## Lab-1

#### February 22, 2024

#### LAB 1

**AIDI 2000** 

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```
[]: import pandas as pd
import numpy as np
from tensorflow import keras
from sklearn.preprocessing import StandardScaler, LabelEncoder
from tensorflow.keras import layers, Sequential
from sklearn.model_selection import train_test_split
from matplotlib import pyplot as plt
```

Import Dataset

```
[]: data = pd.read_csv('spotify_data_12_20_2023.csv', dtype="object")
    data.dropna(inplace=True)
    print(data.columns)
```

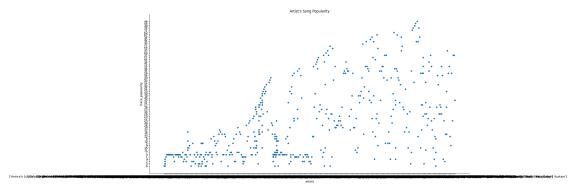
Describe Dataset

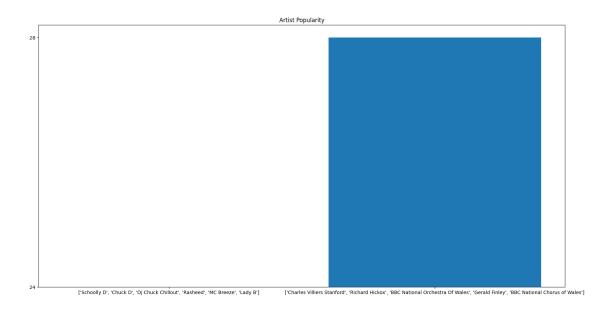
### []: print(data.describe())

	album_id	album_name	album_popularity	\
count	583	583	583	
unique	181	156	71	
top	6Mzrj457wfDueNk3eZJ3Z8	Handel: St. John Passion	6	

```
freq
                                  39
                                                             39
                                                                               40
                                                                    artists
           album_type
                   583
                                                                        583
    count
                     2
    unique
                                                                         348
                 album
                        ['George Frideric Handel', 'Martin Kleitmann',...
    top
    freq
                   460
                           artist_0
                                                                     artist_1 \
    count
                                583
                                                                          583
                                 107
    unique
                                                                           183
    top
            George Frideric Handel
                                     Choir of St. John's College, Cambridge
                                 128
    freq
                      artist_2
                                     artist_3
                                                   artist_4 ... time_signature
    count
                           583
                                          583
                                                         583
                                                                            583
    unique
                           187
                                          204
                                                         244
                                                                              4
                                                                              4
    top
            Andrew Nethsingha
                                Mária Zádori
                                               Judit Nemeth ...
    freq
                            41
                                           39
                                                          39
                                                                            416
                                                      track href
                                                                             type \
    count
                                                             583
                                                                              583
                                                             583
    unique
                                                                                1
    top
            https://api.spotify.com/v1/tracks/6fEJOpOzapwJ... audio_features
    freq
                                                               1
                                                                              583
                                               uri valence explicit \
                                                                 583
                                               583
                                                        583
    count
                                                        427
    unique
                                               583
    top
             spotify:track:6fEJOpOzapwJn7gpfekiBY
                                                    0.0354
                                                               false
                                                          5
                                                                 494
    freq
           track_popularity release_year release_month
                                                      583 583
                         583
                                       583
    count
                          65
                                        44
                                                       12
    unique
                                                             1
                                      2014
    top
                           0
                                                 January
                                        53
                                                      136 583
    freq
                         111
    [4 rows x 49 columns]
    Plot Graphs
[]: df = pd.DataFrame(data)
     df.plot(kind='scatter', x='artists', y='track_popularity', figsize=(20, 10),
             title="Artist's Song Popularity")
     plt.gca().spines[['top', 'right']].set_visible(False)
```

```
x1 = df['artists']
y1 = df['artist_popularity']
plt.figure(figsize =(20, 10))
plt.bar(x1[0:10], y1[0:10])
plt.title('Artist Popularity')
plt.show()
```





## Select Fields for data processing

```
X_categorical = df[categorical_features]
X_numeric = df[numeric_features]
```

Prepare the Data for Processing

```
[]: scaler = StandardScaler()
X_numeric_scaled = scaler.fit_transform(X_numeric)

label_encoders = {}
X_categorical_encoded = pd.DataFrame()

for feature in categorical_features:
    label_encoder = LabelEncoder()
    X_categorical_encoded[feature] = label_encoder.fit_transform(df[feature])
    label_encoders[feature] = label_encoder
```

Split the data

Analyse Model Shape

```
[]: # View the Model Shape
print(X_train.shape)
print(y_train.shape)
```

(466, 16) (466,)

```
[]: print(X_test.shape)
print(y_test.shape)
```

```
(117, 16)
(117,)
```

Reshape the data

```
[]: X_train = X_train.to_numpy().reshape(-1, 16, 1)
X_test = X_test.to_numpy().reshape(-1, 16, 1)

# Preprocess y_test data
y_test = y_test.to_numpy().reshape(-1, 1) # Reshape to a column vector
```

```
Analyse Model Shape
[]: # View the Model Shape
     print(X train.shape)
     print(y_train.shape)
    (466, 16, 1)
    (466, 89)
[]: print(X_test.shape)
    print(y_test.shape)
    (117, 16, 1)
    (117, 89)
    Build and execute the model.
[]: # Defining the model
     model = Sequential([
             layers.Input(shape=(16, 1)),
         layers.Conv1D(32, kernel_size=3, activation='relu'),
         layers.Dense(32, activation='relu'),
         layers.MaxPooling1D(pool_size=2),
         layers.Conv1D(32, kernel_size=3, activation='relu'),
         layers.Dense(32, activation='relu'),
         layers.Flatten(),
         layers.Dense(32, activation='relu'),
         layers.Dense(y_test.shape[1], activation='softmax')
     ])
     # Compiling the model
     model.compile(optimizer='adam', loss='categorical_crossentropy',_
      →metrics=['accuracy'])
     # fitting the model
     h = model.fit(X_train, y_train, epochs=300, verbose=1, validation_data=(X_test,_
      →y_test), batch_size=8)
```

```
Epoch 1/300
0.1288 - val_loss: 3.2983 - val_accuracy: 0.2051
Epoch 2/300
0.2189 - val_loss: 2.6972 - val_accuracy: 0.3846
Epoch 3/300
0.3820 - val_loss: 2.3969 - val_accuracy: 0.4359
Epoch 4/300
59/59 [============== ] - Os 3ms/step - loss: 2.2103 - accuracy:
0.4163 - val_loss: 2.0498 - val_accuracy: 0.5128
Epoch 5/300
59/59 [============== ] - Os 3ms/step - loss: 1.9167 - accuracy:
0.4785 - val_loss: 1.7567 - val_accuracy: 0.5556
Epoch 6/300
0.5365 - val_loss: 1.5064 - val_accuracy: 0.5983
Epoch 7/300
0.5880 - val_loss: 1.3573 - val_accuracy: 0.6496
Epoch 8/300
59/59 [============== ] - Os 3ms/step - loss: 1.2390 - accuracy:
0.6288 - val_loss: 1.3341 - val_accuracy: 0.6325
Epoch 9/300
0.6674 - val_loss: 1.0380 - val_accuracy: 0.7265
Epoch 10/300
0.7039 - val_loss: 0.9401 - val_accuracy: 0.7863
Epoch 11/300
0.7725 - val_loss: 0.8784 - val_accuracy: 0.7949
Epoch 12/300
0.7918 - val_loss: 0.9157 - val_accuracy: 0.7863
Epoch 13/300
0.8133 - val_loss: 0.7984 - val_accuracy: 0.8034
Epoch 14/300
59/59 [============== ] - Os 3ms/step - loss: 0.6007 - accuracy:
0.7961 - val_loss: 0.7147 - val_accuracy: 0.8205
Epoch 15/300
0.8004 - val_loss: 0.6583 - val_accuracy: 0.8632
Epoch 16/300
59/59 [============= ] - Os 3ms/step - loss: 0.4361 - accuracy:
```

```
0.8627 - val_loss: 0.7933 - val_accuracy: 0.8120
Epoch 17/300
59/59 [============= ] - Os 3ms/step - loss: 0.4521 - accuracy:
0.8627 - val_loss: 0.6721 - val_accuracy: 0.8718
Epoch 18/300
59/59 [============== ] - Os 3ms/step - loss: 0.3753 - accuracy:
0.8755 - val_loss: 0.5871 - val_accuracy: 0.9145
Epoch 19/300
0.8991 - val_loss: 0.5320 - val_accuracy: 0.9231
Epoch 20/300
0.9206 - val_loss: 0.4754 - val_accuracy: 0.9402
Epoch 21/300
0.9142 - val_loss: 0.5949 - val_accuracy: 0.8803
Epoch 22/300
0.9099 - val_loss: 0.5156 - val_accuracy: 0.9316
Epoch 23/300
0.9313 - val_loss: 0.5772 - val_accuracy: 0.9231
Epoch 24/300
0.9206 - val_loss: 0.6225 - val_accuracy: 0.8974
Epoch 25/300
0.9077 - val_loss: 0.6196 - val_accuracy: 0.8974
Epoch 26/300
0.9120 - val_loss: 0.5494 - val_accuracy: 0.9316
Epoch 27/300
0.8798 - val_loss: 0.5647 - val_accuracy: 0.9060
Epoch 28/300
0.9270 - val_loss: 0.4522 - val_accuracy: 0.9316
Epoch 29/300
59/59 [============== ] - Os 2ms/step - loss: 0.1652 - accuracy:
0.9399 - val_loss: 0.4864 - val_accuracy: 0.9316
Epoch 30/300
0.9142 - val_loss: 0.6544 - val_accuracy: 0.9231
Epoch 31/300
0.9506 - val_loss: 0.4736 - val_accuracy: 0.9402
Epoch 32/300
59/59 [============== ] - Os 2ms/step - loss: 0.1742 - accuracy:
```

```
0.9335 - val_loss: 0.5243 - val_accuracy: 0.8889
Epoch 33/300
59/59 [============== ] - Os 2ms/step - loss: 0.2580 - accuracy:
0.9120 - val_loss: 0.4540 - val_accuracy: 0.9487
Epoch 34/300
59/59 [============== ] - Os 3ms/step - loss: 0.1305 - accuracy:
0.9485 - val_loss: 0.5130 - val_accuracy: 0.9402
Epoch 35/300
0.9657 - val_loss: 0.5079 - val_accuracy: 0.9573
Epoch 36/300
0.9635 - val_loss: 0.5151 - val_accuracy: 0.9316
Epoch 37/300
0.9700 - val_loss: 0.5057 - val_accuracy: 0.9573
Epoch 38/300
0.9807 - val_loss: 0.5822 - val_accuracy: 0.9231
Epoch 39/300
0.9635 - val_loss: 0.5383 - val_accuracy: 0.9487
Epoch 40/300
0.9592 - val_loss: 0.7079 - val_accuracy: 0.9060
Epoch 41/300
0.9592 - val_loss: 0.5337 - val_accuracy: 0.9316
59/59 [============== ] - Os 3ms/step - loss: 0.1117 - accuracy:
0.9700 - val_loss: 0.4902 - val_accuracy: 0.9573
Epoch 43/300
0.9871 - val_loss: 0.5305 - val_accuracy: 0.9487
Epoch 44/300
0.9785 - val loss: 0.5598 - val accuracy: 0.9145
Epoch 45/300
59/59 [============== ] - Os 4ms/step - loss: 0.1889 - accuracy:
0.9485 - val_loss: 0.5869 - val_accuracy: 0.9316
Epoch 46/300
0.9549 - val_loss: 0.5730 - val_accuracy: 0.9145
Epoch 47/300
0.9700 - val_loss: 0.4340 - val_accuracy: 0.9487
Epoch 48/300
59/59 [============== ] - Os 3ms/step - loss: 0.0446 - accuracy:
```

```
0.9871 - val_loss: 0.4636 - val_accuracy: 0.9573
Epoch 49/300
59/59 [============== ] - Os 4ms/step - loss: 0.0366 - accuracy:
0.9871 - val_loss: 0.4811 - val_accuracy: 0.9487
Epoch 50/300
59/59 [============== ] - Os 4ms/step - loss: 0.0210 - accuracy:
0.9957 - val_loss: 0.4861 - val_accuracy: 0.9487
Epoch 51/300
0.9979 - val_loss: 0.4646 - val_accuracy: 0.9658
Epoch 52/300
0.9936 - val_loss: 0.5197 - val_accuracy: 0.9573
Epoch 53/300
0.9957 - val_loss: 0.4859 - val_accuracy: 0.9658
Epoch 54/300
0.9936 - val_loss: 0.5163 - val_accuracy: 0.9487
Epoch 55/300
0.9893 - val_loss: 0.5437 - val_accuracy: 0.9487
Epoch 56/300
0.9957 - val_loss: 0.4782 - val_accuracy: 0.9658
Epoch 57/300
59/59 [============== ] - Os 4ms/step - loss: 0.0259 - accuracy:
0.9936 - val_loss: 0.5027 - val_accuracy: 0.9487
Epoch 58/300
59/59 [============== ] - Os 3ms/step - loss: 0.0206 - accuracy:
0.9957 - val_loss: 0.5084 - val_accuracy: 0.9573
Epoch 59/300
0.9785 - val_loss: 0.7465 - val_accuracy: 0.8803
Epoch 60/300
59/59 [=============== ] - Os 3ms/step - loss: 0.0808 - accuracy:
0.9785 - val loss: 0.5636 - val accuracy: 0.9231
Epoch 61/300
59/59 [============== ] - Os 2ms/step - loss: 0.1387 - accuracy:
0.9421 - val_loss: 0.7919 - val_accuracy: 0.8889
Epoch 62/300
0.9249 - val_loss: 0.5340 - val_accuracy: 0.9060
Epoch 63/300
0.9700 - val_loss: 0.5884 - val_accuracy: 0.9231
Epoch 64/300
59/59 [============== ] - Os 2ms/step - loss: 0.2966 - accuracy:
```

```
0.9313 - val_loss: 0.5691 - val_accuracy: 0.9316
Epoch 65/300
0.9721 - val_loss: 0.6544 - val_accuracy: 0.8974
Epoch 66/300
59/59 [============== ] - Os 4ms/step - loss: 0.0380 - accuracy:
0.9893 - val_loss: 0.4850 - val_accuracy: 0.9402
Epoch 67/300
0.9957 - val_loss: 0.5634 - val_accuracy: 0.9487
Epoch 68/300
0.9936 - val_loss: 0.4615 - val_accuracy: 0.9658
Epoch 69/300
0.9979 - val_loss: 0.4745 - val_accuracy: 0.9658
Epoch 70/300
1.0000 - val_loss: 0.5375 - val_accuracy: 0.9573
Epoch 71/300
1.0000 - val_loss: 0.5168 - val_accuracy: 0.9658
Epoch 72/300
1.0000 - val_loss: 0.5321 - val_accuracy: 0.9658
Epoch 73/300
1.0000 - val_loss: 0.5583 - val_accuracy: 0.9573
1.0000 - val_loss: 0.5616 - val_accuracy: 0.9573
Epoch 75/300
1.0000 - val_loss: 0.5588 - val_accuracy: 0.9487
Epoch 76/300
1.0000 - val_loss: 0.5721 - val_accuracy: 0.9573
Epoch 77/300
59/59 [============== ] - Os 2ms/step - loss: 0.0029 - accuracy:
1.0000 - val_loss: 0.5511 - val_accuracy: 0.9658
Epoch 78/300
1.0000 - val_loss: 0.5609 - val_accuracy: 0.9573
Epoch 79/300
1.0000 - val_loss: 0.5742 - val_accuracy: 0.9487
Epoch 80/300
59/59 [============== ] - Os 2ms/step - loss: 0.0024 - accuracy:
```

```
1.0000 - val_loss: 0.5678 - val_accuracy: 0.9573
Epoch 81/300
1.0000 - val_loss: 0.5799 - val_accuracy: 0.9573
Epoch 82/300
59/59 [============== ] - Os 2ms/step - loss: 0.0020 - accuracy:
1.0000 - val_loss: 0.5784 - val_accuracy: 0.9573
Epoch 83/300
1.0000 - val_loss: 0.5858 - val_accuracy: 0.9487
Epoch 84/300
1.0000 - val_loss: 0.6049 - val_accuracy: 0.9487
Epoch 85/300
1.0000 - val_loss: 0.6101 - val_accuracy: 0.9402
Epoch 86/300
1.0000 - val_loss: 0.5848 - val_accuracy: 0.9487
Epoch 87/300
1.0000 - val_loss: 0.5802 - val_accuracy: 0.9487
Epoch 88/300
1.0000 - val_loss: 0.6003 - val_accuracy: 0.9573
Epoch 89/300
1.0000 - val_loss: 0.6019 - val_accuracy: 0.9573
1.0000 - val_loss: 0.5951 - val_accuracy: 0.9658
Epoch 91/300
1.0000 - val_loss: 0.5959 - val_accuracy: 0.9573
Epoch 92/300
1.0000 - val loss: 0.6064 - val accuracy: 0.9573
Epoch 93/300
59/59 [============= ] - Os 3ms/step - loss: 0.0011 - accuracy:
1.0000 - val_loss: 0.6090 - val_accuracy: 0.9573
Epoch 94/300
1.0000 - val_loss: 0.6321 - val_accuracy: 0.9487
Epoch 95/300
1.0000 - val_loss: 0.6149 - val_accuracy: 0.9487
Epoch 96/300
```

```
accuracy: 1.0000 - val_loss: 0.6139 - val_accuracy: 0.9487
Epoch 97/300
accuracy: 1.0000 - val_loss: 0.6227 - val_accuracy: 0.9573
Epoch 98/300
accuracy: 1.0000 - val_loss: 0.6276 - val_accuracy: 0.9573
Epoch 99/300
59/59 [============ ] - Os 3ms/step - loss: 8.1724e-04 -
accuracy: 1.0000 - val_loss: 0.6295 - val_accuracy: 0.9573
Epoch 100/300
59/59 [============= ] - Os 3ms/step - loss: 7.5687e-04 -
accuracy: 1.0000 - val_loss: 0.6288 - val_accuracy: 0.9573
Epoch 101/300
59/59 [============= ] - Os 3ms/step - loss: 7.4797e-04 -
accuracy: 1.0000 - val_loss: 0.6291 - val_accuracy: 0.9573
Epoch 102/300
accuracy: 1.0000 - val_loss: 0.6216 - val_accuracy: 0.9573
Epoch 103/300
59/59 [============ ] - Os 3ms/step - loss: 7.2240e-04 -
accuracy: 1.0000 - val_loss: 0.6437 - val_accuracy: 0.9402
Epoch 104/300
accuracy: 1.0000 - val_loss: 0.6403 - val_accuracy: 0.9487
Epoch 105/300
59/59 [============= ] - Os 3ms/step - loss: 6.9134e-04 -
accuracy: 1.0000 - val_loss: 0.6358 - val_accuracy: 0.9573
Epoch 106/300
59/59 [============= ] - Os 5ms/step - loss: 6.4818e-04 -
accuracy: 1.0000 - val_loss: 0.6480 - val_accuracy: 0.9487
Epoch 107/300
59/59 [============= ] - Os 3ms/step - loss: 6.0035e-04 -
accuracy: 1.0000 - val_loss: 0.6507 - val_accuracy: 0.9487
Epoch 108/300
59/59 [============= ] - Os 3ms/step - loss: 5.5155e-04 -
accuracy: 1.0000 - val loss: 0.6602 - val accuracy: 0.9487
Epoch 109/300
59/59 [============== ] - Os 3ms/step - loss: 0.0016 - accuracy:
1.0000 - val_loss: 0.6101 - val_accuracy: 0.9487
Epoch 110/300
59/59 [============= ] - Os 3ms/step - loss: 0.2606 - accuracy:
0.9614 - val_loss: 5.0929 - val_accuracy: 0.6923
Epoch 111/300
0.7361 - val_loss: 0.5261 - val_accuracy: 0.8632
Epoch 112/300
59/59 [============= ] - Os 3ms/step - loss: 0.2119 - accuracy:
```

```
0.9335 - val_loss: 0.4882 - val_accuracy: 0.9060
Epoch 113/300
0.9506 - val_loss: 0.3923 - val_accuracy: 0.9231
Epoch 114/300
59/59 [============== ] - Os 3ms/step - loss: 0.0439 - accuracy:
0.9957 - val_loss: 0.3869 - val_accuracy: 0.9573
Epoch 115/300
0.9936 - val_loss: 0.4074 - val_accuracy: 0.9487
Epoch 116/300
0.9979 - val_loss: 0.4007 - val_accuracy: 0.9573
Epoch 117/300
0.9979 - val_loss: 0.4143 - val_accuracy: 0.9573
Epoch 118/300
1.0000 - val_loss: 0.4112 - val_accuracy: 0.9658
Epoch 119/300
1.0000 - val_loss: 0.4429 - val_accuracy: 0.9402
Epoch 120/300
0.9979 - val_loss: 0.4387 - val_accuracy: 0.9487
Epoch 121/300
0.9979 - val_loss: 0.4431 - val_accuracy: 0.9487
Epoch 122/300
1.0000 - val_loss: 0.4338 - val_accuracy: 0.9487
Epoch 123/300
1.0000 - val_loss: 0.4569 - val_accuracy: 0.9487
Epoch 124/300
59/59 [=============== ] - Os 3ms/step - loss: 0.0047 - accuracy:
1.0000 - val loss: 0.4568 - val accuracy: 0.9487
Epoch 125/300
59/59 [============== ] - Os 3ms/step - loss: 0.0053 - accuracy:
1.0000 - val_loss: 0.4549 - val_accuracy: 0.9487
Epoch 126/300
1.0000 - val_loss: 0.4811 - val_accuracy: 0.9487
Epoch 127/300
1.0000 - val_loss: 0.4913 - val_accuracy: 0.9487
Epoch 128/300
59/59 [============== ] - Os 3ms/step - loss: 0.0031 - accuracy:
```

```
1.0000 - val_loss: 0.5162 - val_accuracy: 0.9487
Epoch 129/300
1.0000 - val_loss: 0.5041 - val_accuracy: 0.9573
Epoch 130/300
59/59 [============== ] - Os 2ms/step - loss: 0.0029 - accuracy:
1.0000 - val_loss: 0.5165 - val_accuracy: 0.9487
Epoch 131/300
59/59 [============== ] - Os 3ms/step - loss: 0.0024 - accuracy:
1.0000 - val_loss: 0.5167 - val_accuracy: 0.9487
Epoch 132/300
1.0000 - val_loss: 0.5312 - val_accuracy: 0.9573
Epoch 133/300
1.0000 - val_loss: 0.5339 - val_accuracy: 0.9573
Epoch 134/300
1.0000 - val_loss: 0.5412 - val_accuracy: 0.9487
Epoch 135/300
1.0000 - val_loss: 0.5352 - val_accuracy: 0.9487
Epoch 136/300
1.0000 - val_loss: 0.5492 - val_accuracy: 0.9402
Epoch 137/300
1.0000 - val_loss: 0.5590 - val_accuracy: 0.9402
Epoch 138/300
1.0000 - val_loss: 0.5585 - val_accuracy: 0.9402
Epoch 139/300
1.0000 - val_loss: 0.5592 - val_accuracy: 0.9402
Epoch 140/300
59/59 [=============== ] - Os 2ms/step - loss: 0.0012 - accuracy:
1.0000 - val loss: 0.5558 - val accuracy: 0.9487
Epoch 141/300
59/59 [============== ] - Os 3ms/step - loss: 0.0012 - accuracy:
1.0000 - val_loss: 0.5788 - val_accuracy: 0.9487
Epoch 142/300
1.0000 - val_loss: 0.5722 - val_accuracy: 0.9402
Epoch 143/300
1.0000 - val_loss: 0.5716 - val_accuracy: 0.9402
Epoch 144/300
59/59 [========== ] - Os 3ms/step - loss: 9.7921e-04 -
```

```
accuracy: 1.0000 - val_loss: 0.5765 - val_accuracy: 0.9573
Epoch 145/300
59/59 [============= ] - Os 3ms/step - loss: 0.0010 - accuracy:
1.0000 - val_loss: 0.5874 - val_accuracy: 0.9487
Epoch 146/300
59/59 [============= ] - Os 4ms/step - loss: 8.9194e-04 -
accuracy: 1.0000 - val_loss: 0.5915 - val_accuracy: 0.9402
Epoch 147/300
59/59 [============ ] - Os 3ms/step - loss: 9.0571e-04 -
accuracy: 1.0000 - val_loss: 0.5771 - val_accuracy: 0.9487
Epoch 148/300
59/59 [============= ] - Os 3ms/step - loss: 7.8802e-04 -
accuracy: 1.0000 - val_loss: 0.5919 - val_accuracy: 0.9402
Epoch 149/300
59/59 [============= ] - Os 3ms/step - loss: 7.6733e-04 -
accuracy: 1.0000 - val_loss: 0.5971 - val_accuracy: 0.9402
Epoch 150/300
accuracy: 1.0000 - val_loss: 0.5865 - val_accuracy: 0.9487
Epoch 151/300
59/59 [============ ] - Os 4ms/step - loss: 7.2566e-04 -
accuracy: 1.0000 - val_loss: 0.6053 - val_accuracy: 0.9487
Epoch 152/300
59/59 [============ ] - Os 4ms/step - loss: 6.6622e-04 -
accuracy: 1.0000 - val_loss: 0.6088 - val_accuracy: 0.9402
Epoch 153/300
59/59 [============== ] - Os 3ms/step - loss: 6.2435e-04 -
accuracy: 1.0000 - val_loss: 0.6038 - val_accuracy: 0.9402
Epoch 154/300
accuracy: 1.0000 - val_loss: 0.6011 - val_accuracy: 0.9402
Epoch 155/300
59/59 [============= ] - Os 3ms/step - loss: 5.7622e-04 -
accuracy: 1.0000 - val_loss: 0.6026 - val_accuracy: 0.9402
Epoch 156/300
59/59 [============ ] - Os 2ms/step - loss: 5.4689e-04 -
accuracy: 1.0000 - val loss: 0.6157 - val accuracy: 0.9402
Epoch 157/300
accuracy: 1.0000 - val_loss: 0.6118 - val_accuracy: 0.9402
Epoch 158/300
59/59 [============= ] - Os 3ms/step - loss: 4.9661e-04 -
accuracy: 1.0000 - val_loss: 0.6251 - val_accuracy: 0.9402
Epoch 159/300
59/59 [============ ] - Os 3ms/step - loss: 4.6517e-04 -
accuracy: 1.0000 - val_loss: 0.6189 - val_accuracy: 0.9402
Epoch 160/300
59/59 [========== ] - Os 3ms/step - loss: 4.8997e-04 -
```

```
accuracy: 1.0000 - val_loss: 0.6340 - val_accuracy: 0.9402
Epoch 161/300
accuracy: 1.0000 - val_loss: 0.6270 - val_accuracy: 0.9402
Epoch 162/300
accuracy: 1.0000 - val_loss: 0.6316 - val_accuracy: 0.9402
Epoch 163/300
59/59 [============= ] - Os 5ms/step - loss: 3.9423e-04 -
accuracy: 1.0000 - val_loss: 0.6336 - val_accuracy: 0.9402
Epoch 164/300
59/59 [============= ] - Os 2ms/step - loss: 3.9202e-04 -
accuracy: 1.0000 - val_loss: 0.6348 - val_accuracy: 0.9402
Epoch 165/300
59/59 [============== ] - Os 4ms/step - loss: 3.5065e-04 -
accuracy: 1.0000 - val_loss: 0.6405 - val_accuracy: 0.9402
Epoch 166/300
accuracy: 1.0000 - val_loss: 0.6348 - val_accuracy: 0.9402
Epoch 167/300
59/59 [============ ] - Os 3ms/step - loss: 3.2651e-04 -
accuracy: 1.0000 - val_loss: 0.6378 - val_accuracy: 0.9402
Epoch 168/300
59/59 [============= ] - Os 4ms/step - loss: 3.2025e-04 -
accuracy: 1.0000 - val_loss: 0.6610 - val_accuracy: 0.9402
Epoch 169/300
accuracy: 1.0000 - val_loss: 0.6420 - val_accuracy: 0.9402
Epoch 170/300
accuracy: 1.0000 - val_loss: 0.6563 - val_accuracy: 0.9402
Epoch 171/300
accuracy: 1.0000 - val_loss: 0.6508 - val_accuracy: 0.9402
Epoch 172/300
59/59 [============ ] - Os 3ms/step - loss: 2.6683e-04 -
accuracy: 1.0000 - val loss: 0.6577 - val accuracy: 0.9402
Epoch 173/300
accuracy: 1.0000 - val_loss: 0.6550 - val_accuracy: 0.9402
Epoch 174/300
accuracy: 1.0000 - val_loss: 0.6652 - val_accuracy: 0.9402
Epoch 175/300
59/59 [============ ] - Os 2ms/step - loss: 2.3381e-04 -
accuracy: 1.0000 - val_loss: 0.6610 - val_accuracy: 0.9402
Epoch 176/300
59/59 [========== ] - Os 3ms/step - loss: 2.2649e-04 -
```

```
accuracy: 1.0000 - val_loss: 0.6649 - val_accuracy: 0.9402
Epoch 177/300
accuracy: 1.0000 - val_loss: 0.6682 - val_accuracy: 0.9402
Epoch 178/300
accuracy: 1.0000 - val_loss: 0.6710 - val_accuracy: 0.9402
Epoch 179/300
accuracy: 1.0000 - val_loss: 0.6747 - val_accuracy: 0.9402
Epoch 180/300
59/59 [============= ] - Os 3ms/step - loss: 1.8727e-04 -
accuracy: 1.0000 - val_loss: 0.6788 - val_accuracy: 0.9402
Epoch 181/300
accuracy: 1.0000 - val_loss: 0.6728 - val_accuracy: 0.9402
Epoch 182/300
accuracy: 1.0000 - val_loss: 0.6849 - val_accuracy: 0.9402
Epoch 183/300
accuracy: 1.0000 - val_loss: 0.6791 - val_accuracy: 0.9402
Epoch 184/300
59/59 [============= ] - Os 3ms/step - loss: 1.6976e-04 -
accuracy: 1.0000 - val_loss: 0.6858 - val_accuracy: 0.9402
Epoch 185/300
accuracy: 1.0000 - val_loss: 0.6863 - val_accuracy: 0.9402
Epoch 186/300
accuracy: 1.0000 - val_loss: 0.6952 - val_accuracy: 0.9402
Epoch 187/300
accuracy: 1.0000 - val_loss: 0.6854 - val_accuracy: 0.9402
Epoch 188/300
59/59 [============ ] - Os 4ms/step - loss: 1.4558e-04 -
accuracy: 1.0000 - val loss: 0.6992 - val accuracy: 0.9402
Epoch 189/300
accuracy: 1.0000 - val_loss: 0.6892 - val_accuracy: 0.9402
Epoch 190/300
accuracy: 1.0000 - val_loss: 0.6868 - val_accuracy: 0.9402
Epoch 191/300
59/59 [============ ] - Os 4ms/step - loss: 1.2741e-04 -
accuracy: 1.0000 - val_loss: 0.7118 - val_accuracy: 0.9402
Epoch 192/300
59/59 [========== ] - Os 4ms/step - loss: 1.2091e-04 -
```

```
accuracy: 1.0000 - val_loss: 0.6945 - val_accuracy: 0.9402
Epoch 193/300
accuracy: 1.0000 - val_loss: 0.7119 - val_accuracy: 0.9402
Epoch 194/300
accuracy: 1.0000 - val_loss: 0.7010 - val_accuracy: 0.9402
Epoch 195/300
59/59 [============= ] - Os 4ms/step - loss: 1.0558e-04 -
accuracy: 1.0000 - val_loss: 0.7159 - val_accuracy: 0.9402
Epoch 196/300
59/59 [============= ] - Os 4ms/step - loss: 1.0023e-04 -
accuracy: 1.0000 - val_loss: 0.7137 - val_accuracy: 0.9402
Epoch 197/300
59/59 [============= ] - Os 3ms/step - loss: 9.5609e-05 -
accuracy: 1.0000 - val_loss: 0.7138 - val_accuracy: 0.9402
Epoch 198/300
accuracy: 1.0000 - val_loss: 0.7114 - val_accuracy: 0.9402
Epoch 199/300
59/59 [============ ] - Os 4ms/step - loss: 9.1386e-05 -
accuracy: 1.0000 - val_loss: 0.7200 - val_accuracy: 0.9402
Epoch 200/300
accuracy: 1.0000 - val_loss: 0.7185 - val_accuracy: 0.9402
Epoch 201/300
accuracy: 1.0000 - val_loss: 0.7208 - val_accuracy: 0.9402
59/59 [============ ] - Os 4ms/step - loss: 8.8762e-05 -
accuracy: 1.0000 - val_loss: 0.7153 - val_accuracy: 0.9402
Epoch 203/300
accuracy: 1.0000 - val_loss: 0.7331 - val_accuracy: 0.9402
Epoch 204/300
59/59 [============= ] - Os 4ms/step - loss: 7.7397e-05 -
accuracy: 1.0000 - val loss: 0.7315 - val accuracy: 0.9402
Epoch 205/300
accuracy: 1.0000 - val_loss: 0.7298 - val_accuracy: 0.9402
Epoch 206/300
accuracy: 1.0000 - val_loss: 0.7386 - val_accuracy: 0.9487
Epoch 207/300
59/59 [============ ] - Os 4ms/step - loss: 7.0942e-05 -
accuracy: 1.0000 - val_loss: 0.7391 - val_accuracy: 0.9487
Epoch 208/300
59/59 [=========== ] - Os 5ms/step - loss: 7.0117e-05 -
```

```
accuracy: 1.0000 - val_loss: 0.7372 - val_accuracy: 0.9487
Epoch 209/300
accuracy: 1.0000 - val_loss: 0.7456 - val_accuracy: 0.9402
Epoch 210/300
accuracy: 1.0000 - val_loss: 0.7457 - val_accuracy: 0.9402
Epoch 211/300
59/59 [============= ] - Os 4ms/step - loss: 6.0883e-05 -
accuracy: 1.0000 - val_loss: 0.7416 - val_accuracy: 0.9487
Epoch 212/300
59/59 [============= ] - Os 3ms/step - loss: 5.4056e-05 -
accuracy: 1.0000 - val_loss: 0.7459 - val_accuracy: 0.9487
Epoch 213/300
accuracy: 1.0000 - val_loss: 0.7521 - val_accuracy: 0.9487
Epoch 214/300
accuracy: 1.0000 - val_loss: 0.7404 - val_accuracy: 0.9487
Epoch 215/300
accuracy: 1.0000 - val_loss: 0.7553 - val_accuracy: 0.9487
Epoch 216/300
accuracy: 1.0000 - val_loss: 0.7451 - val_accuracy: 0.9402
Epoch 217/300
accuracy: 1.0000 - val_loss: 0.7630 - val_accuracy: 0.9487
accuracy: 1.0000 - val_loss: 0.7625 - val_accuracy: 0.9487
Epoch 219/300
accuracy: 1.0000 - val_loss: 0.7517 - val_accuracy: 0.9487
Epoch 220/300
59/59 [============= ] - Os 2ms/step - loss: 4.3609e-05 -
accuracy: 1.0000 - val loss: 0.7674 - val accuracy: 0.9487
Epoch 221/300
accuracy: 1.0000 - val_loss: 0.7739 - val_accuracy: 0.9487
Epoch 222/300
59/59 [============= ] - Os 3ms/step - loss: 4.0557e-05 -
accuracy: 1.0000 - val_loss: 0.7641 - val_accuracy: 0.9487
Epoch 223/300
59/59 [============= ] - Os 3ms/step - loss: 4.1311e-05 -
accuracy: 1.0000 - val_loss: 0.7792 - val_accuracy: 0.9487
Epoch 224/300
59/59 [========== ] - Os 4ms/step - loss: 3.5765e-05 -
```

```
accuracy: 1.0000 - val_loss: 0.7724 - val_accuracy: 0.9487
Epoch 225/300
accuracy: 1.0000 - val_loss: 0.7791 - val_accuracy: 0.9487
Epoch 226/300
accuracy: 1.0000 - val_loss: 0.7847 - val_accuracy: 0.9487
Epoch 227/300
59/59 [============= ] - Os 3ms/step - loss: 3.4415e-05 -
accuracy: 1.0000 - val_loss: 0.7788 - val_accuracy: 0.9487
Epoch 228/300
59/59 [============= ] - Os 3ms/step - loss: 3.1673e-05 -
accuracy: 1.0000 - val_loss: 0.7886 - val_accuracy: 0.9487
Epoch 229/300
59/59 [============== ] - Os 3ms/step - loss: 3.4159e-05 -
accuracy: 1.0000 - val_loss: 0.7952 - val_accuracy: 0.9487
Epoch 230/300
accuracy: 1.0000 - val_loss: 0.7791 - val_accuracy: 0.9487
Epoch 231/300
accuracy: 1.0000 - val_loss: 0.7870 - val_accuracy: 0.9487
Epoch 232/300
accuracy: 1.0000 - val_loss: 0.7895 - val_accuracy: 0.9487
Epoch 233/300
accuracy: 1.0000 - val_loss: 0.7868 - val_accuracy: 0.9487
59/59 [============ ] - Os 4ms/step - loss: 2.5134e-05 -
accuracy: 1.0000 - val_loss: 0.7902 - val_accuracy: 0.9487
Epoch 235/300
accuracy: 1.0000 - val_loss: 0.7978 - val_accuracy: 0.9487
Epoch 236/300
59/59 [============= ] - Os 3ms/step - loss: 2.3049e-05 -
accuracy: 1.0000 - val loss: 0.7940 - val accuracy: 0.9487
Epoch 237/300
accuracy: 1.0000 - val_loss: 0.7952 - val_accuracy: 0.9487
Epoch 238/300
accuracy: 1.0000 - val_loss: 0.7997 - val_accuracy: 0.9487
Epoch 239/300
59/59 [============ ] - Os 3ms/step - loss: 2.1572e-05 -
accuracy: 1.0000 - val_loss: 0.7968 - val_accuracy: 0.9487
Epoch 240/300
59/59 [========== ] - Os 3ms/step - loss: 2.0466e-05 -
```

```
accuracy: 1.0000 - val_loss: 0.8083 - val_accuracy: 0.9487
Epoch 241/300
accuracy: 1.0000 - val_loss: 0.8072 - val_accuracy: 0.9487
Epoch 242/300
accuracy: 1.0000 - val_loss: 0.8155 - val_accuracy: 0.9487
Epoch 243/300
59/59 [============= ] - Os 3ms/step - loss: 1.8557e-05 -
accuracy: 1.0000 - val_loss: 0.8072 - val_accuracy: 0.9487
Epoch 244/300
59/59 [============= ] - Os 3ms/step - loss: 1.7848e-05 -
accuracy: 1.0000 - val_loss: 0.8175 - val_accuracy: 0.9487
Epoch 245/300
59/59 [============= ] - Os 3ms/step - loss: 1.7412e-05 -
accuracy: 1.0000 - val_loss: 0.8194 - val_accuracy: 0.9487
Epoch 246/300
accuracy: 1.0000 - val_loss: 0.8189 - val_accuracy: 0.9487
Epoch 247/300
accuracy: 1.0000 - val_loss: 0.8178 - val_accuracy: 0.9487
Epoch 248/300
accuracy: 1.0000 - val_loss: 0.8193 - val_accuracy: 0.9487
Epoch 249/300
accuracy: 1.0000 - val_loss: 0.8308 - val_accuracy: 0.9487
Epoch 250/300
59/59 [============ ] - Os 5ms/step - loss: 1.4523e-05 -
accuracy: 1.0000 - val_loss: 0.8170 - val_accuracy: 0.9487
Epoch 251/300
accuracy: 1.0000 - val_loss: 0.8319 - val_accuracy: 0.9487
Epoch 252/300
59/59 [============ ] - Os 4ms/step - loss: 1.3789e-05 -
accuracy: 1.0000 - val loss: 0.8236 - val accuracy: 0.9487
Epoch 253/300
accuracy: 1.0000 - val_loss: 0.8312 - val_accuracy: 0.9487
Epoch 254/300
59/59 [============= ] - Os 3ms/step - loss: 1.2720e-05 -
accuracy: 1.0000 - val_loss: 0.8396 - val_accuracy: 0.9487
Epoch 255/300
59/59 [============= ] - Os 3ms/step - loss: 1.2348e-05 -
accuracy: 1.0000 - val_loss: 0.8318 - val_accuracy: 0.9487
Epoch 256/300
59/59 [=========== ] - Os 3ms/step - loss: 1.2505e-05 -
```

```
accuracy: 1.0000 - val_loss: 0.8421 - val_accuracy: 0.9487
Epoch 257/300
accuracy: 1.0000 - val_loss: 0.8455 - val_accuracy: 0.9487
Epoch 258/300
accuracy: 1.0000 - val_loss: 0.8436 - val_accuracy: 0.9487
Epoch 259/300
59/59 [============= ] - Os 2ms/step - loss: 1.0756e-05 -
accuracy: 1.0000 - val_loss: 0.8441 - val_accuracy: 0.9487
Epoch 260/300
59/59 [============= ] - Os 3ms/step - loss: 1.0923e-05 -
accuracy: 1.0000 - val_loss: 0.8478 - val_accuracy: 0.9487
Epoch 261/300
59/59 [============= ] - Os 3ms/step - loss: 9.8826e-06 -
accuracy: 1.0000 - val_loss: 0.8479 - val_accuracy: 0.9487
Epoch 262/300
59/59 [============= ] - Os 3ms/step - loss: 9.6836e-06 -
accuracy: 1.0000 - val_loss: 0.8484 - val_accuracy: 0.9487
Epoch 263/300
accuracy: 1.0000 - val_loss: 0.8470 - val_accuracy: 0.9487
Epoch 264/300
accuracy: 1.0000 - val_loss: 0.8495 - val_accuracy: 0.9487
Epoch 265/300
accuracy: 1.0000 - val_loss: 0.8689 - val_accuracy: 0.9487
Epoch 266/300
accuracy: 1.0000 - val_loss: 0.8588 - val_accuracy: 0.9487
Epoch 267/300
accuracy: 1.0000 - val_loss: 0.8616 - val_accuracy: 0.9487
Epoch 268/300
59/59 [============ ] - Os 4ms/step - loss: 7.7837e-06 -
accuracy: 1.0000 - val loss: 0.8617 - val accuracy: 0.9487
Epoch 269/300
accuracy: 1.0000 - val_loss: 0.8654 - val_accuracy: 0.9487
Epoch 270/300
59/59 [============== ] - Os 6ms/step - loss: 7.0212e-06 -
accuracy: 1.0000 - val_loss: 0.8656 - val_accuracy: 0.9487
Epoch 271/300
59/59 [============ ] - Os 4ms/step - loss: 7.0086e-06 -
accuracy: 1.0000 - val_loss: 0.8627 - val_accuracy: 0.9487
Epoch 272/300
59/59 [========== ] - Os 2ms/step - loss: 6.7167e-06 -
```

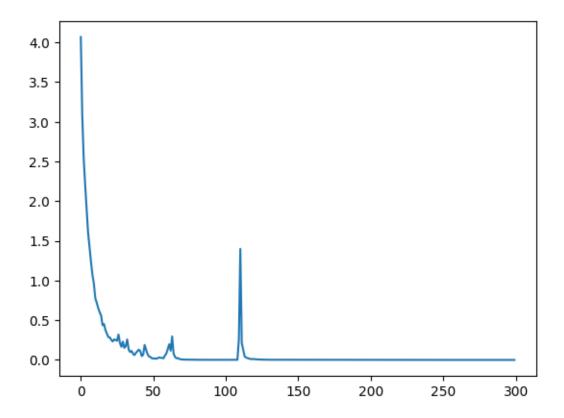
```
accuracy: 1.0000 - val_loss: 0.8690 - val_accuracy: 0.9487
Epoch 273/300
accuracy: 1.0000 - val_loss: 0.8707 - val_accuracy: 0.9487
Epoch 274/300
accuracy: 1.0000 - val_loss: 0.8705 - val_accuracy: 0.9487
Epoch 275/300
59/59 [============ ] - Os 3ms/step - loss: 6.1511e-06 -
accuracy: 1.0000 - val_loss: 0.8679 - val_accuracy: 0.9487
Epoch 276/300
accuracy: 1.0000 - val_loss: 0.8673 - val_accuracy: 0.9487
Epoch 277/300
accuracy: 1.0000 - val_loss: 0.8757 - val_accuracy: 0.9487
Epoch 278/300
accuracy: 1.0000 - val_loss: 0.8683 - val_accuracy: 0.9487
Epoch 279/300
accuracy: 1.0000 - val_loss: 0.8779 - val_accuracy: 0.9487
Epoch 280/300
accuracy: 1.0000 - val_loss: 0.8888 - val_accuracy: 0.9487
Epoch 281/300
accuracy: 1.0000 - val_loss: 0.8785 - val_accuracy: 0.9487
accuracy: 1.0000 - val_loss: 0.8954 - val_accuracy: 0.9487
Epoch 283/300
59/59 [============= ] - Os 3ms/step - loss: 4.7192e-06 -
accuracy: 1.0000 - val_loss: 0.8910 - val_accuracy: 0.9487
Epoch 284/300
59/59 [============ ] - Os 3ms/step - loss: 4.4680e-06 -
accuracy: 1.0000 - val loss: 0.8865 - val accuracy: 0.9487
Epoch 285/300
accuracy: 1.0000 - val_loss: 0.8908 - val_accuracy: 0.9487
Epoch 286/300
59/59 [============= ] - Os 3ms/step - loss: 4.0748e-06 -
accuracy: 1.0000 - val_loss: 0.8883 - val_accuracy: 0.9487
Epoch 287/300
59/59 [============ ] - Os 4ms/step - loss: 3.9970e-06 -
accuracy: 1.0000 - val_loss: 0.9022 - val_accuracy: 0.9487
Epoch 288/300
```

```
Epoch 289/300
  accuracy: 1.0000 - val_loss: 0.9022 - val_accuracy: 0.9573
  Epoch 290/300
  accuracy: 1.0000 - val loss: 0.8995 - val accuracy: 0.9487
  Epoch 291/300
  59/59 [============= ] - Os 3ms/step - loss: 3.4540e-06 -
  accuracy: 1.0000 - val_loss: 0.9011 - val_accuracy: 0.9487
  Epoch 292/300
  59/59 [============= ] - Os 3ms/step - loss: 3.8034e-06 -
  accuracy: 1.0000 - val_loss: 0.8888 - val_accuracy: 0.9487
  Epoch 293/300
  59/59 [============== ] - Os 4ms/step - loss: 3.6778e-06 -
  accuracy: 1.0000 - val_loss: 0.9016 - val_accuracy: 0.9487
  Epoch 294/300
  accuracy: 1.0000 - val_loss: 0.9014 - val_accuracy: 0.9487
  Epoch 295/300
  59/59 [============ ] - Os 3ms/step - loss: 3.0718e-06 -
  accuracy: 1.0000 - val_loss: 0.9035 - val_accuracy: 0.9487
  Epoch 296/300
  accuracy: 1.0000 - val_loss: 0.9143 - val_accuracy: 0.9487
  Epoch 297/300
  accuracy: 1.0000 - val_loss: 0.9043 - val_accuracy: 0.9487
  accuracy: 1.0000 - val_loss: 0.9167 - val_accuracy: 0.9487
  Epoch 299/300
  accuracy: 1.0000 - val_loss: 0.9112 - val_accuracy: 0.9487
  Epoch 300/300
  accuracy: 1.0000 - val loss: 0.9179 - val accuracy: 0.9487
[]: model.summary()
  Model: "sequential_69"
   Layer (type)
                    Output Shape
                                     Param #
  ______
   conv1d_109 (Conv1D)
                     (None, 14, 32)
                                     128
   Layer (type)
                    Output Shape
                                    Param #
```

accuracy: 1.0000 - val\_loss: 0.8904 - val\_accuracy: 0.9487

```
conv1d_109 (Conv1D)
                             (None, 14, 32)
                                                   128
    dense_286 (Dense)
                             (None, 14, 32)
                                                   1056
    max_pooling1d_77 (MaxPooli (None, 7, 32)
    ng1D)
    conv1d_110 (Conv1D)
                             (None, 5, 32)
                                                   3104
    dense_287 (Dense)
                             (None, 5, 32)
                                                   1056
                             (None, 160)
    flatten_71 (Flatten)
    dense_288 (Dense)
                             (None, 32)
                                                   5152
    dense_289 (Dense)
                             (None, 89)
                                                   2937
   Total params: 13433 (52.47 KB)
   Trainable params: 13433 (52.47 KB)
   Non-trainable params: 0 (0.00 Byte)
   ______
   Evaluate the model
[]: score=model.evaluate(X_test, y_test)
    print("Test loss:",score[0])
    print("Test accuracy",score[1])
   1/4 [=====>...] - ETA: Os - loss: 0.2138 - accuracy: 0.9375
   0.9487
   Test loss: 0.9179329872131348
   Test accuracy 0.9487179517745972
   Test loss: 0.9179329872131348
   Test accuracy 0.9487179517745972
   Plot the model performance
[]: plt.plot(h.history['loss'])
```

# []: [<matplotlib.lines.Line2D at 0x2d2d028a310>]



#### Make Prediction

```
[ ]: result = model.predict(X_test[:7])

1/1 [========] - 0s 180ms/step
1/1 [======] - 0s 180ms/step
Compare Prediction against test value
[ ]: result[:7].argmax(axis=1)
[ ]: array([66, 80, 38, 73, 65, 66, 38], dtype=int64)
[ ]: y_test[:7].argmax(axis=1)
[ ]: array([66, 73, 38, 73, 65, 66, 38], dtype=int64)
[ ]: result
[ ]: array([[1.89866194e-29, 0.00000000e+00, 0.00000000e+00, 0.00000000e+00, 5.69825823e-35, 7.81600583e-28, 1.93848942e-29, 4.45189004e-31, 1.10324277e-35, 8.11767999e-31, 1.61683253e-32, 1.23624127e-30, 0.00000000e+00, 1.05047862e-36, 0.00000000e+00, 1.10588053e-35, 1.89874048e-30, 8.48238836e-34, 8.78467429e-31, 2.46712809e-37,
```

```
1.21756200e-31, 1.08824546e-35, 5.97217307e-37, 0.00000000e+00,
1.45048533e-26, 1.11941866e-30, 0.00000000e+00, 0.00000000e+00,
2.39625993e-32, 8.38856701e-24, 2.27189861e-36, 4.21633749e-21,
6.61542093e-34, 0.00000000e+00, 1.15919161e-35, 1.38811175e-35,
8.99209226e-19, 0.00000000e+00, 6.37342662e-20, 3.31173346e-32,
1.82945727e-32, 0.00000000e+00, 0.0000000e+00, 1.99557907e-23,
0.0000000e+00, 1.08655881e-17, 0.0000000e+00, 0.0000000e+00,
0.00000000e+00, 4.34141092e-20, 0.00000000e+00, 0.00000000e+00,
1.14390535e-19, 0.00000000e+00, 3.09835960e-36, 4.71294194e-17,
6.65969422e-30, 0.00000000e+00, 4.85339397e-16, 0.00000000e+00,
1.06850360e-14, 1.33017991e-22, 1.17412535e-23, 4.38577143e-35,
4.33652809e-35, 3.22795399e-02, 9.67719734e-01, 1.55260436e-25,
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