - loss: 0.4401 - val accuracy: 0.8678 - val loss: 0.40
57
Epoch 95/150
39/51 -
                         — 0s 1ms/step - accuracy: 0.8485 - loss: 0.4411
Epoch 95: val_loss did not improve from 0.36641
                          - 0s 2ms/step - accuracy: 0.8469 - loss: 0.4417 - val accuracy: 0.8703 - val loss: 0.38
51/51 -
31
Fnoch 96/150
37/51 -
                         - 0s 1ms/step - accuracy: 0.8548 - loss: 0.4014
Epoch 96: val loss improved from 0.36641 to 0.34974, saving model to saved models/Asthma audioclassification.ker
51/51 -
                     ——— 0s 3ms/step - accuracy: 0.8537 - loss: 0.4032 - val accuracy: 0.8853 - val loss: 0.34
97
Epoch 97/150
33/51 -
                         — 0s 3ms/step - accuracy: 0.8520 - loss: 0.3690
Epoch 97: val loss did not improve from 0.34974
51/51 -
                         – 0s 4ms/step - accuracy: 0.8481 - loss: 0.3863 - val accuracy: 0.8828 - val loss: 0.37
57
Epoch 98/150
38/51 -
                         - 0s 1ms/step - accuracy: 0.8441 - loss: 0.4189
Epoch 98: val_loss did not improve from 0.34974
51/51 -
                          - 0s 2ms/step - accuracy: 0.8486 - loss: 0.4066 - val accuracy: 0.8678 - val loss: 0.38
13
Epoch 99/150
46/51 -
                        — 0s 1ms/step - accuracy: 0.8648 - loss: 0.3707
Epoch 99: val loss did not improve from 0.34974
                         – 0s 2ms/step - accuracy: 0.8634 - loss: 0.3758 - val accuracy: 0.8778 - val loss: 0.41
51/51
28
Epoch 100/150
                         - 0s 2ms/step - accuracy: 0.8448 - loss: 0.4287
Epoch 100: val_loss did not improve from 0.34974
51/51 •
                          - 0s 3ms/step - accuracy: 0.8470 - loss: 0.4206 - val accuracy: 0.8703 - val loss: 0.37
73
Epoch 101/150
                        — 0s 1ms/step - accuracy: 0.8380 - loss: 0.4025
39/51
Epoch 101: val loss improved from 0.34974 to 0.33734, saving model to saved models/Asthma audioclassification.ke
ras
51/51
                         — 0s 3ms/step - accuracy: 0.8384 - loss: 0.3996 - val accuracy: 0.8828 - val loss: 0.33
73
Epoch 102/150
37/51 -
                         — 0s 1ms/step - accuracy: 0.8744 - loss: 0.3333
Epoch 102: val loss did not improve from 0.33734
                         - 0s 2ms/step - accuracy: 0.8697 - loss: 0.3420 - val_accuracy: 0.8703 - val loss: 0.38
51/51
85
Epoch 103/150
                         — 0s 2ms/step - accuracy: 0.8282 - loss: 0.4718
```

Epoch 103: val_loss did not improve from 0.33734

```
51/51 -
                      ----- 0s 2ms/step - accuracy: 0.8296 - loss: 0.4599 - val accuracy: 0.8579 - val loss: 0.37
45
Epoch 104/150
                        — 0s 1ms/step - accuracy: 0.8515 - loss: 0.4220
42/51
Epoch 104: val loss did not improve from 0.33734
51/51 -
                         — 0s 2ms/step - accuracy: 0.8471 - loss: 0.4265 - val accuracy: 0.8828 - val loss: 0.37
32
Epoch 105/150
                         - 0s 1ms/step - accuracy: 0.8304 - loss: 0.4479
Epoch 105: val_loss did not improve from 0.33734
51/51 -
                          - 0s 3ms/step - accuracy: 0.8335 - loss: 0.4398 - val accuracy: 0.8554 - val loss: 0.37
69
Epoch 106/150
                       — 0s 1ms/step - accuracy: 0.8517 - loss: 0.3906
42/51 -
Epoch 106: val loss did not improve from 0.33734
                         - 0s 3ms/step - accuracy: 0.8502 - loss: 0.3931 - val_accuracy: 0.8878 - val_loss: 0.38
51/51
25
Epoch 107/150
45/51 -
                         — 0s 1ms/step - accuracy: 0.8325 - loss: 0.4195
Epoch 107: val_loss did not improve from 0.33734
                         – 0s 2ms/step - accuracy: 0.8339 - loss: 0.4188 - val accuracy: 0.8903 - val loss: 0.34
33
Epoch 108/150
39/51
                         — 0s 1ms/step - accuracy: 0.8593 - loss: 0.3645
Epoch 108: val loss did not improve from 0.33734
51/51 -
                          - 0s 2ms/step - accuracy: 0.8581 - loss: 0.3672 - val accuracy: 0.8753 - val loss: 0.40
13
Epoch 109/150
42/51 -
                         - 0s 3ms/step - accuracy: 0.8706 - loss: 0.3584
Epoch 109: val loss did not improve from 0.33734
                         – 0s 3ms/step - accuracy: 0.8678 - loss: 0.3668 - val accuracy: 0.8753 - val loss: 0.37
51/51
27
Epoch 110/150
38/51 -
                         - 0s 1ms/step - accuracy: 0.8543 - loss: 0.3783
Epoch 110: val loss did not improve from 0.33734
                         - 0s 2ms/step - accuracy: 0.8520 - loss: 0.3892 - val_accuracy: 0.8728 - val_loss: 0.38
51/51 -
04
Epoch 111/150
                         — 0s 2ms/step - accuracy: 0.8406 - loss: 0.4490
Epoch 111: val loss did not improve from 0.33734
                         – 0s 3ms/step - accuracy: 0.8472 - loss: 0.4254 - val accuracy: 0.8853 - val loss: 0.35
80
Epoch 112/150
                         — 0s 2ms/step - accuracy: 0.8744 - loss: 0.3588
48/51
Epoch 112: val loss improved from 0.33734 to 0.31969, saving model to saved models/Asthma audioclassification.ke
ras
51/51
                       —— 0s 4ms/step - accuracy: 0.8733 - loss: 0.3600 - val accuracy: 0.8953 - val loss: 0.31
97
Epoch 113/150
46/51 -
                         - 0s 1ms/step - accuracy: 0.8585 - loss: 0.3823
Epoch 113: val loss did not improve from 0.31969
                         - 0s 2ms/step - accuracy: 0.8591 - loss: 0.3814 - val accuracy: 0.8903 - val loss: 0.35
51/51 -
72
Epoch 114/150
43/51
                        — 0s 1ms/step - accuracy: 0.8893 - loss: 0.3259
Epoch 114: val loss did not improve from 0.31969
                         – 0s 2ms/step - accuracy: 0.8867 - loss: 0.3300 - val_accuracy: 0.9002 - val_loss: 0.34
51/51 -
32
Epoch 115/150
                         — 0s 1ms/step - accuracy: 0.8680 - loss: 0.3237
37/51
Epoch 115: val loss did not improve from 0.31969
51/51
                         – 0s 2ms/step - accuracy: 0.8683 - loss: 0.3280 - val accuracy: 0.8653 - val loss: 0.41
78
Epoch 116/150
                         - 0s 1ms/step - accuracy: 0.8447 - loss: 0.3964
36/51 -
Epoch 116: val loss did not improve from 0.31969
                         - 0s 2ms/step - accuracy: 0.8490 - loss: 0.3885 - val_accuracy: 0.9052 - val_loss: 0.32
51/51 -
44
Epoch 117/150
                         — 0s 1ms/step - accuracy: 0.8534 - loss: 0.3843
Epoch 117: val_loss did not improve from 0.31969
                          - 0s 3ms/step - accuracy: 0.8540 - loss: 0.3847 - val accuracy: 0.8953 - val loss: 0.36
51/51 -
71
Epoch 118/150
                         - 0s 1ms/step - accuracy: 0.8619 - loss: 0.3878
41/51
Epoch 118: val loss did not improve from 0.31969
                         – 0s 2ms/step - accuracy: 0.8610 - loss: 0.3871 - val_accuracy: 0.8903 - val loss: 0.33
51/51
89
Epoch 119/150
                         - 0s 1ms/step - accuracy: 0.8539 - loss: 0.3830
Epoch 119: val loss did not improve from 0.31969
                          - 0s 2ms/step - accuracy: 0.8590 - loss: 0.3736 - val accuracy: 0.9077 - val loss: 0.33
51/51 -
```

36

```
Epoch 120/150
38/51 -
                         0s 1ms/step - accuracy: 0.8614 - loss: 0.3886
Epoch 120: val loss did not improve from 0.31969
51/51 -
                         - 0s 3ms/step - accuracy: 0.8606 - loss: 0.3870 - val accuracy: 0.8953 - val loss: 0.36
13
Epoch 121/150
43/51
                        — 0s 1ms/step - accuracy: 0.8675 - loss: 0.3824
Epoch 121: val loss did not improve from 0.31969
51/51
                         – 0s 2ms/step - accuracy: 0.8661 - loss: 0.3817 - val accuracy: 0.8628 - val loss: 0.36
96
Epoch 122/150
31/51
                        — 0s 2ms/step - accuracy: 0.8377 - loss: 0.4009
Epoch 122: val loss did not improve from 0.31969
                         — 0s 2ms/step - accuracy: 0.8451 - loss: 0.3918 - val_accuracy: 0.8878 - val loss: 0.32
51/51 -
Epoch 123/150
43/51 -
                         - 0s 1ms/step - accuracy: 0.8449 - loss: 0.4090
Epoch 123: val loss did not improve from 0.31969
                          - 0s 3ms/step - accuracy: 0.8451 - loss: 0.4089 - val accuracy: 0.8778 - val loss: 0.32
35
Epoch 124/150
43/51 -
                         — 0s 1ms/step - accuracy: 0.8741 - loss: 0.3822
Epoch 124: val loss did not improve from 0.31969
                         – 0s 2ms/step - accuracy: 0.8730 - loss: 0.3808 - val_accuracy: 0.8878 - val<sub>_</sub>loss: 0.32
51/51
22
Epoch 125/150
                         — 0s 1ms/step - accuracy: 0.8724 - loss: 0.3345
Epoch 125: val loss did not improve from 0.31969
                         – 0s 3ms/step - accuracy: 0.8717 - loss: 0.3362 - val accuracy: 0.8978 - val loss: 0.34
51/51
46
Epoch 126/150
47/51
                         — 0s 1ms/step - accuracy: 0.8554 - loss: 0.3589
Epoch 126: val loss did not improve from 0.31969
                         - 0s 2ms/step - accuracy: 0.8548 - loss: 0.3615 - val_accuracy: 0.8504 - val_loss: 0.40
51/51 -
54
Epoch 127/150
                      ____ 0s 1ms/step - accuracy: 0.8429 - loss: 0.4493
Epoch 127: val_loss improved from 0.31969 to 0.31810, saving model to saved_models/Asthma_audioclassification.ke
51/51 -
                      —— 0s 3ms/step - accuracy: 0.8448 - loss: 0.4370 - val accuracy: 0.8953 - val loss: 0.31
81
Epoch 128/150
                         — 0s 2ms/step - accuracy: 0.8403 - loss: 0.4237
Epoch 128: val loss did not improve from 0.31810
                         — 0s 2ms/step - accuracy: 0.8508 - loss: 0.4072 - val accuracy: 0.8928 - val loss: 0.32
51/51 -
36
Epoch 129/150
                         — 0s 1ms/step - accuracy: 0.8811 - loss: 0.3632
46/51 -
Epoch 129: val loss did not improve from 0.31810
51/51 -
                         - 0s 2ms/step - accuracy: 0.8804 - loss: 0.3623 - val_accuracy: 0.9027 - val_loss: 0.34
37
Epoch 130/150
39/51 -
                      0s 1ms/step - accuracy: 0.8932 - loss: 0.3156
Epoch 130: val loss did not improve from 0.31810
51/51
                         — 0s 2ms/step - accuracy: 0.8913 - loss: 0.3184 - val accuracy: 0.8803 - val loss: 0.35
Epoch 131/150
                         - 0s 2ms/step - accuracy: 0.8471 - loss: 0.4151
32/51 -
Epoch 131: val loss did not improve from 0.31810
                          - 0s 4ms/step - accuracy: 0.8547 - loss: 0.4015 - val accuracy: 0.8853 - val loss: 0.34
51/51 -
27
Epoch 132/150
                         - 0s 2ms/step - accuracy: 0.8792 - loss: 0.3355
31/51
Epoch 132: val loss improved from 0.31810 to 0.31356, saving model to saved models/Asthma audioclassification.ke
ras
51/51 -
                        — 0s 3ms/step - accuracy: 0.8775 - loss: 0.3399 - val_accuracy: 0.8953 - val_loss: 0.31
36
Epoch 133/150
43/51
                         — 0s 1ms/step - accuracy: 0.8780 - loss: 0.3313
Epoch 133: val loss did not improve from 0.31356
                         – 0s 2ms/step - accuracy: 0.8796 - loss: 0.3279 - val_accuracy: 0.8828 - val_loss: 0.32
51/51
Epoch 134/150
44/51 •
                         — 0s 1ms/step - accuracy: 0.8802 - loss: 0.3090
Epoch 134: val loss did not improve from 0.31356
51/51 -
                         – 0s 3ms/step - accuracy: 0.8794 - loss: 0.3127 - val accuracy: 0.8928 - val loss: 0.34
14
Epoch 135/150
35/51 -
                         — 0s 2ms/step - accuracy: 0.8676 - loss: 0.3291
Epoch 135: val loss did not improve from 0.31356
51/51 -
                          - 0s 2ms/step - accuracy: 0.8694 - loss: 0.3299 - val_accuracy: 0.8953 - val_loss: 0.31
```

Epoch 136/150

```
- 0s 1ms/step - accuracy: 0.8790 - loss: 0.3142
        Epoch 136: val loss did not improve from 0.31356
                                  - 0s 2ms/step - accuracy: 0.8776 - loss: 0.3176 - val accuracy: 0.8853 - val loss: 0.37
        79
        Epoch 137/150
        40/51
                                 — 0s 1ms/step - accuracy: 0.8837 - loss: 0.3114
        Epoch 137: val loss did not improve from 0.31356
        51/51
                                  – 0s 2ms/step - accuracy: 0.8843 - loss: 0.3105 - val accuracy: 0.8903 - val loss: 0.34
        43
        Epoch 138/150
        27/51
                                  - 0s 3ms/step - accuracy: 0.8869 - loss: 0.3012
        Epoch 138: val_loss did not improve from 0.31356
                                  - 0s 3ms/step - accuracy: 0.8787 - loss: 0.3179 - val accuracy: 0.8928 - val loss: 0.37
        51/51 -
        52
        Epoch 139/150
                                 — 0s 1ms/step - accuracy: 0.8505 - loss: 0.3914
        43/51
        Epoch 139: val loss did not improve from 0.31356
                                  - 0s 2ms/step - accuracy: 0.8535 - loss: 0.3846 - val_accuracy: 0.8878 - val_loss: 0.35
        51/51 -
        05
        Epoch 140/150
        43/51 -
                                  - 0s 1ms/step - accuracy: 0.8705 - loss: 0.3484
        Epoch 140: val_loss did not improve from 0.31356
                                  – 0s 2ms/step - accuracy: 0.8706 - loss: 0.3462 - val accuracy: 0.8928 - val loss: 0.32
        51/51
        50
        Epoch 141/150
        39/51 -
                                  - 0s 1ms/step - accuracy: 0.8746 - loss: 0.3370
        Epoch 141: val loss did not improve from 0.31356
                                  - 0s 2ms/step - accuracy: 0.8749 - loss: 0.3355 - val_accuracy: 0.9002 - val_loss: 0.34
        51/51 -
        02
        Epoch 142/150
        42/51 -
                                 — 0s 1ms/step - accuracy: 0.8777 - loss: 0.3379
        Epoch 142: val loss did not improve from 0.31356
                                  - 0s 2ms/step - accuracy: 0.8776 - loss: 0.3406 - val accuracy: 0.8978 - val loss: 0.32
        33
        Epoch 143/150
                                 — 0s 2ms/step - accuracy: 0.8529 - loss: 0.3707
        33/51
        Epoch 143: val loss did not improve from 0.31356
                                  - 0s 3ms/step - accuracy: 0.8581 - loss: 0.3632 - val_accuracy: 0.8579 - val loss: 0.39
        51/51 -
        60
        Epoch 144/150
        46/51 -
                                  - 0s 1ms/step - accuracy: 0.8429 - loss: 0.4176
        Epoch 144: val loss did not improve from 0.31356
        51/51 -
                                  - 0s 2ms/step - accuracy: 0.8454 - loss: 0.4117 - val accuracy: 0.8903 - val loss: 0.32
        72
        Epoch 145/150
                                 — 0s 1ms/step - accuracy: 0.8815 - loss: 0.3589
        45/51 -
        Epoch 145: val loss improved from 0.31356 to 0.30762, saving model to saved models/Asthma audioclassification.ke
        ras
        51/51 -
                                 — 0s 4ms/step - accuracy: 0.8817 - loss: 0.3566 - val accuracy: 0.9052 - val loss: 0.30
        76
        Epoch 146/150
        39/51
                                  - 0s 1ms/step - accuracy: 0.8735 - loss: 0.3328
        Epoch 146: val loss did not improve from 0.30762
        51/51 •
                                  – 0s 2ms/step - accuracy: 0.8727 - loss: 0.3307 - val accuracy: 0.8978 - val loss: 0.33
        82
        Epoch 147/150
        37/51 -
                                  - 0s 1ms/step - accuracy: 0.8843 - loss: 0.2964
        Epoch 147: val loss did not improve from 0.30762
        51/51 -
                                  - 0s 2ms/step - accuracy: 0.8850 - loss: 0.2965 - val accuracy: 0.8978 - val loss: 0.33
        61
        Epoch 148/150
        25/51
                                 — 0s 2ms/step - accuracy: 0.8851 - loss: 0.3296
        Epoch 148: val_loss did not improve from 0.30762
                                  – 0s 2ms/step - accuracy: 0.8714 - loss: 0.3589 - val accuracy: 0.8878 - val loss: 0.34
        51/51
        69
        Epoch 149/150
        36/51
                                 — 0s 1ms/step - accuracy: 0.8650 - loss: 0.3601
        Epoch 149: val loss did not improve from 0.30762
        51/51
                                 — 0s 2ms/step - accuracy: 0.8653 - loss: 0.3586 - val_accuracy: 0.9077 - val_loss: 0.36
        44
        Epoch 150/150
        43/51 -
                                  - 0s 1ms/step - accuracy: 0.8573 - loss: 0.3610
        Epoch 150: val loss did not improve from 0.30762
                                  - 0s 2ms/step - accuracy: 0.8597 - loss: 0.3550 - val_accuracy: 0.9002 - val_loss: 0.38
        51/51 -
        48
        Training time 0:00:27.224420
In [58]: model.evaluate(X_test,y_test,verbose=0)
Out[58]: [0.3847678303718567, 0.9002493619918823]
```

In [59]: filename="Asthma Detection Dataset Version 2\\asthma\\Asthma_RS_20 (1).wav"
y, sr = librosa.load(filename, duration=3, offset=0.5)

```
mfcc = np.mean(librosa.feature.mfcc(y=y, sr=sr, n_mfcc=40).T, axis=0)
         mfcc=mfcc.reshape(1,-1)
         print(mfcc)
         print(mfcc.shape)
         predicted_label=np.argmax(model.predict(mfcc), axis=-1)
         print(predicted label)
         prediction class = labelencoder.inverse transform(predicted label)
         prediction class
        [[-548.83496
                         50.119514 45.78449
                                                  43.384907
                                                                40.99531
            37.92376
                        34.31658
                                     30.903397
                                                  27.728077
                                                                24.546072
                                                 13.384048
            21.37009
                         18.457767
                                     15.855426
                                                                11.001728
             8.931421
                          7.295938
                                      5.979423
                                                   4.9392314
                                                                 4.3285704
             4.1767793
                         4.285747
                                      4.5153503
                                                  4.891299
                                                                 5.3693304
             5.767992
                         5.9868765 6.0878544
                                                  6.109491
                                                                5.9857993
                                                   4.4640946
             5.6918344
                         5.316192
                                       4.9219117
                                                                 3.9046216
             3.3001502
                          2.7046251
                                       2.0866601
                                                  1.4203374
                                                                0.7722891]]
        (1.40)
        1/1 -
                               — 0s 60ms/step
        [0]
Out[59]: array(['asthma'], dtype='<U9')</pre>
In [60]: filename="Asthma Detection Dataset Version 2\Bronchial\Bronchial Sc 1 (1).wav"
         y, sr = librosa.load(filename, duration=3, offset=0.5)
         mfcc = np.mean(librosa.feature.mfcc(y=y, sr=sr, n mfcc=40).T, axis=0)
         mfcc=mfcc.reshape(1,-1)
         print(mfcc)
         print(mfcc.shape)
         predicted label=np.argmax(model.predict(mfcc), axis=-1)
         print(predicted label)
         prediction class = labelencoder.inverse transform(predicted label)
         prediction_class
        [[-3.8983014e+02 1.8178264e+02 2.1946053e+01 -1.5424136e+01 1.3209252e+01 1.8196669e+01 7.7502036e+00 6.2696522e-01
           6.6663593e-01 5.2164898e+00 9.5117359e+00 6.4938064e+00
          -2.1969423e+00 4.3192797e+00 4.2819500e+00 3.9133751e+00
           1.8397973e-01 9.0698242e+00 3.1407807e+00 4.5960984e+00
          -2.8902512e+00 2.9254174e+00 -1.1361605e-01 6.3802347e+00
          -9.9346149e-01 5.8656082e+00 3.1182034e+00 6.5837083e+00
          -2.4030316e+00 1.1249828e+00 -2.7190251e+00 3.4650650e+00
           (1.40)
        1/1
                               0s 22ms/step
        [1]
Out[60]: array(['bronchial'], dtype='<U9')</pre>
In [61]: filename="Asthma Detection Dataset Version 2\Bronchial\Bronchial Sc 1 (8).wav"
         y, sr = librosa.load(filename, duration=3, offset=0.5)
         mfcc = np.mean(librosa.feature.mfcc(y=y, sr=sr, n_mfcc=40).T, axis=0)
         mfcc=mfcc.reshape(1,-1)
         print(mfcc)
         print(mfcc.shape)
         predicted_label=np.argmax(model.predict(mfcc), axis=-1)
         print(predicted label)
         prediction_class = labelencoder.inverse_transform(predicted_label)
         prediction class
        [[-4.5608768e+02 1.7521689e+02 5.2230350e+01 -1.3782601e+00
          -4.3475981e+00 3.0727172e+00 -6.5864635e-01 2.1587160e+00
           1.4168003e+01 2.2654135e+01 1.4551707e+01 3.4448764e+00
           4.7594452e-01 8.8251476e+00 8.9229660e+00 5.2574186e+00
          -1.7476784e+00 2.4893077e-01 2.7135258e+00 7.2230697e+00 5.8852682e+00 7.4056950e+00 2.2178140e+00 1.3237664e+00
           9.3586162e-02 2.8307562e+00 8.4737867e-01 1.9972440e+00
           1.8681623e+00 5.0061994e+00 1.6391280e+00 3.8139215e-01
          -1.1640850e+00 1.9350014e+00 1.4548314e+00 1.4240146e+00
          -7.2621983e-01 9.7914219e-01 1.7796893e+00 3.4094405e+00]]
        (1, 40)
        1/1
                               - 0s 38ms/step
Out[61]: array(['bronchial'], dtype='<U9')</pre>
In [62]: filename="Asthma Detection Dataset Version 2\pneumonia\Pneumonia_10T (5).wav"
         y, sr = librosa.load(filename, duration=3, offset=0.5)
         mfcc = np.mean(librosa.feature.mfcc(y=y, sr=sr, n_mfcc=40).T, axis=0)
         mfcc=mfcc.reshape(1,-1)
         print(mfcc)
         print(mfcc.shape)
         predicted_label=np.argmax(model.predict(mfcc), axis=-1)
         print(predicted label)
         prediction class = labelencoder.inverse transform(predicted label)
```

```
prediction_class
        [[-385.1813
                       113.2246
                                     87.42339
                                                  64.046234
                                                              45.475803
            31.254955
                       19.582722
                                     10.06814
                                                  3.4867895
                                                              1.1309445
            2.2522495
                         4.13407
                                     5.300241
                                                   5.9561915
                                                               6.156539
                        5.2794957 4.7358866
            5.829645
                                                  4.2662005
                                                               3.8962717
            3.2000952
                        2.0528636 1.4345835
                                                  2.102308
                                                               3.0735042
                         2.614724
                                      1.9900856
                                                  1.740515
            3.2031195
                                                               1.8590331
            1.8391153
                         1.2675236
                                      0.5058664
                                                   0.3985395
                                                               0.7895945
            0.881756
                                                 2.1741264
                                                               3.1455822]]
                         0.6754371
                                      1.0320972
        (1, 40)
        1/1
                              - 0s 21ms/step
        [4]
Out[62]: array(['pneumonia'], dtype='<U9')</pre>
In [ ]: filename=
         y, sr = librosa.load(filename, duration=3, offset=0.5)
         mfcc = np.mean(librosa.feature.mfcc(y=y, sr=sr, n_mfcc=40).T, axis=0)
         mfcc=mfcc.reshape(1,-1)
         print(mfcc)
         print(mfcc.shape)
         predicted_label=np.argmax(model.predict(mfcc), axis=-1)
         print(predicted_label)
         prediction_class = labelencoder.inverse_transform(predicted_label)
         prediction_class
```

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