

## final (2)

May 28, 2023

```
[2]: # Import libraries and magics

import numpy as np
import pandas as pd
from scipy import stats
from pandas import DatetimeIndex as dt
from seaborn import pairplot
import matplotlib.pyplot as plt
%matplotlib inline
plt.style.use('bmh')
import pandas as pd
import seaborn as sns
import numpy as np
import os
import matplotlib.pyplot as plt
%matplotlib inline
from IPython.display import display
pd.options.display.max_columns = None
import warnings
warnings.filterwarnings("ignore")
from statsmodels.stats.outliers_influence import variance_inflation_factor
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report, confusion_matrix

from sklearn.model_selection import train_test_split, GridSearchCV,
↳RandomizedSearchCV, cross_val_score
from sklearn.preprocessing import StandardScaler, OneHotEncoder,
↳OrdinalEncoder, PolynomialFeatures
from sklearn.pipeline import Pipeline
from sklearn.compose import ColumnTransformer
from sklearn.impute import SimpleImputer
```

```
import joblib
from sklearn.linear_model import LinearRegression, Lasso, LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import r2_score, accuracy_score, confusion_matrix, \
    classification_report
```

[2]:

```
[3]: # Going through the data
data = pd.read_csv('starcraft_player_data.csv')
data
```

```
[3]:
```

	GameID	LeagueIndex	Age	HoursPerWeek	TotalHours	APM	\
0	52	5	27	10	3000	143.7180	
1	55	5	23	10	5000	129.2322	
2	56	4	30	10	200	69.9612	
3	57	3	19	20	400	107.6016	
4	58	3	32	10	500	122.8908	
...	...	...	...	...	...	...	
3390	10089	8	?	?	?	259.6296	
3391	10090	8	?	?	?	314.6700	
3392	10092	8	?	?	?	299.4282	
3393	10094	8	?	?	?	375.8664	
3394	10095	8	?	?	?	348.3576	

	SelectByHotkeys	AssignToHotkeys	UniqueHotkeys	MinimapAttacks	\
0	0.003515	0.000220	7	0.000110	
1	0.003304	0.000259	4	0.000294	
2	0.001101	0.000336	4	0.000294	
3	0.001034	0.000213	1	0.000053	
4	0.001136	0.000327	2	0.000000	
...	...	...	...	...	
3390	0.020425	0.000743	9	0.000621	
3391	0.028043	0.001157	10	0.000246	
3392	0.028341	0.000860	7	0.000338	
3393	0.036436	0.000594	5	0.000204	
3394	0.029855	0.000811	4	0.000224	

	MinimapRightClicks	NumberOfPACs	GapBetweenPACs	ActionLatency	\
0	0.000392	0.004849	32.6677	40.8673	
1	0.000432	0.004307	32.9194	42.3454	
2	0.000461	0.002926	44.6475	75.3548	
3	0.000543	0.003783	29.2203	53.7352	
4	0.001329	0.002368	22.6885	62.0813	
...	...	...	...	...	
3390	0.000146	0.004555	18.6059	42.8342	

3391	0.001083	0.004259	14.3023	36.1156
3392	0.000169	0.004439	12.4028	39.5156
3393	0.000780	0.004346	11.6910	34.8547
3394	0.001315	0.005566	20.0537	33.5142

	ActionsInPAC	TotalMapExplored	WorkersMade	UniqueUnitsMade	\
0	4.7508	28	0.001397		6
1	4.8434	22	0.001193		5
2	4.0430	22	0.000745		6
3	4.9155	19	0.000426		7
4	9.3740	15	0.001174		4
...	...	...	...	...	
3390	6.2754	46	0.000877		5
3391	7.1965	16	0.000788		4
3392	6.3979	19	0.001260		4
3393	7.9615	15	0.000613		6
3394	6.3719	27	0.001566		7

	ComplexUnitsMade	ComplexAbilitiesUsed
0	0.000000	0.000000
1	0.000000	0.000208
2	0.000000	0.000189
3	0.000000	0.000384
4	0.000000	0.000019
...	...	...
3390	0.000000	0.000000
3391	0.000000	0.000000
3392	0.000000	0.000000
3393	0.000000	0.000631
3394	0.000457	0.000895

[3395 rows x 20 columns]

```
[4]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3395 entries, 0 to 3394
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   GameID                 3395 non-null   int64
1   LeagueIndex            3395 non-null   int64
2   Age                    3395 non-null   object
3   HoursPerWeek           3395 non-null   object
4   TotalHours              3395 non-null   object
5   APM                    3395 non-null   float64
6   SelectByHotkeys        3395 non-null   float64
```

```

7   AssignToHotkeys      3395 non-null float64
8   UniqueHotkeys       3395 non-null int64
9   MinimapAttacks      3395 non-null float64
10  MinimapRightClicks  3395 non-null float64
11  NumberOfPACs        3395 non-null float64
12  GapBetweenPACs      3395 non-null float64
13  ActionLatency        3395 non-null float64
14  ActionsInPAC        3395 non-null float64
15  TotalMapExplored     3395 non-null int64
16  WorkersMade          3395 non-null float64
17  UniqueUnitsMade      3395 non-null int64
18  ComplexUnitsMade     3395 non-null float64
19  ComplexAbilitiesUsed 3395 non-null float64
dtypes: float64(12), int64(5), object(3)
memory usage: 530.6+ KB

```

```
[5]: data.isnull().sum()
```

```

[5]: GameID          0
     LeagueIndex     0
     Age             0
     HoursPerWeek    0
     TotalHours      0
     APM             0
     SelectByHotkeys 0
     AssignToHotkeys 0
     UniqueHotkeys   0
     MinimapAttacks  0
     MinimapRightClicks 0
     NumberOfPACs    0
     GapBetweenPACs  0
     ActionLatency   0
     ActionsInPAC    0
     TotalMapExplored 0
     WorkersMade     0
     UniqueUnitsMade 0
     ComplexUnitsMade 0
     ComplexAbilitiesUsed 0
dtype: int64

```

0.1 We convert these 3 columns to numeric datatypes from object and find out that there are null values that are present. We use errors='coerce' since there are '?' values present in these 3 features. We convert it to null from '?'

```
[6]: data['Age'] = pd.to_numeric(data['Age'], errors='coerce')
data['HoursPerWeek'] = pd.to_numeric(data['HoursPerWeek'], errors='coerce')
data['TotalHours'] = pd.to_numeric(data['TotalHours'], errors='coerce')
```

```
[7]: data.isnull().sum()
```

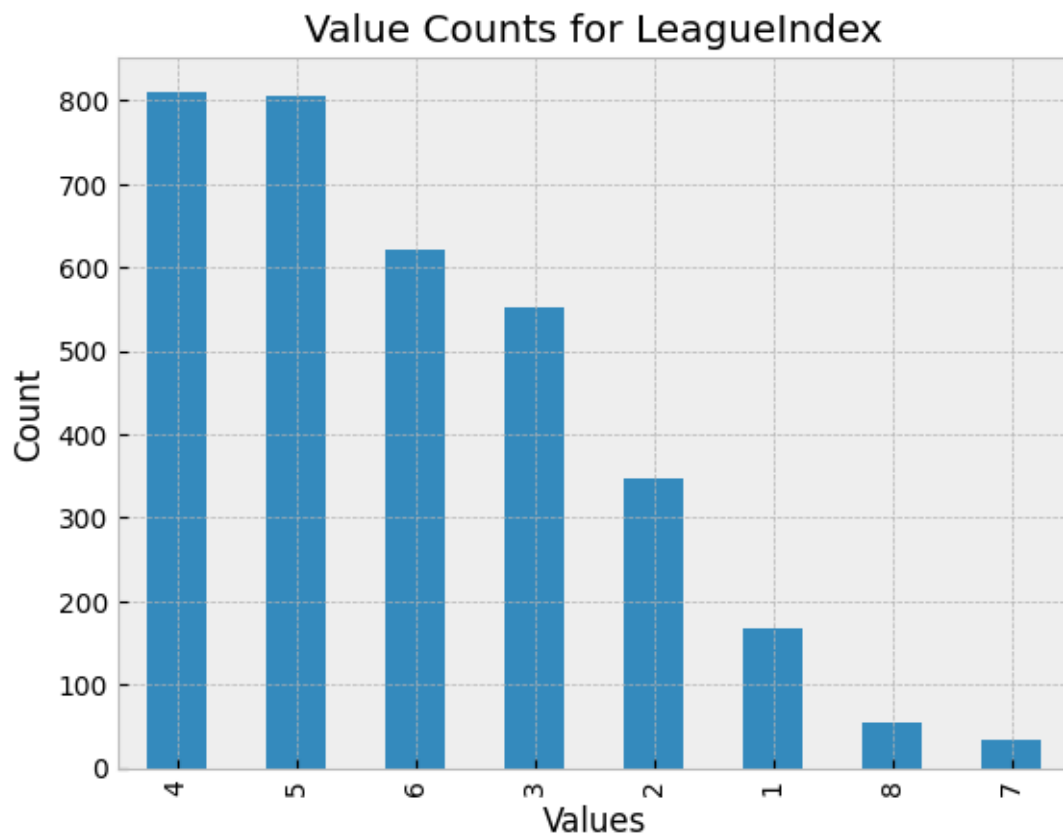
```
[7]: GameID                0
LeagueIndex             0
Age                    55
HoursPerWeek           56
TotalHours             57
APM                    0
SelectByHotkeys        0
AssignToHotkeys        0
UniqueHotkeys          0
MinimapAttacks         0
MinimapRightClicks     0
NumberOfPACs           0
GapBetweenPACs         0
ActionLatency          0
ActionsInPAC           0
TotalMapExplored       0
WorkersMade            0
UniqueUnitsMade        0
ComplexUnitsMade       0
ComplexAbilitiesUsed   0
dtype: int64
```

0.2 Conclusions we can form from this data:

1. We can see that for LeagueIndex we don't have many instances in the range 7-8.
2. The null values in those 3 features are about 1% of the entire dataset and we can use some imputing techniques for it.
3. We can come up with individual techniques to fill in the null values so that noise is not added to our dataset.

```
[8]: interesting_columns=[ 'LeagueIndex']
for column in interesting_columns:
    plt.figure() # Create a new figure for each column
    data[column].value_counts().plot(kind='bar')
    plt.title(f'Value Counts for {column}')
```

```
plt.xlabel('Values')
plt.ylabel('Count')
plt.show()
```



0.3 For age, we see that the LeagueIndex 8 has all missing values. Furthermore, we also see that as LeagueIndex increases the the distribution of the age decreases even though the mean remains relatively same. We take the median of LeagueIndex 7 to impute the null values.

```
[9]: import matplotlib.pyplot as plt

# Create a range of LeagueIndex values from 1 to 8
league_index_range = range(1, 9)

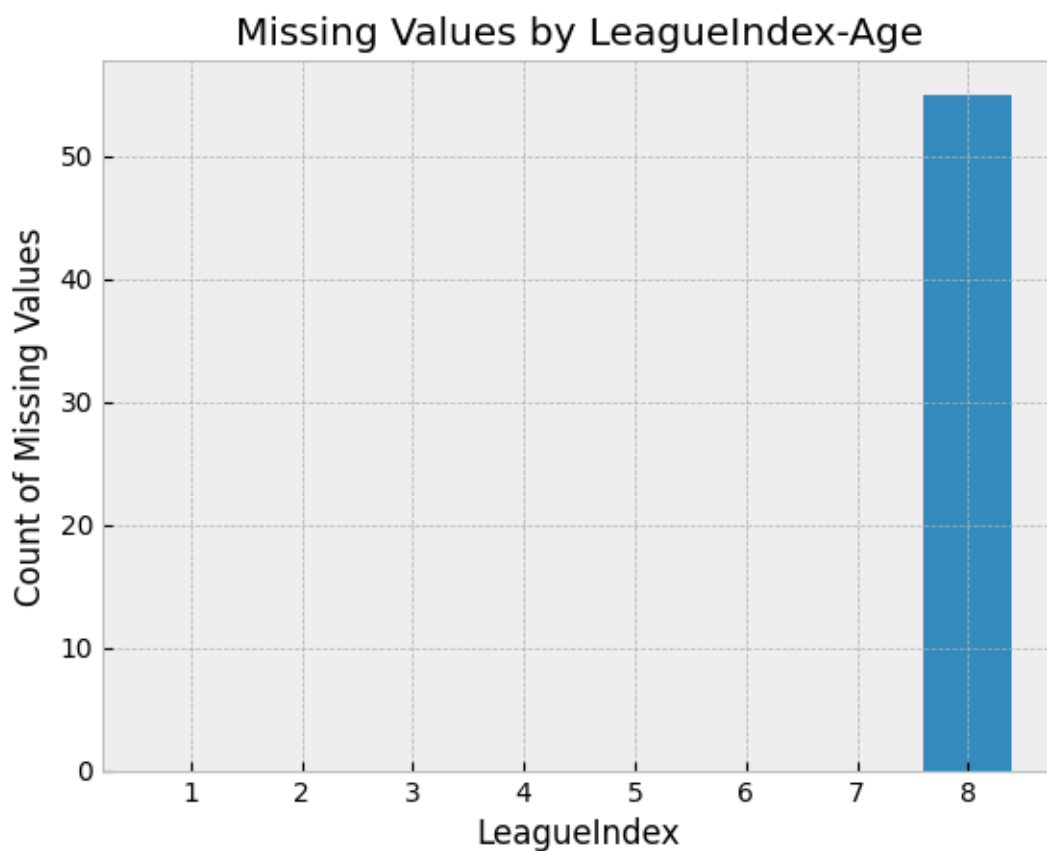
# Initialize a list to store the count of missing values for each LeagueIndex
# category
missing_values_count = []
```

```

# Iterate over each LeagueIndex category
for index in league_index_range:
    count = data[(data["LeagueIndex"] == index) & (data["Age"].isnull())].
    ↪shape[0]
    missing_values_count.append(count)

# Plot the bar chart
plt.bar(league_index_range, missing_values_count)
plt.xlabel("LeagueIndex")
plt.ylabel("Count of Missing Values")
plt.title("Missing Values by LeagueIndex-Age")
plt.show()

```



```

[10]: import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(15, 10))
sns.set_style("whitegrid") # Set the style of the plot

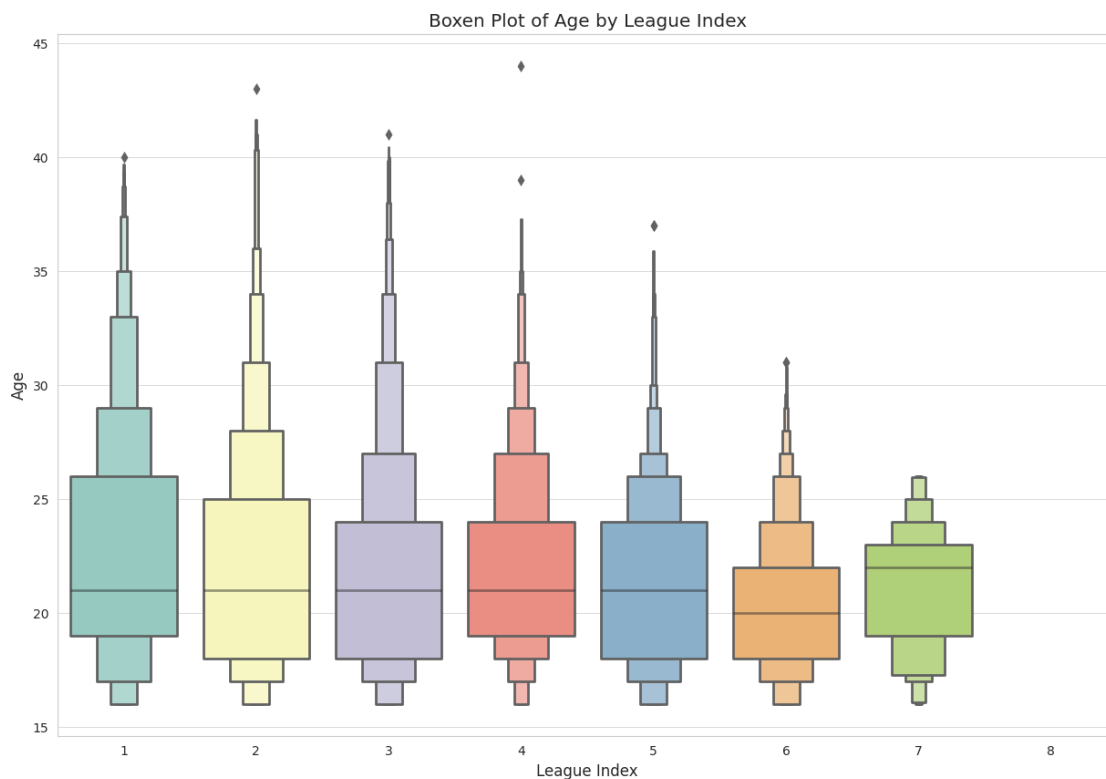
```

```

sns.boxenplot(data=data, x="LeagueIndex", y="Age", palette="Set3",
               k_depth="proportion")
plt.xlabel("League Index")
plt.ylabel("Age")
plt.title("Boxen Plot of Age by League Index")

plt.show()

```



```

[11]: med_val = data[data["LeagueIndex"] == 7]["Age"].median()
      data["Age"] = data["Age"].fillna(med_val)

```

0.4 For HoursPerWeek we see that most null values are present in leagueIndex 8 and one null value is present in LeagueIndex 5. The hours per week spent increases as the LeagueIndex increases which makes logical sense. Since there is quite a difference in the mean of all leagueIndex, we fill with individual median LeagueIndex values.

```

[12]: import matplotlib.pyplot as plt

      # Create a range of LeagueIndex values from 1 to 8
      league_index_range = range(1, 9)

```



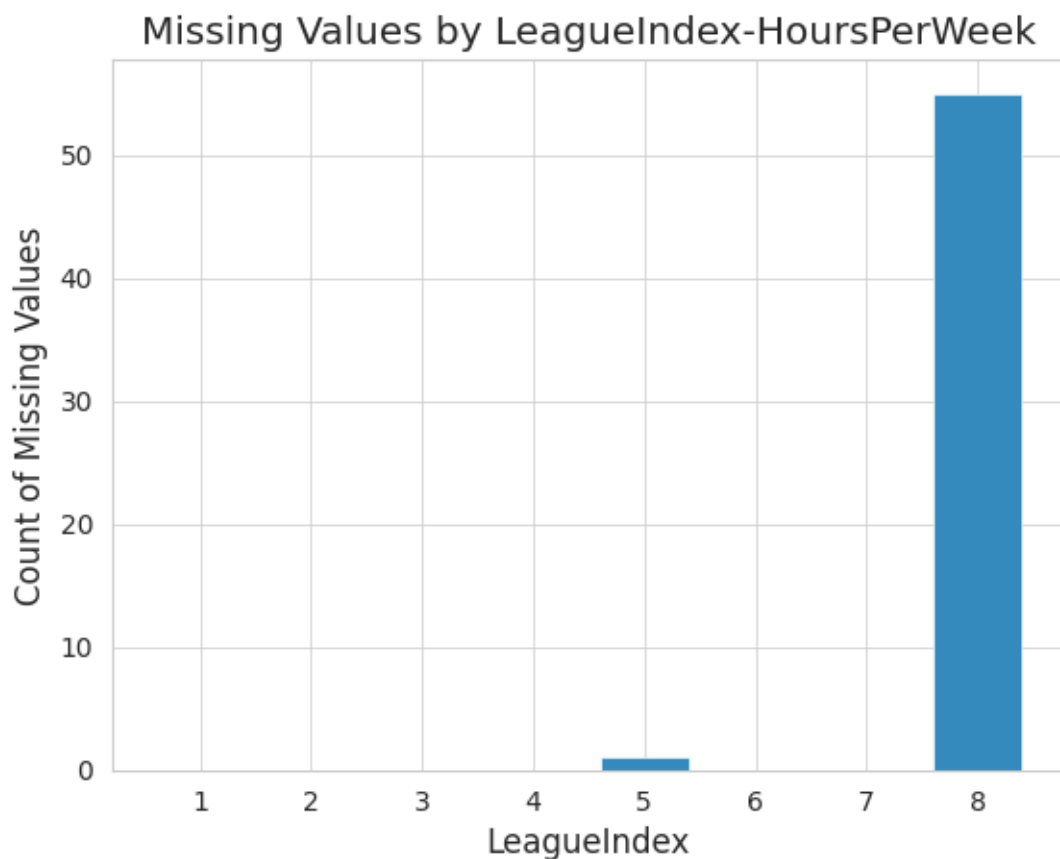
```

# Initialize a list to store the count of missing values for each LeagueIndex
↳ category
missing_values_count = []

# Iterate over each LeagueIndex category
for index in league_index_range:
    count = data[(data["LeagueIndex"] == index) & (data["HoursPerWeek"].
↳ isnull())].shape[0]
    missing_values_count.append(count)

# Plot the bar chart
plt.bar(league_index_range, missing_values_count)
plt.xlabel("LeagueIndex")
plt.ylabel("Count of Missing Values")
plt.title("Missing Values by LeagueIndex-HoursPerWeek")
plt.show()

```

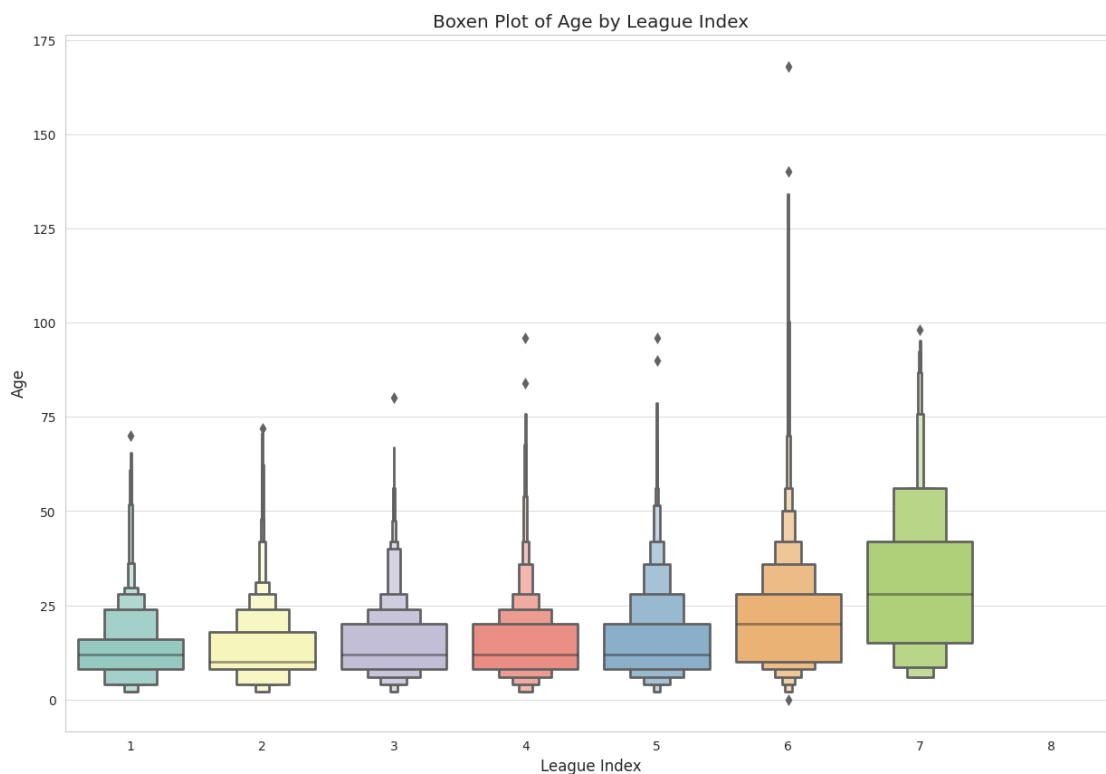


```
[13]: import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(15, 10))
sns.set_style("whitegrid") # Set the style of the plot

sns.boxenplot(data=data, x="LeagueIndex", y="HoursPerWeek", palette="Set3",
              ↳k_depth="proportion")
plt.xlabel("League Index")
plt.ylabel("Age")
plt.title("Boxen Plot of Age by League Index")

plt.show()
```



```
[14]: med_val_15 = data[data["LeagueIndex"] == 5]["HoursPerWeek"].median()
data.loc[
    (data["HoursPerWeek"].isnull()) & (data["LeagueIndex"] == 5), "HoursPerWeek"
] = med_val_15

# Now imputing the missing value for player with missing hoursperweek for
↳leagueindex = 8
med_val_18 = data[data["LeagueIndex"] == 7]["HoursPerWeek"].median()
```

```
data.loc[
    (data["HoursPerWeek"].isnull()) & (data["LeagueIndex"] == 8), "HoursPerWeek"
] = med_val_18
```

- 0.5 Logically the total hours increases as the LeagueIndex increases, since the more you play the better you are. Since the box plot is on the log normal scale, there is quite a big difference between the mean total hours per leagueindex. Yet again, most of the missing values are from LeagueIndex 8 and a few from index 5. We fill these with the median values from the individual Total Hours per leagueIndex.

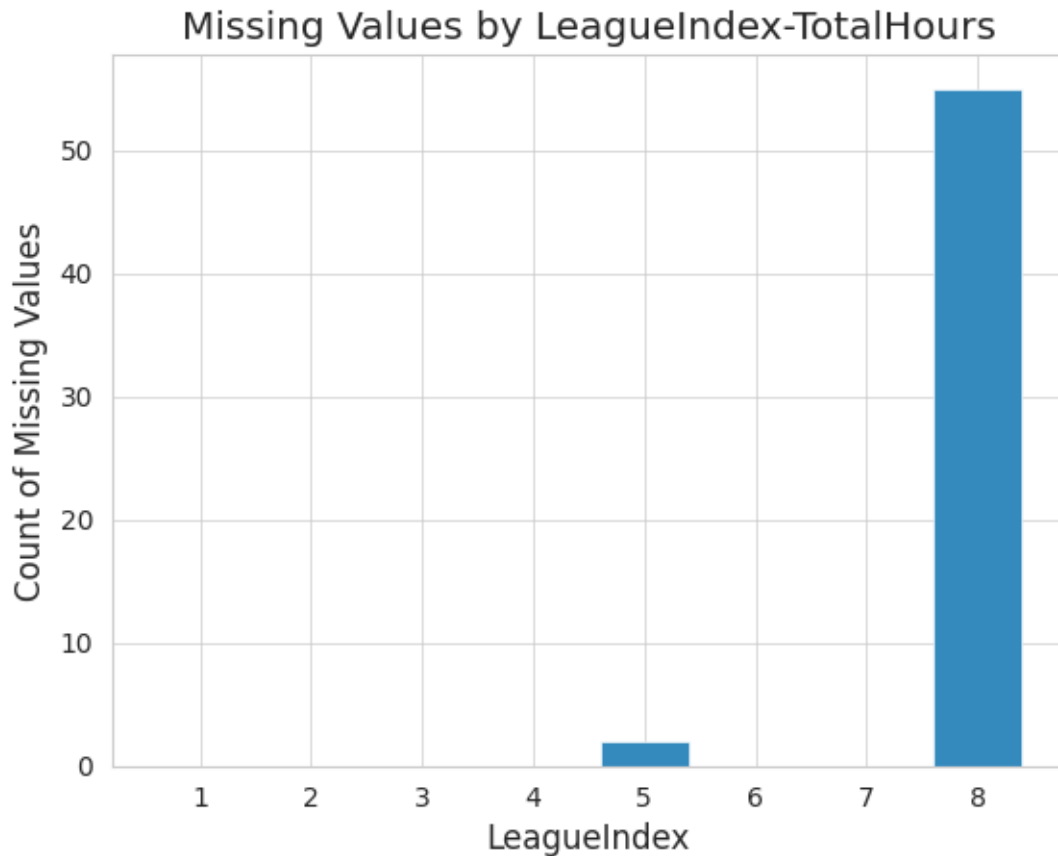
```
[15]: import matplotlib.pyplot as plt

# Create a range of LeagueIndex values from 1 to 8
league_index_range = range(1, 9)

# Initialize a list to store the count of missing values for each LeagueIndex
# category
missing_values_count = []

# Iterate over each LeagueIndex category
for index in league_index_range:
    count = data[(data["LeagueIndex"] == index) & (data["TotalHours"].
# isnull())].shape[0]
    missing_values_count.append(count)

# Plot the bar chart
plt.bar(league_index_range, missing_values_count)
plt.xlabel("LeagueIndex")
plt.ylabel("Count of Missing Values")
plt.title("Missing Values by LeagueIndex-TotalHours")
plt.show()
```



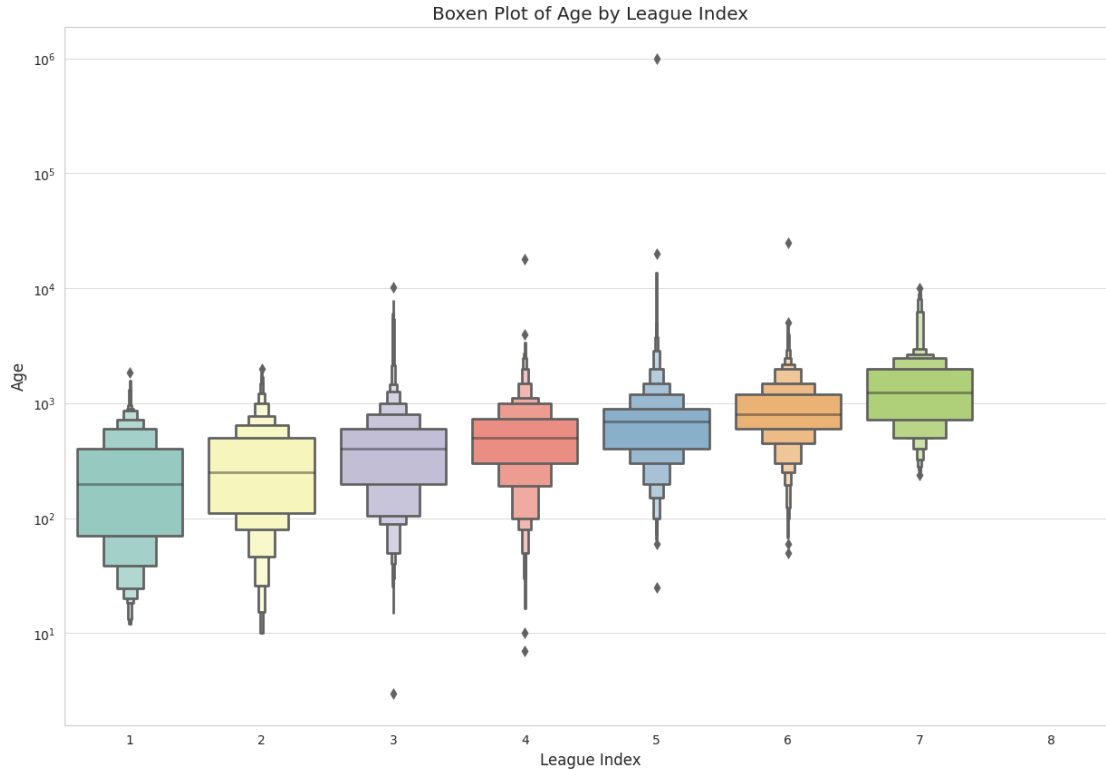
```
[16]: import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(15, 10))
sns.set_style("whitegrid") # Set the style of the plot

sns.boxenplot(data=data, x="LeagueIndex", y="TotalHours", palette="Set3",
              k_depth="proportion")
plt.xlabel("League Index")
plt.ylabel("Age")
plt.title("Boxen Plot of Age by League Index")

plt.yscale("log") # Change the scale of the y-axis to logarithmic

plt.show()
```



```
[17]: med_val_15 = data[data["LeagueIndex"] == 5]["TotalHours"].median()
data.loc[
    (data["TotalHours"].isnull()) & (data["LeagueIndex"] == 5), "TotalHours"
] = med_val_15

# Now imputing the missing value for player with missing hoursperweek for
↳ leagueindex = 8
med_val_18 = data[data["LeagueIndex"] == 7]["TotalHours"].median()
data.loc[
    (data["TotalHours"].isnull()) & (data["LeagueIndex"] == 8), "TotalHours"
] = med_val_18
```

```
[18]: data.describe()
```

```
[18]:
```

	GameID	LeagueIndex	Age	HoursPerWeek	TotalHours \
count	3395.000000	3395.000000	3395.000000	3395.000000	3395.000000
mean	4805.012371	4.184094	21.653608	16.105449	964.959647
std	2719.944851	1.517327	4.172356	11.961811	17172.134959
min	52.000000	1.000000	16.000000	0.000000	3.000000
25%	2464.500000	3.000000	19.000000	8.000000	300.000000
50%	4874.000000	4.000000	21.000000	12.000000	500.000000
75%	7108.500000	5.000000	24.000000	20.000000	800.000000

max	10095.000000	8.000000	44.000000	168.000000	1000000.000000
-----	--------------	----------	-----------	------------	----------------

	APM	SelectByHotkeys	AssignToHotkeys	UniqueHotkeys	\
count	3395.000000	3395.000000	3395.000000	3395.000000	
mean	117.046947	0.004299	0.000374	4.364654	
std	51.945291	0.005284	0.000225	2.360333	
min	22.059600	0.000000	0.000000	0.000000	
25%	79.900200	0.001258	0.000204	3.000000	
50%	108.010200	0.002500	0.000353	4.000000	
75%	142.790400	0.005133	0.000499	6.000000	
max	389.831400	0.043088	0.001752	10.000000	

	MinimapAttacks	MinimapRightClicks	NumberOfPACs	GapBetweenPACs	\
count	3395.000000	3395.000000	3395.000000	3395.000000	
mean	0.000098	0.000387	0.003463	40.361562	
std	0.000166	0.000377	0.000992	17.153570	
min	0.000000	0.000000	0.000679	6.666700	
25%	0.000000	0.000140	0.002754	28.957750	
50%	0.000040	0.000281	0.003395	36.723500	
75%	0.000119	0.000514	0.004027	48.290500	
max	0.003019	0.004041	0.007971	237.142900	

	ActionLatency	ActionsInPAC	TotalMapExplored	WorkersMade	\
count	3395.000000	3395.000000	3395.000000	3395.000000	
mean	63.739403	5.272988	22.131664	0.001032	
std	19.238869	1.494835	7.431719	0.000519	
min	24.093600	2.038900	5.000000	0.000077	
25%	50.446600	4.272850	17.000000	0.000683	
50%	60.931800	5.095500	22.000000	0.000905	
75%	73.681300	6.033600	27.000000	0.001259	
max	176.372100	18.558100	58.000000	0.005149	

	UniqueUnitsMade	ComplexUnitsMade	ComplexAbilitiesUsed
count	3395.000000	3395.000000	3395.000000
mean	6.534021	0.000059	0.000142
std	1.857697	0.000111	0.000265
min	2.000000	0.000000	0.000000
25%	5.000000	0.000000	0.000000
50%	6.000000	0.000000	0.000020
75%	8.000000	0.000086	0.000181
max	13.000000	0.000902	0.003084

```
[19]: # Since it is not possible to play the entire week
data.loc[data["HoursPerWeek"] > 120, "HoursPerWeek"] = 120
```

```
[20]: data.describe()
```

[20] :

	GameID	LeagueIndex	Age	HoursPerWeek	TotalHours \
count	3395.000000	3395.000000	3395.000000	3395.000000	3395.000000
mean	4805.012371	4.184094	21.653608	16.085420	964.959647
std	2719.944851	1.517327	4.172356	11.752646	17172.134959
min	52.000000	1.000000	16.000000	0.000000	3.000000
25%	2464.500000	3.000000	19.000000	8.000000	300.000000
50%	4874.000000	4.000000	21.000000	12.000000	500.000000
75%	7108.500000	5.000000	24.000000	20.000000	800.000000
max	10095.000000	8.000000	44.000000	120.000000	1000000.000000

	APM	SelectByHotkeys	AssignToHotkeys	UniqueHotkeys \
count	3395.000000	3395.000000	3395.000000	3395.000000
mean	117.046947	0.004299	0.000374	4.364654
std	51.945291	0.005284	0.000225	2.360333
min	22.059600	0.000000	0.000000	0.000000
25%	79.900200	0.001258	0.000204	3.000000
50%	108.010200	0.002500	0.000353	4.000000
75%	142.790400	0.005133	0.000499	6.000000
max	389.831400	0.043088	0.001752	10.000000

	MinimapAttacks	MinimapRightClicks	NumberOfPACs	GapBetweenPACs \
count	3395.000000	3395.000000	3395.000000	3395.000000
mean	0.000098	0.000387	0.003463	40.361562
std	0.000166	0.000377	0.000992	17.153570
min	0.000000	0.000000	0.000679	6.666700
25%	0.000000	0.000140	0.002754	28.957750
50%	0.000040	0.000281	0.003395	36.723500
75%	0.000119	0.000514	0.004027	48.290500
max	0.003019	0.004041	0.007971	237.142900

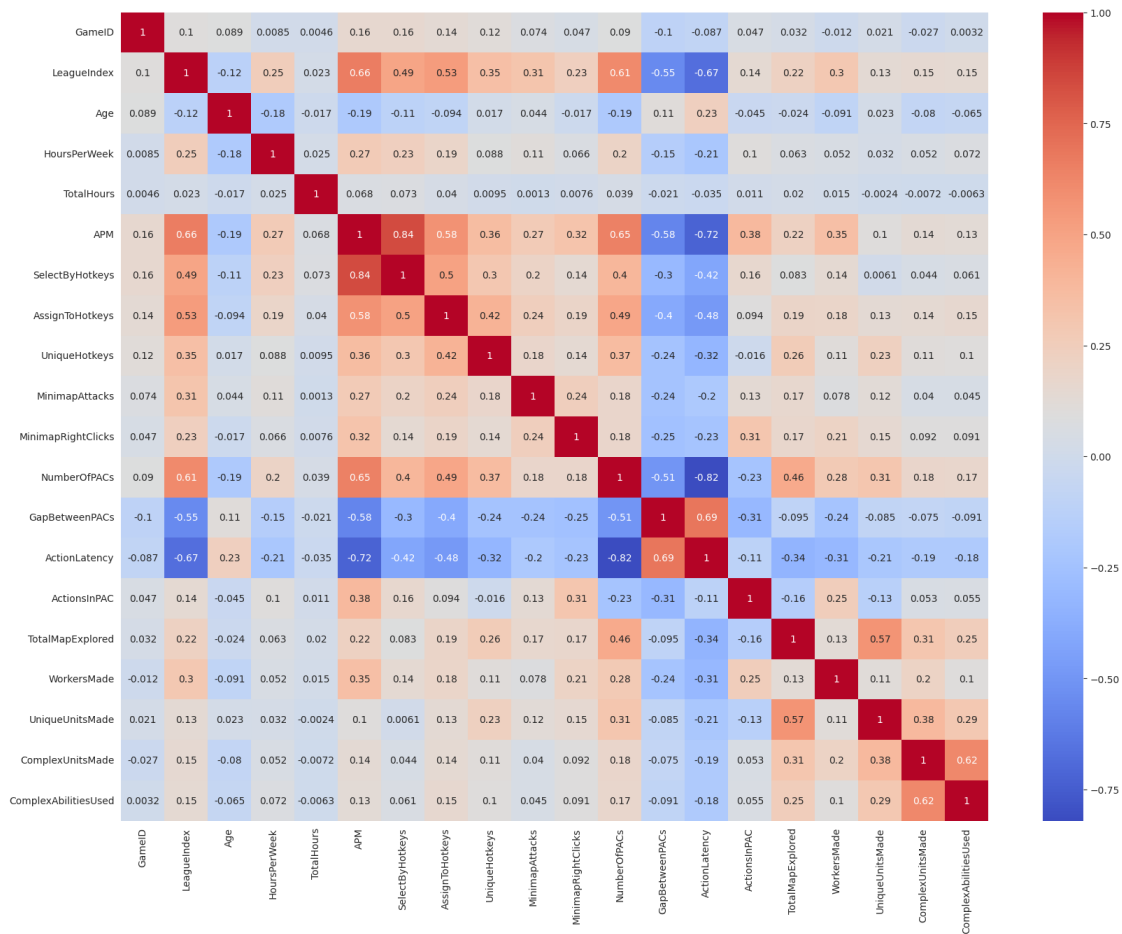
	ActionLatency	ActionsInPAC	TotalMapExplored	WorkersMade \
count	3395.000000	3395.000000	3395.000000	3395.000000
mean	63.739403	5.272988	22.131664	0.001032
std	19.238869	1.494835	7.431719	0.000519
min	24.093600	2.038900	5.000000	0.000077
25%	50.446600	4.272850	17.000000	0.000683
50%	60.931800	5.095500	22.000000	0.000905
75%	73.681300	6.033600	27.000000	0.001259
max	176.372100	18.558100	58.000000	0.005149

	UniqueUnitsMade	ComplexUnitsMade	ComplexAbilitiesUsed
count	3395.000000	3395.000000	3395.000000
mean	6.534021	0.000059	0.000142
std	1.857697	0.000111	0.000265
min	2.000000	0.000000	0.000000
25%	5.000000	0.000000	0.000000
50%	6.000000	0.000000	0.000020

75%	8.000000	0.000086	0.000181
max	13.000000	0.000902	0.003084

0.6 We can see that there is high correlation between LeagueIndex and predictor variables, as is indicated by the heatmap where there are multiple values greater than 0.5 and lesser than -0.5. We see that multicollinearity is present in the data which would indicate that we can drop some features since some features would be redundant.

```
[21]: plt.figure(figsize=(20,15))
sns.heatmap(data.corr(), annot=True, cmap='coolwarm')
plt.show()
```





**0.7** We are removing the GameID and LeagueIndex features. GameID offers no valuable information intuitively and the correlation is also low with LeagueIndex. We can try to use techniques like Variance inflation factor to measure multicollienarity in the features.

Interpretation of VIF:

VIF=1 There is no multicollinearity; the predictor variable is not correlated with other predictors.

VIF > 1 and < 5: There is moderate multicollinearity, indicating a moderate correlation with other predictors.

VIF >= 5: There is a high degree of multicollinearity, indicating a strong correlation with other predictors.

```
[22]: data_re = data.drop(["LeagueIndex", "GameID"], axis=1)
```

```
[23]: vif_data = pd.DataFrame()
vif_data["feature"] = data_re.columns

# calculating VIF for each feature
vif_data["VIF"] = [
    variance_inflation_factor(data_re.values, i)
    for i in range(len(data_re.columns))
]
```

```
[24]: vif_data.sort_values(["VIF"], ascending=False)
```

```
[24]:
```

	feature	VIF
3	APM	232.141090
9	NumberOfPACs	87.009324
12	ActionsInPAC	68.410297
11	ActionLatency	40.244447
0	Age	29.483299
4	SelectByHotkeys	23.235798
15	UniqueUnitsMade	22.060698
13	TotalMapExplored	18.184561
10	GapBetweenPACs	14.802954
5	AssignToHotkeys	6.491549
14	WorkersMade	6.445501
6	UniqueHotkeys	5.897251
1	HoursPerWeek	3.202471
8	MinimapRightClicks	2.636975
16	ComplexUnitsMade	2.323360
17	ComplexAbilitiesUsed	2.134550
7	MinimapAttacks	1.579499
2	TotalHours	1.010096

Next we also remove some features where the correlation is low with the target variable and the

VIF factor is high. In this case “ActionsInPAC”, “UniqueUnitsMade”

```
[25]: gamedf_indpt = data.drop(
        ["LeagueIndex", "ActionsInPAC", "UniqueUnitsMade"], axis=1
    )
    vif_data = pd.DataFrame()
    vif_data["feature"] = gamedf_indpt.columns

    # calculating VIF for each feature
    vif_data["VIF"] = [
        variance_inflation_factor(gamedf_indpt.values, i)
        for i in range(len(gamedf_indpt.columns))
    ]
    vif_data.sort_values(["VIF"], ascending=False)
```

```
[25]:
```

	feature	VIF
4	APM	51.618980
10	NumberOfPACs	34.791021
1	Age	28.122009
12	ActionLatency	26.916903
13	TotalMapExplored	14.854583
11	GapBetweenPACs	14.741981
5	SelectByHotkeys	8.119519
6	AssignToHotkeys	6.481163
14	WorkersMade	6.460107
7	UniqueHotkeys	5.857128
0	GameID	4.340584
2	HoursPerWeek	3.166164
9	MinimapRightClicks	2.620920
15	ComplexUnitsMade	2.238089
16	ComplexAbilitiesUsed	2.112293
8	MinimapAttacks	1.573936
3	TotalHours	1.009968

We also remove APM since it is highly correlated with ActionLatency which is further highly correlated with LeagueIndex

```
[26]: gamedf_indpt = data.drop(
        ["LeagueIndex", "ActionsInPAC", "UniqueUnitsMade", "APM"], axis=1
    )
    vif_data = pd.DataFrame()
    vif_data["feature"] = gamedf_indpt.columns

    # calculating VIF for each feature
    vif_data["VIF"] = [
        variance_inflation_factor(gamedf_indpt.values, i)
        for i in range(len(gamedf_indpt.columns))
    ]
```

```
vif_data.sort_values(["VIF"], ascending=False)
```

```
[26]:
```

	feature	VIF
1	Age	27.691425
11	ActionLatency	26.866244
9	NumberOfPACs	23.400289
12	TotalMapExplored	14.853234
10	GapBetweenPACs	13.937734
5	AssignToHotkeys	6.461673
6	UniqueHotkeys	5.856953
13	WorkersMade	5.568088
0	GameID	4.308569
2	HoursPerWeek	3.098915
4	SelectByHotkeys	2.507283
8	MinimapRightClicks	2.372744
14	ComplexUnitsMade	2.237685
15	ComplexAbilitiesUsed	2.112131
7	MinimapAttacks	1.570528
3	TotalHours	1.009875

```
[27]: tmp=data  
data=data.drop(["ActionsInPAC", "UniqueUnitsMade", "APM"], axis=1)
```

## 0.8 Steps before feeding the data into the model”

1. We split our dataset in train and test dividing it into a 80:20 split. We do a stratified split since we want to divide the classes in a proportionate manner between the train and test dataset.
2. Furthermore we use a pipeline to standardize our dataset. We fit it on the training set and just transform it on the testing dataset.
3. I use F1 macro score since the dataset is imbalanced, and it is a multi class classification along with the fact that all classes are of equal importance.

```
[76]: X = data.drop(columns=['LeagueIndex'])  
y = data['LeagueIndex']  
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,  
↪stratify=y, random_state=0)  
  
X_train.shape, y_train.shape, X_test.shape, y_test.shape
```

```
[76]: ((2716, 16), (2716,), (679, 16), (679,))
```

```
[77]: num_2scale_attribs = list(X_train.columns)
```

```
# Pipeline for numerical attributes to be scaled
num_2scale_pipeline = Pipeline([('std_scaler', StandardScaler())])

train_set_prepared = num_2scale_pipeline.fit_transform(X_train)
```

```
[78]: train_set_prepared.shape
```

```
[78]: (2716, 16)
```

```
[79]: test_set_prepared = num_2scale_pipeline.transform(X_test)
```

```
[80]: df_train_set_prepared = pd.DataFrame(data=train_set_prepared,
      ↪ columns=num_2scale_attribs)

df_train_set_prepared
```

```
[80]:
```

	GameID	Age	HoursPerWeek	TotalHours	SelectByHotkeys	\
0	1.318432	-0.156888	-0.010148	-0.033614	-0.647269	
1	1.155848	-0.873436	1.010672	-0.002345	0.186175	
2	0.366676	-0.634587	1.010672	-0.012768	1.516507	
3	-1.013281	-1.351135	-0.860831	-0.019543	-0.710278	
4	1.599758	1.276208	-0.690695	-0.017980	-0.082602	
...	...	...	...	...	...	
2711	0.227109	0.081962	-0.860831	-0.033614	-0.071696	
2712	-0.788586	-0.156888	0.330125	-0.012768	-0.118653	
2713	-0.349060	-1.351135	-0.180285	-0.036220	-0.394137	
2714	1.402830	-1.351135	-0.690695	-0.028403	0.286839	
2715	1.080950	-0.156888	-0.010148	-0.012768	3.447158	

	AssignToHotkeys	UniqueHotkeys	MinimapAttacks	MinimapRightClicks	\
0	-0.786519	-1.859269	-0.601104	-0.685494	
1	0.592114	1.504724	-0.601104	-0.440049	
2	2.064971	0.663726	0.005912	2.201970	
3	-0.478394	-1.438770	-0.566649	-0.930200	
4	-0.143502	0.243227	-0.146569	1.028012	
...	...	...	...	...	
2711	0.896602	0.243227	3.547702	0.295307	
2712	0.248477	-0.597771	8.596995	-0.306486	
2713	0.374336	-0.597771	-0.545344	0.064697	
2714	0.485182	-0.597771	-0.421196	1.967961	
2715	1.195197	2.345722	0.633667	0.224789	

	NumberOfPACs	GapBetweenPACs	ActionLatency	TotalMapExplored	\
0	0.105141	-0.175572	-0.541348	-1.228037	
1	0.830187	-0.922867	-0.389448	1.180543	
2	1.339523	-1.595382	-1.465478	-0.558987	

3	-1.611500	3.704689	2.495767	0.110063
4	-0.351144	1.588008	0.087515	-1.361847
...	...	...	...	...
2711	0.092809	-0.988751	-0.425834	1.046733
2712	1.165477	-0.770078	-0.825397	0.377683
2713	-0.068947	0.588303	-0.521695	0.377683
2714	-1.046775	-0.717350	0.367619	0.243873
2715	0.852495	-1.000858	-1.259717	-0.425177

	WorkersMade	ComplexUnitsMade	ComplexAbilitiesUsed
0	0.228589	-0.534712	-0.533526
1	0.940189	-0.272220	-0.533526
2	0.626469	-0.534712	2.018332
3	-1.032991	-0.088638	0.190684
4	-0.146098	-0.534712	-0.533526
...	...	...	...
2711	-0.911631	-0.534712	-0.533526
2712	3.038640	-0.231699	-0.533526
2713	0.175446	2.032074	0.939938
2714	0.035378	0.845552	2.275640
2715	0.333492	-0.534712	-0.533526

[2716 rows x 16 columns]

## 0.9 Model Selection

1. We use random forest classifier, gradient boosting, logistic regression and neural networks and see that it overfits on the training data and that is evident by the testing accuracy that is slightly low(42%).
2. Furthermore, we see that the model classifies the LeagueIndex 8 datapoints at a 100% accuracy as compared to the other LeagueIndex's.
3. This is clearly erroneous since we have filled AGE, TOTAL\_HOURS, HOURS\_PER\_WEEK with the median of LeagueIndex 7. This is easy for the model to identify.
4. I tried using for SMOTE for generating synthetic data since the classes were imbalanced but that does not increase the performance on the test dataset.
5. Cross validation was applied, and a 95% confidence interval was also calculated. However, on unseen data the accuracy did not increase more than 42%.
6. Hyperparameter tuning was also done for developing a robust model.
7. In this specific case, we are trying to predict one out of eight possible ranks for each player. Since the ranks are closely related and can vary by only one position, we introduced an error range of plus or minus 1. This means that if our model predicts a rank that is one position higher or lower than the actual rank, we consider it an acceptable prediction. By incorporating this error range, our accuracy significantly improves to 88%. This means that in nearly 9 out of 10 cases, our model predicts the rank either correctly or within one position of the true rank. This level of performance is quite promising, considering the closely related nature of the ranks. Even if the model predicts a rank that is off by one, it is still considered

a valuable prediction since it is very close to the actual rank.

8. Dropping the features that have a high VIF value does not help increase the performance on the test dataset.

```
[64]: # Random Forest classifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import GridSearchCV, cross_val_score, \
    ↪StratifiedKFold
from sklearn.metrics import classification_report
import numpy as np
import scipy.stats as stats

from keras.layers import Dropout
from sklearn.utils import class_weight
from imblearn.over_sampling import SMOTE
from sklearn.preprocessing import StandardScaler
from keras.optimizers import Adam

# Apply SMOTE to handle class imbalance
smote = SMOTE()
#df_train_set_prepared, y_train = smote.fit_resample(df_train_set_prepared, \
    ↪y_train)
# Create the Random Forest classifier
rf = RandomForestClassifier(bootstrap=True, oob_score=True, max_samples=0.8, \
    ↪random_state=3)

# Define the parameter grid for tuning
param_grid = {'n_estimators': [100],
              'max_depth': [None, 5, 10, 15],
              'min_samples_split': [10],
              'min_samples_leaf': [4],
              'max_features': ['sqrt', 'log2']}

# Perform grid search with cross-validation
cv = StratifiedKFold(n_splits=10, shuffle=True, random_state=42) # Stratified \
    ↪K-fold for multi-class
grid_search = GridSearchCV(rf, param_grid, scoring='f1_macro', n_jobs=-1, \
    ↪cv=cv, refit=True)
grid_search.fit(df_train_set_prepared, y_train)

# Get the best parameters and best estimator from grid search
best_params = grid_search.best_params_
best_estimator = grid_search.best_estimator_

# Use the best estimator to make predictions on the training set
y_train_model_rf = best_estimator.predict(df_train_set_prepared)
```

```

# Perform cross-validation on the best estimator
scores_model_rf = cross_val_score(best_estimator, df_train_set_prepared,
    ↪y_train, cv=cv, scoring='f1_macro')

# Print the cross-validation scores
print('Cross-Validation Scores:')
print(scores_model_rf)
confidence = 0.95
print('Training F1 Score:', np.mean(scores_model_rf))
print('95% CI:', stats.t.interval(confidence, len(scores_model_rf) - 1,
    loc=np.mean(scores_model_rf),
    scale=np.std(scores_model_rf, ddof=1) / np.
    ↪sqrt(len(scores_model_rf))))

# Evaluate performance on the test set
y_test_model_rf = best_estimator.predict(test_set_prepared)
print('Test Set Evaluation:')
print(classification_report(y_test, y_test_model_rf))

```

Cross-Validation Scores:

```

[0.42197991 0.4265486 0.41008572 0.44823598 0.39075007 0.4320571
 0.4404112 0.41490338 0.44287559 0.42667718]

```

Training F1 Score: 0.4254524710587309

95% CI: (0.4131623994947289, 0.43774254262273293)

Test Set Evaluation:

	precision	recall	f1-score	support
1	0.57	0.39	0.46	33
2	0.34	0.30	0.32	70
3	0.37	0.24	0.29	111
4	0.36	0.46	0.40	162
5	0.37	0.44	0.40	161
6	0.56	0.50	0.53	124
7	0.00	0.00	0.00	7
8	1.00	1.00	1.00	11
accuracy			0.41	679
macro avg	0.45	0.42	0.43	679
weighted avg	0.42	0.41	0.41	679

```
[ ]: best_params
```

```
[ ]: {'max_depth': None,
      'max_features': 'sqrt',

```

```
'min_samples_leaf': 4,
'min_samples_split': 10,
'n_estimators': 95}
```

```
[56]: y_test_model_rf = best_estimator.predict(df_train_set_prepared)
print('Train Set Evaluation:')
print(classification_report(y_train, y_test_model_rf))
# This is actually the train set result, did not run it again since I didn't
# save my model, we can clearly see that the model is overfitting
# on the training data.
```

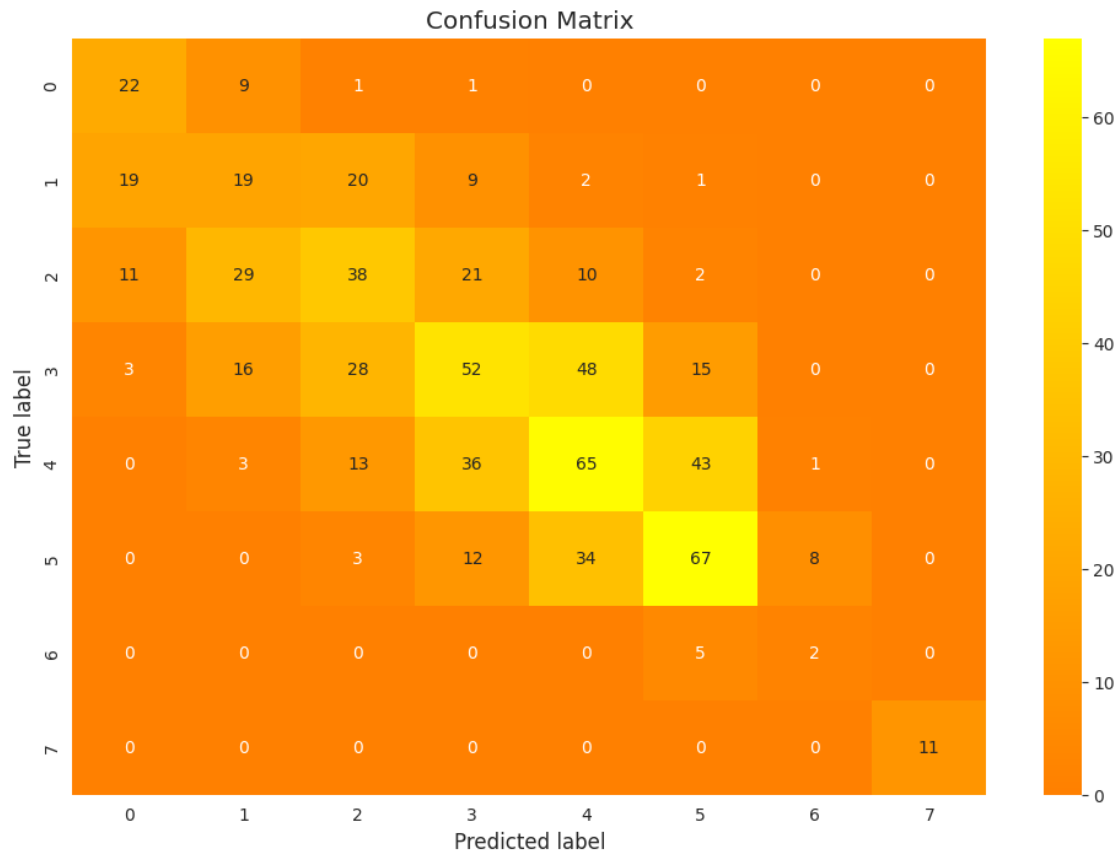
Train Set Evaluation:

	precision	recall	f1-score	support
1	0.93	0.99	0.96	649
2	0.94	0.97	0.95	649
3	0.94	0.91	0.92	649
4	0.94	0.86	0.89	649
5	0.93	0.87	0.90	649
6	0.87	0.94	0.90	649
7	0.98	1.00	0.99	649
8	1.00	1.00	1.00	649
accuracy			0.94	5192
macro avg	0.94	0.94	0.94	5192
weighted avg	0.94	0.94	0.94	5192

```
[57]: def draw_confusion_matrix(cm):
plt.figure(figsize=(12,8))
sns.heatmap(cm,annot=True,fmt="d", center=0, cmap='autumn')
plt.title("Confusion Matrix")
plt.ylabel('True label')
plt.xlabel('Predicted label')
plt.show()
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix
y_test_model_rf = best_estimator.predict(test_set_prepared)
print('Test Set Evaluation:')
cm_rf = confusion_matrix(y_test, y_test_model_rf)
draw_confusion_matrix(cm_rf)
```

Test Set Evaluation:





```
[65]: def calculate_accuracy(true_labels, predicted_labels):
    total_predictions = len(true_labels)
    correct_predictions = 0

    for true_label, predicted_label in zip(true_labels, predicted_labels):
        if abs(predicted_label - true_label) <= 1:
            correct_predictions += 1

    accuracy = (correct_predictions / total_predictions) * 100
    return accuracy
calculate_accuracy(y_test, y_test_model_rf)
```

[65]: 87.77614138438881

```
[ ]: # Gradient Boosting Classifier
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.model_selection import GridSearchCV, cross_val_score,
    StratifiedKFold
from sklearn.metrics import classification_report
import numpy as np
```

```

import scipy.stats as stats

# Create the Gradient Boosting classifier
gb = GradientBoostingClassifier(random_state=3)

# Define the parameter grid for tuning
param_grid = {'n_estimators': [100, 200, 300],
              'learning_rate': [0.1, 0.01, 0.001],
              'max_depth': [3, 5, 7]}

# Perform grid search with cross-validation
cv = StratifiedKFold(n_splits=10, shuffle=True, random_state=42) # Stratified
    ↳ K-fold for multi-class
grid_search = GridSearchCV(gb, param_grid, scoring='f1_macro', n_jobs=-1,
    ↳ cv=cv, refit=True)
grid_search.fit(df_train_set_prepared, y_train)

# Get the best parameters and best estimator from grid search
best_params = grid_search.best_params_
best_estimator = grid_search.best_estimator_

# Use the best estimator to make predictions on the training set
y_train_model_gb = best_estimator.predict(df_train_set_prepared)

# Perform cross-validation on the best estimator
scores_model_gb = cross_val_score(best_estimator, df_train_set_prepared,
    ↳ y_train, cv=cv, scoring='f1_macro')

# Print the cross-validation scores
print('Cross-Validation Scores:')
print(scores_model_gb)
confidence = 0.95
print('Training F1 Score:', np.mean(scores_model_gb))
print('95% CI:', stats.t.interval(confidence, len(scores_model_gb) - 1,
                                loc=np.mean(scores_model_gb),
                                scale=np.std(scores_model_gb, ddof=1) / np.
    ↳ sqrt(len(scores_model_gb))))

# Evaluate performance on the test set
y_test_model_gb = best_estimator.predict(test_set_prepared)
print('Test Set Evaluation:')
print(classification_report(y_test, y_test_model_gb))

```

Cross-Validation Scores:

```

[0.46310942 0.40819944 0.38094769 0.42026116 0.40590591 0.39066955
 0.47724717 0.38618565 0.45498668 0.41665838]

```

Training F1 Score: 0.4204171040090823  
 95% CI: (0.3962922725246684, 0.44454193549349613)  
 Test Set Evaluation:

	precision	recall	f1-score	support
1	0.58	0.45	0.51	33
2	0.34	0.31	0.33	70
3	0.37	0.23	0.28	111
4	0.37	0.45	0.41	162
5	0.37	0.44	0.40	161
6	0.55	0.53	0.54	124
7	1.00	0.14	0.25	7
8	1.00	1.00	1.00	11
accuracy			0.42	679
macro avg	0.57	0.45	0.46	679
weighted avg	0.43	0.42	0.41	679

```
[66]: #Logistic Regression
log_reg = LogisticRegression(multi_class='multinomial', solver='saga',
    ↪random_state=0)

param_grid = {'penalty': ['l1', 'l2', 'none', 'elasticnet'],
              'C': np.linspace(0.5, 1.5, 100)}

grid_search = GridSearchCV(log_reg, param_grid,
                           scoring='accuracy',
                           n_jobs=-1,
                           cv=10,
                           refit=True)
grid_search.fit(train_set_prepared, y_train)

grid_search.best_params_
```

```
[66]: {'C': 0.5202020202020202, 'penalty': 'l1'}
```

```
[67]: log_reg_final = grid_search.best_estimator_

y_train_model6_logreg = log_reg_final.predict(train_set_prepared)
```

```
[68]: scores_model6_logreg = cross_val_score(log_reg_final,
                                             train_set_prepared, y_train,
                                             n_jobs=-1,
                                             cv=10,
                                             scoring='accuracy')
```

```

print(scores_model6_logreg)
confidence = 0.95
print('Training accuracy: ', accuracy_score(y_train, y_train_model6_logreg))
print('95% CI: ', stats.t.interval(confidence, len(scores_model6_logreg) - 1,
                                   loc=scores_model6_logreg.mean(),
                                   scale=scores_model6_logreg.std(ddof=1)/np.
                                   sqrt(len(scores_model6_logreg))))

```

```

[0.40073529 0.43014706 0.40808824 0.40441176 0.42279412 0.41544118
 0.42066421 0.4501845  0.39114391 0.4095941 ]

```

Training accuracy: 0.43041237113402064

95% CI: (0.4033476521249959, 0.4272932204602006)

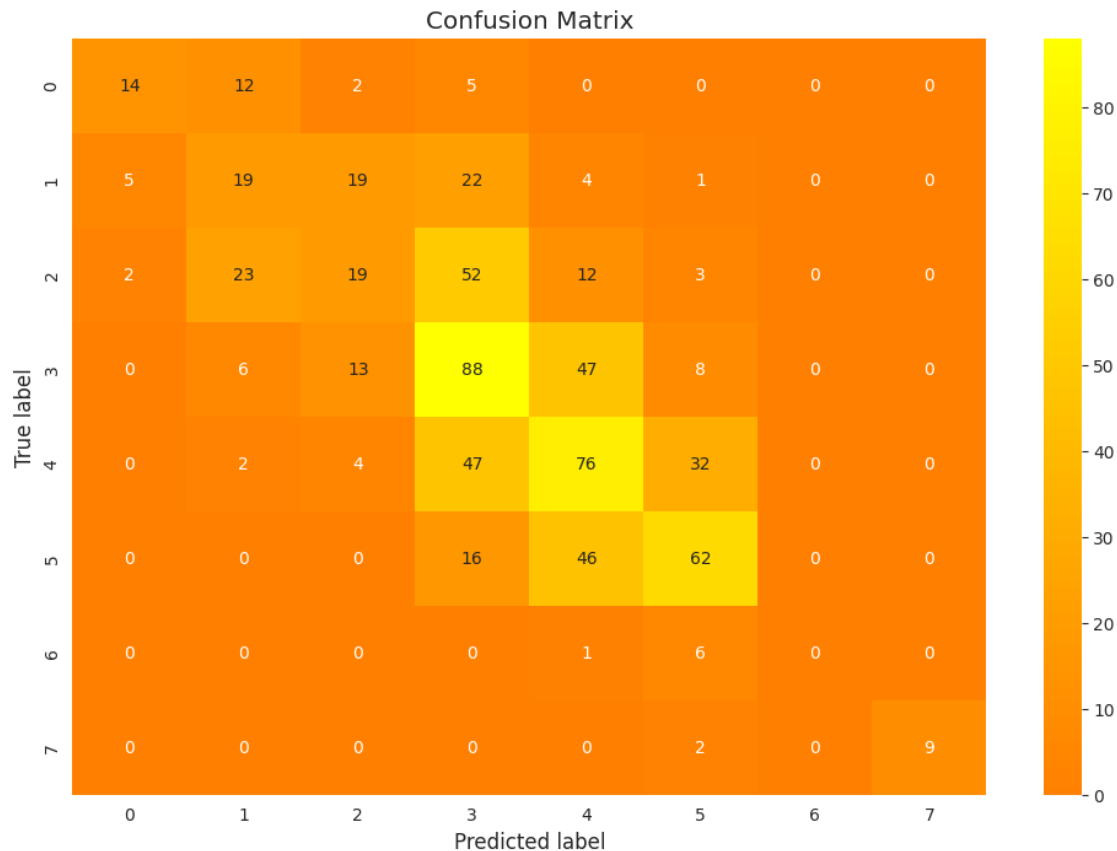
```

[70]: def draw_confusion_matrix(cm):
        plt.figure(figsize=(12,8))
        sns.heatmap(cm,annot=True,fmt="d", center=0, cmap='autumn')
        plt.title("Confusion Matrix")
        plt.ylabel('True label')
        plt.xlabel('Predicted label')
        plt.show()

y_test_model_rf = log_reg_final.predict(test_set_prepared)
print('Test Set Evaluation:')
cm_rf = confusion_matrix(y_test, y_test_model_rf)
draw_confusion_matrix(cm_rf)

```

Test Set Evaluation:



```
[71]: def calculate_accuracy(true_labels, predicted_labels):
    total_predictions = len(true_labels)
    correct_predictions = 0

    for true_label, predicted_label in zip(true_labels, predicted_labels):
        if abs(predicted_label - true_label) <= 1:
            correct_predictions += 1

    accuracy = (correct_predictions / total_predictions) * 100
    return accuracy
calculate_accuracy(y_test, y_test_model_rf)
```

[71]: 86.74521354933727

```
[81]: import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.metrics import classification_report
from keras.models import Sequential
from keras.layers import Dense
from keras.wrappers.scikit_learn import KerasClassifier
```

```

from sklearn.model_selection import GridSearchCV, StratifiedKFold
from sklearn.metrics import f1_score

# Define a function to create the neural network model
def create_model(learn_rate=0.01):
    model = Sequential()
    model.add(Dense(128, activation='relu', input_shape=(16,)))
    model.add(Dropout(0.2))
    model.add(Dense(128, activation='relu'))
    model.add(Dropout(0.2))
    model.add(Dense(8, activation='softmax'))
    optimizer = Adam(lr=learn_rate)
    model.compile(optimizer=optimizer, loss='categorical_crossentropy',
metrics=['accuracy'])
    return model

# Modify param_grid
param_grid = {'batch_size': [16,32],
              'epochs': [200],
              'learn_rate': [0.001, 0.01, 0.1]}

# Create the KerasClassifier wrapper
model = KerasClassifier(build_fn=create_model)

# Perform grid search with cross-validation
cv = StratifiedKFold(n_splits=5, shuffle=True, random_state=42) # Stratified
K-fold for multi-class
grid_search = GridSearchCV(estimator=model, param_grid=param_grid,
scoring='f1_macro', cv=cv, refit=True)
grid_search.fit(df_train_set_prepared, y_train)

# Get the best parameters and best estimator from grid search
best_params = grid_search.best_params_
best_estimator = grid_search.best_estimator_

# Use the best estimator to make predictions on the training set
y_train_model_nn = best_estimator.predict(df_train_set_prepared)

# Convert the predicted labels to match the shape of y_train (if necessary)

# Perform cross-validation on the best estimator
scores_model_nn = cross_val_score(best_estimator, df_train_set_prepared,
y_train, cv=cv, scoring='f1_macro')

# Print the cross-validation scores
print('Cross-Validation Scores:')
print(scores_model_nn)
confidence = 0.95

```

```

print('Training F1 Score:', np.mean(scores_model_nn))
print('95% CI:', stats.t.interval(confidence, len(scores_model_nn) - 1,
                                loc=np.mean(scores_model_nn),
                                scale=np.std(scores_model_nn, ddof=1) / np.
                                sqrt(len(scores_model_nn))))

# Evaluate performance on the test set
y_test_model_nn = best_estimator.predict(test_set_prepared)
# Convert the predicted labels to match the shape of y_test (if necessary)

print('Test Set Evaluation:')
print(classification_report(y_test, y_test_model_nn))

```

Streaming output truncated to the last 5000 lines.

```

Epoch 118/200
68/68 [=====] - 0s 2ms/step - loss: 0.7708 - accuracy:
0.6861
Epoch 119/200
68/68 [=====] - 0s 2ms/step - loss: 0.7098 - accuracy:
0.7124
Epoch 120/200
68/68 [=====] - 0s 2ms/step - loss: 0.7560 - accuracy:
0.7073
Epoch 121/200
68/68 [=====] - 0s 2ms/step - loss: 0.7428 - accuracy:
0.6931
Epoch 122/200
68/68 [=====] - 0s 2ms/step - loss: 0.7232 - accuracy:
0.7082
Epoch 123/200
68/68 [=====] - 0s 2ms/step - loss: 0.7768 - accuracy:
0.6898
Epoch 124/200
68/68 [=====] - 0s 2ms/step - loss: 0.7748 - accuracy:
0.6815
Epoch 125/200
68/68 [=====] - 0s 2ms/step - loss: 0.7383 - accuracy:
0.7055
Epoch 126/200
68/68 [=====] - 0s 2ms/step - loss: 0.7386 - accuracy:
0.7073
Epoch 127/200
68/68 [=====] - 0s 2ms/step - loss: 0.7513 - accuracy:
0.7082
Epoch 128/200
68/68 [=====] - 0s 2ms/step - loss: 0.7089 - accuracy:
0.7257

```

Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7242 - accuracy:  
0.7174  
Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7273 - accuracy:  
0.7101  
Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6669 - accuracy:  
0.7345  
Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7189 - accuracy:  
0.7096  
Epoch 133/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7038 - accuracy:  
0.7119  
Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6919 - accuracy:  
0.7266  
Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7164 - accuracy:  
0.7220  
Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7686 - accuracy:  
0.7082  
Epoch 137/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7664 - accuracy:  
0.7096  
Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7760 - accuracy:  
0.7069  
Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7234 - accuracy:  
0.7096  
Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7290 - accuracy:  
0.7193  
Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7270 - accuracy:  
0.7069  
Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7000 - accuracy:  
0.7303  
Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7161 - accuracy:  
0.7257  
Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7075 - accuracy:  
0.7303



Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7210 - accuracy:  
0.7225  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7009 - accuracy:  
0.7266  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6912 - accuracy:  
0.7289  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6659 - accuracy:  
0.7414  
Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6911 - accuracy:  
0.7239  
Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6794 - accuracy:  
0.7243  
Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6587 - accuracy:  
0.7312  
Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7280 - accuracy:  
0.7312  
Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6932 - accuracy:  
0.7317  
Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7618 - accuracy:  
0.7027  
Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6824 - accuracy:  
0.7271  
Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6883 - accuracy:  
0.7464  
Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6728 - accuracy:  
0.7294  
Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6703 - accuracy:  
0.7276  
Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6513 - accuracy:  
0.7478  
Epoch 160/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7161 - accuracy:  
0.7262

Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6899 - accuracy:  
0.7349  
Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7075 - accuracy:  
0.7073  
Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6887 - accuracy:  
0.7331  
Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6874 - accuracy:  
0.7253  
Epoch 165/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6242 - accuracy:  
0.7520  
Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6863 - accuracy:  
0.7395  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6885 - accuracy:  
0.7395  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6669 - accuracy:  
0.7446  
Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6882 - accuracy:  
0.7289  
Epoch 170/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6737 - accuracy:  
0.7372  
Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7110 - accuracy:  
0.7340  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7072 - accuracy:  
0.7317  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6981 - accuracy:  
0.7391  
Epoch 174/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6867 - accuracy:  
0.7308  
Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6757 - accuracy:  
0.7391  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6633 - accuracy:  
0.7474

Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6909 - accuracy: 0.7455

Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6385 - accuracy: 0.7529

Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6339 - accuracy: 0.7506

Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6612 - accuracy: 0.7547

Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6706 - accuracy: 0.7382

Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6739 - accuracy: 0.7510

Epoch 183/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6850 - accuracy: 0.7354

Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6705 - accuracy: 0.7400

Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6852 - accuracy: 0.7441

Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6524 - accuracy: 0.7515

Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6219 - accuracy: 0.7533

Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6177 - accuracy: 0.7621

Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6541 - accuracy: 0.7524

Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6645 - accuracy: 0.7561

Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6325 - accuracy: 0.7607

Epoch 192/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6560 - accuracy: 0.7506

Epoch 193/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6785 - accuracy: 0.7363

Epoch 194/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6213 - accuracy: 0.7515

Epoch 195/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6638 - accuracy: 0.7520

Epoch 196/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6485 - accuracy: 0.7566

Epoch 197/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6509 - accuracy: 0.7566

Epoch 198/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6909 - accuracy: 0.7395

Epoch 199/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6677 - accuracy: 0.7497

Epoch 200/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6733 - accuracy: 0.7437  
17/17 [=====] - 0s 1ms/step

Epoch 1/200  
68/68 [=====] - 1s 2ms/step - loss: 1.5961 - accuracy: 0.3290

Epoch 2/200  
68/68 [=====] - 0s 2ms/step - loss: 1.4153 - accuracy: 0.3594

Epoch 3/200  
68/68 [=====] - 0s 3ms/step - loss: 1.3686 - accuracy: 0.3912

Epoch 4/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3705 - accuracy: 0.4004

Epoch 5/200  
68/68 [=====] - 0s 3ms/step - loss: 1.3607 - accuracy: 0.3990

Epoch 6/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3492 - accuracy: 0.4027

Epoch 7/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3383 - accuracy: 0.4013

Epoch 8/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3353 - accuracy:

0.4100  
Epoch 9/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3155 - accuracy:  
0.4114  
Epoch 10/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2963 - accuracy:  
0.4229  
Epoch 11/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2796 - accuracy:  
0.4326  
Epoch 12/200  
68/68 [=====] - 0s 3ms/step - loss: 1.2739 - accuracy:  
0.4363  
Epoch 13/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3222 - accuracy:  
0.4202  
Epoch 14/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2686 - accuracy:  
0.4372  
Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2409 - accuracy:  
0.4330  
Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2397 - accuracy:  
0.4501  
Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2320 - accuracy:  
0.4501  
Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2233 - accuracy:  
0.4694  
Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2104 - accuracy:  
0.4694  
Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1999 - accuracy:  
0.4666  
Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1872 - accuracy:  
0.4855  
Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1858 - accuracy:  
0.4818  
Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1694 - accuracy:  
0.4818  
Epoch 24/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1499 - accuracy:

```

0.4929
Epoch 25/200
68/68 [=====] - 0s 2ms/step - loss: 1.1558 - accuracy:
0.4906
Epoch 26/200
68/68 [=====] - 0s 2ms/step - loss: 1.1765 - accuracy:
0.4883
Epoch 27/200
68/68 [=====] - 0s 2ms/step - loss: 1.1532 - accuracy:
0.5007
Epoch 28/200
68/68 [=====] - 0s 2ms/step - loss: 1.1533 - accuracy:
0.5090
Epoch 29/200
68/68 [=====] - 0s 2ms/step - loss: 1.1283 - accuracy:
0.5085
Epoch 30/200
68/68 [=====] - 0s 2ms/step - loss: 1.0954 - accuracy:
0.5177
Epoch 31/200
68/68 [=====] - 0s 2ms/step - loss: 1.1236 - accuracy:
0.5191
Epoch 32/200
68/68 [=====] - 0s 2ms/step - loss: 1.0817 - accuracy:
0.5255
Epoch 33/200
68/68 [=====] - 0s 2ms/step - loss: 1.0831 - accuracy:
0.5421
Epoch 34/200
68/68 [=====] - 0s 2ms/step - loss: 1.0827 - accuracy:
0.5223
Epoch 35/200
68/68 [=====] - 0s 2ms/step - loss: 1.0851 - accuracy:
0.5255
Epoch 36/200
68/68 [=====] - 0s 2ms/step - loss: 1.0667 - accuracy:
0.5292
Epoch 37/200
68/68 [=====] - 0s 2ms/step - loss: 1.0686 - accuracy:
0.5444
Epoch 38/200
68/68 [=====] - 0s 2ms/step - loss: 1.0697 - accuracy:
0.5366
Epoch 39/200
68/68 [=====] - 0s 2ms/step - loss: 1.0436 - accuracy:
0.5453
Epoch 40/200
68/68 [=====] - 0s 2ms/step - loss: 1.0628 - accuracy:

```

0.5495  
 Epoch 41/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0303 - accuracy:  
 0.5619  
 Epoch 42/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0371 - accuracy:  
 0.5550  
 Epoch 43/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0344 - accuracy:  
 0.5596  
 Epoch 44/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0132 - accuracy:  
 0.5656  
 Epoch 45/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0300 - accuracy:  
 0.5545  
 Epoch 46/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9968 - accuracy:  
 0.5785  
 Epoch 47/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9981 - accuracy:  
 0.5808  
 Epoch 48/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0006 - accuracy:  
 0.5550  
 Epoch 49/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9936 - accuracy:  
 0.5863  
 Epoch 50/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9834 - accuracy:  
 0.5872  
 Epoch 51/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9916 - accuracy:  
 0.5780  
 Epoch 52/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9648 - accuracy:  
 0.6029  
 Epoch 53/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9934 - accuracy:  
 0.5854  
 Epoch 54/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9527 - accuracy:  
 0.6079  
 Epoch 55/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9447 - accuracy:  
 0.6052  
 Epoch 56/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.9633 - accuracy:

0.5757  
Epoch 57/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9523 - accuracy:  
0.5923  
Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0192 - accuracy:  
0.5950  
Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9585 - accuracy:  
0.6065  
Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9309 - accuracy:  
0.6075  
Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9481 - accuracy:  
0.6079  
Epoch 62/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9646 - accuracy:  
0.6001  
Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9521 - accuracy:  
0.6075  
Epoch 64/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9243 - accuracy:  
0.6134  
Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9330 - accuracy:  
0.6130  
Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9327 - accuracy:  
0.6153  
Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9335 - accuracy:  
0.6213  
Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9103 - accuracy:  
0.6309  
Epoch 69/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8815 - accuracy:  
0.6249  
Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9462 - accuracy:  
0.6245  
Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8831 - accuracy:  
0.6420  
Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8969 - accuracy:



```

0.6213
Epoch 73/200
68/68 [=====] - 0s 2ms/step - loss: 0.8768 - accuracy:
0.6337
Epoch 74/200
68/68 [=====] - 0s 2ms/step - loss: 0.8663 - accuracy:
0.6420
Epoch 75/200
68/68 [=====] - 0s 2ms/step - loss: 0.8963 - accuracy:
0.6457
Epoch 76/200
68/68 [=====] - 0s 2ms/step - loss: 0.8767 - accuracy:
0.6406
Epoch 77/200
68/68 [=====] - 0s 2ms/step - loss: 0.8314 - accuracy:
0.6447
Epoch 78/200
68/68 [=====] - 0s 2ms/step - loss: 0.8868 - accuracy:
0.6374
Epoch 79/200
68/68 [=====] - 0s 2ms/step - loss: 0.8731 - accuracy:
0.6378
Epoch 80/200
68/68 [=====] - 0s 2ms/step - loss: 0.8685 - accuracy:
0.6249
Epoch 81/200
68/68 [=====] - 0s 2ms/step - loss: 0.8875 - accuracy:
0.6346
Epoch 82/200
68/68 [=====] - 0s 2ms/step - loss: 0.8416 - accuracy:
0.6576
Epoch 83/200
68/68 [=====] - 0s 2ms/step - loss: 0.8811 - accuracy:
0.6332
Epoch 84/200
68/68 [=====] - 0s 2ms/step - loss: 0.8350 - accuracy:
0.6562
Epoch 85/200
68/68 [=====] - 0s 2ms/step - loss: 0.8324 - accuracy:
0.6535
Epoch 86/200
68/68 [=====] - 0s 2ms/step - loss: 0.9187 - accuracy:
0.6263
Epoch 87/200
68/68 [=====] - 0s 2ms/step - loss: 0.8700 - accuracy:
0.6406
Epoch 88/200
68/68 [=====] - 0s 2ms/step - loss: 0.8353 - accuracy:

```

```

0.6539
Epoch 89/200
68/68 [=====] - 0s 2ms/step - loss: 0.8861 - accuracy:
0.6438
Epoch 90/200
68/68 [=====] - 0s 2ms/step - loss: 0.8742 - accuracy:
0.6424
Epoch 91/200
68/68 [=====] - 0s 2ms/step - loss: 0.8313 - accuracy:
0.6530
Epoch 92/200
68/68 [=====] - 0s 2ms/step - loss: 0.8320 - accuracy:
0.6654
Epoch 93/200
68/68 [=====] - 0s 2ms/step - loss: 0.8149 - accuracy:
0.6549
Epoch 94/200
68/68 [=====] - 0s 2ms/step - loss: 0.8242 - accuracy:
0.6526
Epoch 95/200
68/68 [=====] - 0s 2ms/step - loss: 0.8472 - accuracy:
0.6530
Epoch 96/200
68/68 [=====] - 0s 3ms/step - loss: 0.8419 - accuracy:
0.6567
Epoch 97/200
68/68 [=====] - 0s 2ms/step - loss: 0.8403 - accuracy:
0.6618
Epoch 98/200
68/68 [=====] - 0s 2ms/step - loss: 0.8718 - accuracy:
0.6535
Epoch 99/200
68/68 [=====] - 0s 2ms/step - loss: 0.8358 - accuracy:
0.6595
Epoch 100/200
68/68 [=====] - 0s 2ms/step - loss: 0.8251 - accuracy:
0.6631
Epoch 101/200
68/68 [=====] - 0s 2ms/step - loss: 0.8181 - accuracy:
0.6585
Epoch 102/200
68/68 [=====] - 0s 2ms/step - loss: 0.8112 - accuracy:
0.6696
Epoch 103/200
68/68 [=====] - 0s 2ms/step - loss: 0.8030 - accuracy:
0.6733
Epoch 104/200
68/68 [=====] - 0s 2ms/step - loss: 0.8442 - accuracy:

```

0.6622  
Epoch 105/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8200 - accuracy:  
0.6572  
Epoch 106/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7859 - accuracy:  
0.6848  
Epoch 107/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8339 - accuracy:  
0.6567  
Epoch 108/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8172 - accuracy:  
0.6590  
Epoch 109/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8297 - accuracy:  
0.6788  
Epoch 110/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7911 - accuracy:  
0.6857  
Epoch 111/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7960 - accuracy:  
0.6889  
Epoch 112/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8074 - accuracy:  
0.6815  
Epoch 113/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8099 - accuracy:  
0.6723  
Epoch 114/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8120 - accuracy:  
0.6723  
Epoch 115/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7928 - accuracy:  
0.6756  
Epoch 116/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7799 - accuracy:  
0.6935  
Epoch 117/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8446 - accuracy:  
0.6608  
Epoch 118/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8275 - accuracy:  
0.6650  
Epoch 119/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7702 - accuracy:  
0.6820  
Epoch 120/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7682 - accuracy:

```

0.6861
Epoch 121/200
68/68 [=====] - 0s 2ms/step - loss: 0.7844 - accuracy:
0.6723
Epoch 122/200
68/68 [=====] - 0s 2ms/step - loss: 0.7896 - accuracy:
0.6820
Epoch 123/200
68/68 [=====] - 0s 2ms/step - loss: 0.8235 - accuracy:
0.6599
Epoch 124/200
68/68 [=====] - 0s 2ms/step - loss: 0.8255 - accuracy:
0.6742
Epoch 125/200
68/68 [=====] - 0s 2ms/step - loss: 0.8733 - accuracy:
0.6654
Epoch 126/200
68/68 [=====] - 0s 2ms/step - loss: 0.8331 - accuracy:
0.6696
Epoch 127/200
68/68 [=====] - 0s 2ms/step - loss: 0.8292 - accuracy:
0.6673
Epoch 128/200
68/68 [=====] - 0s 2ms/step - loss: 0.8240 - accuracy:
0.6746
Epoch 129/200
68/68 [=====] - 0s 2ms/step - loss: 0.7614 - accuracy:
0.6908
Epoch 130/200
68/68 [=====] - 0s 2ms/step - loss: 0.8209 - accuracy:
0.6848
Epoch 131/200
68/68 [=====] - 0s 2ms/step - loss: 0.7969 - accuracy:
0.6705
Epoch 132/200
68/68 [=====] - 0s 2ms/step - loss: 0.7813 - accuracy:
0.6838
Epoch 133/200
68/68 [=====] - 0s 2ms/step - loss: 0.7800 - accuracy:
0.6908
Epoch 134/200
68/68 [=====] - 0s 2ms/step - loss: 0.8109 - accuracy:
0.6802
Epoch 135/200
68/68 [=====] - 0s 2ms/step - loss: 0.7901 - accuracy:
0.6852
Epoch 136/200
68/68 [=====] - 0s 2ms/step - loss: 0.7807 - accuracy:

```

0.6825  
Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7589 - accuracy:  
0.6926  
Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7871 - accuracy:  
0.6958  
Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7821 - accuracy:  
0.6783  
Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7407 - accuracy:  
0.6935  
Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7489 - accuracy:  
0.6852  
Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7623 - accuracy:  
0.6829  
Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7485 - accuracy:  
0.6908  
Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7243 - accuracy:  
0.7050  
Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8084 - accuracy:  
0.6903  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7670 - accuracy:  
0.6940  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7655 - accuracy:  
0.6981  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7151 - accuracy:  
0.7110  
Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7507 - accuracy:  
0.7059  
Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8161 - accuracy:  
0.6783  
Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7816 - accuracy:  
0.6940  
Epoch 152/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7724 - accuracy:

0.6884  
Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7275 - accuracy:  
0.7046  
Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7114 - accuracy:  
0.7050  
Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7587 - accuracy:  
0.6931  
Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7123 - accuracy:  
0.7032  
Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7287 - accuracy:  
0.7064  
Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7419 - accuracy:  
0.7032  
Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7565 - accuracy:  
0.7027  
Epoch 160/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7635 - accuracy:  
0.7050  
Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7486 - accuracy:  
0.7013  
Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7133 - accuracy:  
0.7050  
Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7272 - accuracy:  
0.7105  
Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7282 - accuracy:  
0.7138  
Epoch 165/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7612 - accuracy:  
0.7046  
Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7507 - accuracy:  
0.6986  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7397 - accuracy:  
0.7128  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7255 - accuracy:

0.7133  
 Epoch 169/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7158 - accuracy:  
 0.7179  
 Epoch 170/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7469 - accuracy:  
 0.7078  
 Epoch 171/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7510 - accuracy:  
 0.6995  
 Epoch 172/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7544 - accuracy:  
 0.7000  
 Epoch 173/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7581 - accuracy:  
 0.7009  
 Epoch 174/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7215 - accuracy:  
 0.7161  
 Epoch 175/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6925 - accuracy:  
 0.7280  
 Epoch 176/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7523 - accuracy:  
 0.7170  
 Epoch 177/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7405 - accuracy:  
 0.7096  
 Epoch 178/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7083 - accuracy:  
 0.7262  
 Epoch 179/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6938 - accuracy:  
 0.7216  
 Epoch 180/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7156 - accuracy:  
 0.7124  
 Epoch 181/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7470 - accuracy:  
 0.7138  
 Epoch 182/200  
 68/68 [=====] - 0s 3ms/step - loss: 0.7091 - accuracy:  
 0.7050  
 Epoch 183/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7146 - accuracy:  
 0.7211  
 Epoch 184/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7868 - accuracy:

0.6871  
Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7370 - accuracy: 0.7138  
Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7526 - accuracy: 0.7073  
Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7078 - accuracy: 0.7266  
Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6996 - accuracy: 0.7225  
Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7310 - accuracy: 0.7119  
Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7403 - accuracy: 0.7041  
Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7273 - accuracy: 0.7087  
Epoch 192/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7619 - accuracy: 0.7082  
Epoch 193/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7100 - accuracy: 0.7128  
Epoch 194/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7049 - accuracy: 0.7285  
Epoch 195/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6791 - accuracy: 0.7303  
Epoch 196/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7252 - accuracy: 0.7331  
Epoch 197/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7077 - accuracy: 0.7170  
Epoch 198/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7246 - accuracy: 0.7069  
Epoch 199/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7391 - accuracy: 0.7165  
Epoch 200/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6965 - accuracy:



```

0.7161
17/17 [=====] - 0s 1ms/step
Epoch 1/200
68/68 [=====] - 1s 2ms/step - loss: 2.1472 - accuracy:
0.2426
Epoch 2/200
68/68 [=====] - 0s 3ms/step - loss: 1.7977 - accuracy:
0.2316
Epoch 3/200
68/68 [=====] - 0s 2ms/step - loss: 1.8086 - accuracy:
0.2348
Epoch 4/200
68/68 [=====] - 0s 2ms/step - loss: 1.8027 - accuracy:
0.2192
Epoch 5/200
68/68 [=====] - 0s 3ms/step - loss: 1.7936 - accuracy:
0.2320
Epoch 6/200
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:
0.2247
Epoch 7/200
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy:
0.2307
Epoch 8/200
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:
0.2307
Epoch 9/200
68/68 [=====] - 0s 3ms/step - loss: 1.7923 - accuracy:
0.2413
Epoch 10/200
68/68 [=====] - 0s 2ms/step - loss: 1.8052 - accuracy:
0.2330
Epoch 11/200
68/68 [=====] - 0s 2ms/step - loss: 1.8020 - accuracy:
0.2482
Epoch 12/200
68/68 [=====] - 0s 2ms/step - loss: 1.7932 - accuracy:
0.2302
Epoch 13/200
68/68 [=====] - 0s 2ms/step - loss: 1.7918 - accuracy:
0.2486
Epoch 14/200
68/68 [=====] - 0s 2ms/step - loss: 1.7967 - accuracy:
0.2325
Epoch 15/200
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy:
0.2307
Epoch 16/200

```

68/68 [=====] - 0s 2ms/step - loss: 1.7967 - accuracy:  
0.2270  
Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7999 - accuracy:  
0.2297  
Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:  
0.2302  
Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:  
0.2362  
Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7997 - accuracy:  
0.2366  
Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2357  
Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:  
0.2426  
Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2215  
Epoch 24/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8059 - accuracy:  
0.2353  
Epoch 25/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy:  
0.2265  
Epoch 26/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy:  
0.2454  
Epoch 27/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8019 - accuracy:  
0.2320  
Epoch 28/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:  
0.2348  
Epoch 29/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy:  
0.2247  
Epoch 30/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy:  
0.2224  
Epoch 31/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:  
0.2385  
Epoch 32/200

```

68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:
0.2339
Epoch 33/200
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:
0.2330
Epoch 34/200
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:
0.2408
Epoch 35/200
68/68 [=====] - 0s 2ms/step - loss: 1.7983 - accuracy:
0.2145
Epoch 36/200
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:
0.2371
Epoch 37/200
68/68 [=====] - 0s 2ms/step - loss: 1.8126 - accuracy:
0.2399
Epoch 38/200
68/68 [=====] - 0s 2ms/step - loss: 1.7918 - accuracy:
0.2279
Epoch 39/200
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:
0.2380
Epoch 40/200
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:
0.2417
Epoch 41/200
68/68 [=====] - 0s 2ms/step - loss: 1.7994 - accuracy:
0.2302
Epoch 42/200
68/68 [=====] - 0s 2ms/step - loss: 1.7925 - accuracy:
0.2385
Epoch 43/200
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:
0.2343
Epoch 44/200
68/68 [=====] - 0s 2ms/step - loss: 1.7909 - accuracy:
0.2288
Epoch 45/200
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy:
0.2417
Epoch 46/200
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:
0.2334
Epoch 47/200
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:
0.2270
Epoch 48/200

```

68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2178  
Epoch 49/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:  
0.2440  
Epoch 50/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy:  
0.2320  
Epoch 51/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy:  
0.2357  
Epoch 52/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7917 - accuracy:  
0.2413  
Epoch 53/200  
68/68 [=====] - 0s 2ms/step - loss: 2.4085 - accuracy:  
0.2390  
Epoch 54/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2113  
Epoch 55/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2288  
Epoch 56/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:  
0.2399  
Epoch 57/200  
68/68 [=====] - 0s 3ms/step - loss: 1.8003 - accuracy:  
0.2228  
Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8009 - accuracy:  
0.2472  
Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2307  
Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:  
0.2307  
Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:  
0.2459  
Epoch 62/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7916 - accuracy:  
0.2505  
Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:  
0.2366  
Epoch 64/200

68/68 [=====] - 0s 3ms/step - loss: 1.7927 - accuracy: 0.2284  
Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7997 - accuracy: 0.2486  
Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy: 0.2366  
Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7978 - accuracy: 0.2053  
Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy: 0.2339  
Epoch 69/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy: 0.2334  
Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy: 0.2288  
Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7927 - accuracy: 0.2302  
Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy: 0.2311  
Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy: 0.2201  
Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy: 0.2320  
Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy: 0.2284  
Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7919 - accuracy: 0.2390  
Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8001 - accuracy: 0.2224  
Epoch 78/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy: 0.2251  
Epoch 79/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy: 0.2233  
Epoch 80/200

68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:  
0.2390  
Epoch 81/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2371  
Epoch 82/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2380  
Epoch 83/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7919 - accuracy:  
0.2505  
Epoch 84/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7962 - accuracy:  
0.2233  
Epoch 85/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8011 - accuracy:  
0.2454  
Epoch 86/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2302  
Epoch 87/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7923 - accuracy:  
0.2348  
Epoch 88/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:  
0.2302  
Epoch 89/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:  
0.2288  
Epoch 90/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:  
0.2408  
Epoch 91/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:  
0.2449  
Epoch 92/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2288  
Epoch 93/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2477  
Epoch 94/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy:  
0.2357  
Epoch 95/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:  
0.2399  
Epoch 96/200

```

68/68 [=====] - 0s 3ms/step - loss: 1.7966 - accuracy:
0.2274
Epoch 97/200
68/68 [=====] - 0s 2ms/step - loss: 1.7908 - accuracy:
0.2426
Epoch 98/200
68/68 [=====] - 0s 2ms/step - loss: 1.8001 - accuracy:
0.2297
Epoch 99/200
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:
0.2297
Epoch 100/200
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:
0.2366
Epoch 101/200
68/68 [=====] - 0s 3ms/step - loss: 1.7965 - accuracy:
0.2422
Epoch 102/200
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:
0.2334
Epoch 103/200
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:
0.2362
Epoch 104/200
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy:
0.2113
Epoch 105/200
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:
0.2403
Epoch 106/200
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:
0.2362
Epoch 107/200
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:
0.2422
Epoch 108/200
68/68 [=====] - 0s 2ms/step - loss: 1.7930 - accuracy:
0.2371
Epoch 109/200
68/68 [=====] - 0s 2ms/step - loss: 1.7904 - accuracy:
0.2541
Epoch 110/200
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:
0.2380
Epoch 111/200
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:
0.2366
Epoch 112/200

```

68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2468  
Epoch 113/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2445  
Epoch 114/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2274  
Epoch 115/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2376  
Epoch 116/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7913 - accuracy:  
0.2330  
Epoch 117/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2302  
Epoch 118/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:  
0.2371  
Epoch 119/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:  
0.2182  
Epoch 120/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2320  
Epoch 121/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:  
0.2343  
Epoch 122/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:  
0.2394  
Epoch 123/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:  
0.2449  
Epoch 124/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7991 - accuracy:  
0.2293  
Epoch 125/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7994 - accuracy:  
0.2297  
Epoch 126/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2279  
Epoch 127/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2293  
Epoch 128/200



68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:  
0.2371  
Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7918 - accuracy:  
0.2514  
Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7928 - accuracy:  
0.2403  
Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2293  
Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:  
0.2385  
Epoch 133/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2454  
Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy:  
0.2210  
Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8013 - accuracy:  
0.2233  
Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:  
0.2343  
Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2233  
Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy:  
0.2390  
Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:  
0.2353  
Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2353  
Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2334  
Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:  
0.2422  
Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7967 - accuracy:  
0.2320  
Epoch 144/200

68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2408  
Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2468  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy:  
0.2279  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:  
0.2274  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:  
0.2099  
Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:  
0.2362  
Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy:  
0.2311  
Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:  
0.2509  
Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2316  
Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:  
0.2274  
Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:  
0.2523  
Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7926 - accuracy:  
0.2343  
Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:  
0.2288  
Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy:  
0.2385  
Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8003 - accuracy:  
0.2215  
Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7922 - accuracy:  
0.2546  
Epoch 160/200

68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy:  
0.2390  
Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:  
0.2376  
Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:  
0.2408  
Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2330  
Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy:  
0.2325  
Epoch 165/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7924 - accuracy:  
0.2348  
Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2228  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:  
0.2390  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7915 - accuracy:  
0.2500  
Epoch 169/200  
68/68 [=====] - 0s 3ms/step - loss: 1.8047 - accuracy:  
0.2210  
Epoch 170/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7925 - accuracy:  
0.2371  
Epoch 171/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7916 - accuracy:  
0.2403  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2219  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2210  
Epoch 174/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy:  
0.2205  
Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:  
0.2297  
Epoch 176/200

68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:  
0.2261  
Epoch 177/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7963 - accuracy:  
0.2440  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:  
0.2343  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:  
0.2320  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:  
0.2403  
Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:  
0.2353  
Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7931 - accuracy:  
0.2357  
Epoch 183/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy:  
0.2366  
Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8001 - accuracy:  
0.2307  
Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8000 - accuracy:  
0.2270  
Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy:  
0.2445  
Epoch 187/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7921 - accuracy:  
0.2357  
Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:  
0.2297  
Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2371  
Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7923 - accuracy:  
0.2422  
Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:  
0.2376  
Epoch 192/200

```

68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:
0.2256
Epoch 193/200
68/68 [=====] - 0s 2ms/step - loss: 1.7921 - accuracy:
0.2311
Epoch 194/200
68/68 [=====] - 0s 2ms/step - loss: 1.7983 - accuracy:
0.2284
Epoch 195/200
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:
0.2436
Epoch 196/200
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy:
0.2159
Epoch 197/200
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:
0.2270
Epoch 198/200
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy:
0.2371
Epoch 199/200
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy:
0.2320
Epoch 200/200
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:
0.2325
17/17 [=====] - 0s 1ms/step
Epoch 1/200
68/68 [=====] - 1s 2ms/step - loss: 2.5244 - accuracy:
0.2301
Epoch 2/200
68/68 [=====] - 0s 2ms/step - loss: 1.8238 - accuracy:
0.2402
Epoch 3/200
68/68 [=====] - 0s 2ms/step - loss: 1.8257 - accuracy:
0.2186
Epoch 4/200
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:
0.2398
Epoch 5/200
68/68 [=====] - 0s 2ms/step - loss: 1.8070 - accuracy:
0.2301
Epoch 6/200
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:
0.2411
Epoch 7/200
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:
0.2388

```

Epoch 8/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy:  
0.2444

Epoch 9/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy:  
0.2214

Epoch 10/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:  
0.2338

Epoch 11/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:  
0.2333

Epoch 12/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy:  
0.2356

Epoch 13/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:  
0.2421

Epoch 14/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:  
0.2315

Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2453

Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8676 - accuracy:  
0.2273

Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:  
0.2379

Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8024 - accuracy:  
0.2287

Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy:  
0.2384

Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:  
0.2416

Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy:  
0.2301

Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy:  
0.2388

Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:  
0.2453

Epoch 24/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy: 0.2356  
Epoch 25/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy: 0.2315  
Epoch 26/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy: 0.2255  
Epoch 27/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy: 0.2375  
Epoch 28/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8015 - accuracy: 0.2154  
Epoch 29/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy: 0.2292  
Epoch 30/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7974 - accuracy: 0.2416  
Epoch 31/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy: 0.2402  
Epoch 32/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7993 - accuracy: 0.2264  
Epoch 33/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy: 0.2352  
Epoch 34/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy: 0.2398  
Epoch 35/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy: 0.2310  
Epoch 36/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy: 0.2453  
Epoch 37/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy: 0.2250  
Epoch 38/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy: 0.2195  
Epoch 39/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7920 - accuracy: 0.2448

Epoch 40/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7904 - accuracy: 0.2278

Epoch 41/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8007 - accuracy: 0.2540

Epoch 42/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy: 0.2342

Epoch 43/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy: 0.2370

Epoch 44/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy: 0.2439

Epoch 45/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy: 0.2526

Epoch 46/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy: 0.2375

Epoch 47/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy: 0.2329

Epoch 48/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy: 0.2388

Epoch 49/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy: 0.2287

Epoch 50/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7925 - accuracy: 0.2448

Epoch 51/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7991 - accuracy: 0.2421

Epoch 52/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy: 0.2402

Epoch 53/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy: 0.2342

Epoch 54/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7980 - accuracy: 0.2430

Epoch 55/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy: 0.2425



Epoch 56/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy:  
0.2315  
Epoch 57/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7974 - accuracy:  
0.2232  
Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:  
0.2319  
Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2296  
Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2416  
Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:  
0.2347  
Epoch 62/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7971 - accuracy:  
0.2430  
Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy:  
0.2264  
Epoch 64/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:  
0.2453  
Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy:  
0.2379  
Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2375  
Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy:  
0.2379  
Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7974 - accuracy:  
0.2292  
Epoch 69/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2296  
Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy:  
0.2434  
Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy:  
0.2301

Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8313 - accuracy: 0.2462

Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8007 - accuracy: 0.2145

Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy: 0.2439

Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy: 0.2356

Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7924 - accuracy: 0.2333

Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy: 0.2545

Epoch 78/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy: 0.2384

Epoch 79/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy: 0.2278

Epoch 80/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy: 0.2384

Epoch 81/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy: 0.2416

Epoch 82/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy: 0.2310

Epoch 83/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy: 0.2329

Epoch 84/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy: 0.2365

Epoch 85/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy: 0.2356

Epoch 86/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy: 0.2324

Epoch 87/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy: 0.2375

Epoch 88/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7980 - accuracy:  
0.2306

Epoch 89/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8047 - accuracy:  
0.2467

Epoch 90/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7932 - accuracy:  
0.2536

Epoch 91/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:  
0.2292

Epoch 92/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:  
0.2232

Epoch 93/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:  
0.2416

Epoch 94/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:  
0.2310

Epoch 95/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:  
0.2402

Epoch 96/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7918 - accuracy:  
0.2356

Epoch 97/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7995 - accuracy:  
0.2278

Epoch 98/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy:  
0.2416

Epoch 99/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2375

Epoch 100/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy:  
0.2269

Epoch 101/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:  
0.2416

Epoch 102/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2439

Epoch 103/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy:  
0.2517

Epoch 104/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy: 0.2273

Epoch 105/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8225 - accuracy: 0.2232

Epoch 106/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy: 0.2356

Epoch 107/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy: 0.2338

Epoch 108/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy: 0.2338

Epoch 109/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy: 0.2402

Epoch 110/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8009 - accuracy: 0.2287

Epoch 111/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7920 - accuracy: 0.2283

Epoch 112/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy: 0.2416

Epoch 113/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7901 - accuracy: 0.2430

Epoch 114/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy: 0.2209

Epoch 115/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy: 0.2352

Epoch 116/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy: 0.2398

Epoch 117/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy: 0.2273

Epoch 118/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy: 0.2439

Epoch 119/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8007 - accuracy: 0.2301

Epoch 120/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8026 - accuracy: 0.2356

Epoch 121/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy: 0.2347

Epoch 122/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy: 0.2121

Epoch 123/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8020 - accuracy: 0.2283

Epoch 124/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy: 0.2283

Epoch 125/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy: 0.2467

Epoch 126/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8007 - accuracy: 0.2398

Epoch 127/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7930 - accuracy: 0.2384

Epoch 128/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy: 0.2448

Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy: 0.2402

Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy: 0.2370

Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7917 - accuracy: 0.2457

Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy: 0.2306

Epoch 133/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy: 0.2526

Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy: 0.2324

Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy: 0.2384

Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2333  
Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:  
0.2319  
Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy:  
0.2283  
Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7999 - accuracy:  
0.2278  
Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy:  
0.2407  
Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7978 - accuracy:  
0.2356  
Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7989 - accuracy:  
0.2352  
Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy:  
0.2416  
Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2315  
Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2356  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy:  
0.2342  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7921 - accuracy:  
0.2333  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2365  
Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:  
0.2310  
Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7978 - accuracy:  
0.2370  
Epoch 151/200  
68/68 [=====] - 0s 3ms/step - loss: 1.8009 - accuracy:  
0.2296

Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2296  
Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy:  
0.2246  
Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:  
0.2448  
Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7999 - accuracy:  
0.2315  
Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:  
0.2554  
Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:  
0.2342  
Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:  
0.2411  
Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2365  
Epoch 160/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7965 - accuracy:  
0.2439  
Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:  
0.2407  
Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:  
0.2421  
Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:  
0.2430  
Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7999 - accuracy:  
0.2453  
Epoch 165/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2177  
Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7999 - accuracy:  
0.2480  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:  
0.2315

Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2333  
Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7974 - accuracy:  
0.2361  
Epoch 170/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7953 - accuracy:  
0.2342  
Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:  
0.2324  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy:  
0.2398  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7983 - accuracy:  
0.2260  
Epoch 174/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2269  
Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2393  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy:  
0.2315  
Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2365  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:  
0.2356  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:  
0.2250  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy:  
0.2273  
Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:  
0.2448  
Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:  
0.2457  
Epoch 183/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy:  
0.2388



Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy: 0.2269

Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy: 0.2457

Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7931 - accuracy: 0.2416

Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8000 - accuracy: 0.2338

Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy: 0.2204

Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy: 0.2287

Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7923 - accuracy: 0.2411

Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7983 - accuracy: 0.2310

Epoch 192/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy: 0.2411

Epoch 193/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy: 0.2352

Epoch 194/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7991 - accuracy: 0.2301

Epoch 195/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy: 0.2370

Epoch 196/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy: 0.2200

Epoch 197/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy: 0.2292

Epoch 198/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy: 0.2241

Epoch 199/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy: 0.2347

Epoch 200/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7992 - accuracy: 0.2237  
17/17 [=====] - 0s 2ms/step  
Epoch 1/200  
68/68 [=====] - 0s 2ms/step - loss: 2.2274 - accuracy: 0.2388  
Epoch 2/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8034 - accuracy: 0.2352  
Epoch 3/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy: 0.2467  
Epoch 4/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy: 0.2388  
Epoch 5/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8018 - accuracy: 0.2365  
Epoch 6/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy: 0.2292  
Epoch 7/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy: 0.2411  
Epoch 8/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy: 0.2296  
Epoch 9/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7992 - accuracy: 0.2421  
Epoch 10/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8029 - accuracy: 0.2402  
Epoch 11/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy: 0.2356  
Epoch 12/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy: 0.2269  
Epoch 13/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy: 0.2237  
Epoch 14/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy: 0.2296  
Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8101 - accuracy:

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0.2434
Epoch 16/200
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:
0.2273
Epoch 17/200
68/68 [=====] - 0s 2ms/step - loss: 1.8002 - accuracy:
0.2269
Epoch 18/200
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:
0.2241
Epoch 19/200
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:
0.2370
Epoch 20/200
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:
0.2379
Epoch 21/200
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:
0.2324
Epoch 22/200
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:
0.2255
Epoch 23/200
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:
0.2103
Epoch 24/200
68/68 [=====] - 0s 2ms/step - loss: 1.8140 - accuracy:
0.2310
Epoch 25/200
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:
0.2342
Epoch 26/200
68/68 [=====] - 0s 2ms/step - loss: 1.8005 - accuracy:
0.2181
Epoch 27/200
68/68 [=====] - 0s 2ms/step - loss: 1.8007 - accuracy:
0.2338
Epoch 28/200
68/68 [=====] - 0s 2ms/step - loss: 1.8023 - accuracy:
0.2168
Epoch 29/200
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:
0.2485
Epoch 30/200
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:
0.2508
Epoch 31/200
68/68 [=====] - 0s 2ms/step - loss: 1.8026 - accuracy:

```

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0.2425
Epoch 32/200
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:
0.2434
Epoch 33/200
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:
0.2347
Epoch 34/200
68/68 [=====] - 0s 2ms/step - loss: 1.8071 - accuracy:
0.2338
Epoch 35/200
68/68 [=====] - 0s 2ms/step - loss: 1.8600 - accuracy:
0.2250
Epoch 36/200
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:
0.2453
Epoch 37/200
68/68 [=====] - 0s 2ms/step - loss: 1.8912 - accuracy:
0.2490
Epoch 38/200
68/68 [=====] - 0s 2ms/step - loss: 1.8034 - accuracy:
0.2310
Epoch 39/200
68/68 [=====] - 0s 2ms/step - loss: 1.7932 - accuracy:
0.2444
Epoch 40/200
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:
0.2398
Epoch 41/200
68/68 [=====] - 0s 2ms/step - loss: 1.7919 - accuracy:
0.2246
Epoch 42/200
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy:
0.2476
Epoch 43/200
68/68 [=====] - 0s 2ms/step - loss: 1.8034 - accuracy:
0.2306
Epoch 44/200
68/68 [=====] - 0s 3ms/step - loss: 1.7992 - accuracy:
0.2232
Epoch 45/200
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:
0.2315
Epoch 46/200
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:
0.2402
Epoch 47/200
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:

```

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0.2365
Epoch 48/200
68/68 [=====] - 0s 2ms/step - loss: 1.8023 - accuracy:
0.2365
Epoch 49/200
68/68 [=====] - 0s 2ms/step - loss: 1.7974 - accuracy:
0.2338
Epoch 50/200
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:
0.2324
Epoch 51/200
68/68 [=====] - 0s 2ms/step - loss: 1.8029 - accuracy:
0.2250
Epoch 52/200
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy:
0.2384
Epoch 53/200
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:
0.2246
Epoch 54/200
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy:
0.2388
Epoch 55/200
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:
0.2333
Epoch 56/200
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:
0.2264
Epoch 57/200
68/68 [=====] - 0s 2ms/step - loss: 1.8027 - accuracy:
0.2292
Epoch 58/200
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:
0.2264
Epoch 59/200
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:
0.2319
Epoch 60/200
68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy:
0.2310
Epoch 61/200
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:
0.2379
Epoch 62/200
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:
0.2398
Epoch 63/200
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:

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0.2333
Epoch 64/200
68/68 [=====] - 0s 2ms/step - loss: 1.8072 - accuracy:
0.2250
Epoch 65/200
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:
0.2370
Epoch 66/200
68/68 [=====] - 0s 3ms/step - loss: 1.7970 - accuracy:
0.2457
Epoch 67/200
68/68 [=====] - 0s 2ms/step - loss: 1.7967 - accuracy:
0.2301
Epoch 68/200
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:
0.2393
Epoch 69/200
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:
0.2329
Epoch 70/200
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy:
0.2191
Epoch 71/200
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:
0.2287
Epoch 72/200
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:
0.2250
Epoch 73/200
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:
0.2342
Epoch 74/200
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:
0.2232
Epoch 75/200
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy:
0.2246
Epoch 76/200
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy:
0.2499
Epoch 77/200
68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy:
0.2333
Epoch 78/200
68/68 [=====] - 0s 2ms/step - loss: 1.7997 - accuracy:
0.2361
Epoch 79/200
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:

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0.2283
Epoch 80/200
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:
0.2352
Epoch 81/200
68/68 [=====] - 0s 2ms/step - loss: 1.7974 - accuracy:
0.2292
Epoch 82/200
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:
0.2388
Epoch 83/200
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy:
0.2421
Epoch 84/200
68/68 [=====] - 0s 2ms/step - loss: 1.7978 - accuracy:
0.2223
Epoch 85/200
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy:
0.2264
Epoch 86/200
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:
0.2379
Epoch 87/200
68/68 [=====] - 0s 2ms/step - loss: 1.8013 - accuracy:
0.2296
Epoch 88/200
68/68 [=====] - 0s 2ms/step - loss: 1.7931 - accuracy:
0.2273
Epoch 89/200
68/68 [=====] - 0s 2ms/step - loss: 1.8000 - accuracy:
0.2365
Epoch 90/200
68/68 [=====] - 0s 2ms/step - loss: 1.7993 - accuracy:
0.2338
Epoch 91/200
68/68 [=====] - 0s 2ms/step - loss: 1.8009 - accuracy:
0.2186
Epoch 92/200
68/68 [=====] - 0s 2ms/step - loss: 1.7991 - accuracy:
0.2227
Epoch 93/200
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:
0.2315
Epoch 94/200
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:
0.2393
Epoch 95/200
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:

```

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0.2232
Epoch 96/200
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy:
0.2260
Epoch 97/200
68/68 [=====] - 0s 2ms/step - loss: 1.7995 - accuracy:
0.2393
Epoch 98/200
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:
0.2457
Epoch 99/200
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:
0.2306
Epoch 100/200
68/68 [=====] - 0s 2ms/step - loss: 1.7916 - accuracy:
0.2425
Epoch 101/200
68/68 [=====] - 0s 2ms/step - loss: 1.8001 - accuracy:
0.2324
Epoch 102/200
68/68 [=====] - 0s 2ms/step - loss: 1.8013 - accuracy:
0.2471
Epoch 103/200
68/68 [=====] - 0s 2ms/step - loss: 1.7930 - accuracy:
0.2388
Epoch 104/200
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:
0.2347
Epoch 105/200
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:
0.2172
Epoch 106/200
68/68 [=====] - 0s 2ms/step - loss: 1.7983 - accuracy:
0.2324
Epoch 107/200
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:
0.2347
Epoch 108/200
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy:
0.2407
Epoch 109/200
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:
0.2292
Epoch 110/200
68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy:
0.2232
Epoch 111/200
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:

```



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0.2398
Epoch 112/200
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:
0.2384
Epoch 113/200
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:
0.2476
Epoch 114/200
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:
0.2315
Epoch 115/200
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:
0.2379
Epoch 116/200
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:
0.2393
Epoch 117/200
68/68 [=====] - 0s 2ms/step - loss: 1.7997 - accuracy:
0.2209
Epoch 118/200
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:
0.2480
Epoch 119/200
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy:
0.2319
Epoch 120/200
68/68 [=====] - 0s 2ms/step - loss: 1.7992 - accuracy:
0.2209
Epoch 121/200
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:
0.2255
Epoch 122/200
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:
0.2338
Epoch 123/200
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy:
0.2476
Epoch 124/200
68/68 [=====] - 0s 2ms/step - loss: 1.8024 - accuracy:
0.2379
Epoch 125/200
68/68 [=====] - 0s 2ms/step - loss: 1.7995 - accuracy:
0.2310
Epoch 126/200
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:
0.2209
Epoch 127/200
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:

```

0.2260  
Epoch 128/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy: 0.2407  
Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy: 0.2333  
Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy: 0.2338  
Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy: 0.2485  
Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy: 0.2402  
Epoch 133/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7923 - accuracy: 0.2356  
Epoch 134/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7976 - accuracy: 0.2338  
Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy: 0.2434  
Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7980 - accuracy: 0.2457  
Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy: 0.2347  
Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy: 0.2310  
Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy: 0.2361  
Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy: 0.2237  
Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8004 - accuracy: 0.2163  
Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy: 0.2361  
Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:

0.2425  
Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2319  
Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7992 - accuracy:  
0.2407  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2342  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:  
0.2204  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy:  
0.2338  
Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy:  
0.2393  
Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy:  
0.2370  
Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2375  
Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:  
0.2255  
Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:  
0.2375  
Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:  
0.2448  
Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2361  
Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:  
0.2338  
Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:  
0.2425  
Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8035 - accuracy:  
0.2315  
Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:

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0.2338
Epoch 160/200
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:
0.2365
Epoch 161/200
68/68 [=====] - 0s 2ms/step - loss: 1.7991 - accuracy:
0.2260
Epoch 162/200
68/68 [=====] - 0s 2ms/step - loss: 1.7989 - accuracy:
0.2444
Epoch 163/200
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy:
0.2296
Epoch 164/200
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:
0.2287
Epoch 165/200
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:
0.2338
Epoch 166/200
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:
0.2283
Epoch 167/200
68/68 [=====] - 0s 2ms/step - loss: 1.8013 - accuracy:
0.2191
Epoch 168/200
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:
0.2379
Epoch 169/200
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:
0.2315
Epoch 170/200
68/68 [=====] - 0s 2ms/step - loss: 1.8027 - accuracy:
0.2204
Epoch 171/200
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy:
0.2407
Epoch 172/200
68/68 [=====] - 0s 2ms/step - loss: 1.7923 - accuracy:
0.2517
Epoch 173/200
68/68 [=====] - 0s 2ms/step - loss: 1.8003 - accuracy:
0.2352
Epoch 174/200
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:
0.2218
Epoch 175/200
68/68 [=====] - 0s 2ms/step - loss: 1.7974 - accuracy:

```

0.2135  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7918 - accuracy:  
0.2375  
Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:  
0.2250  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8010 - accuracy:  
0.2324  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy:  
0.2315  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7931 - accuracy:  
0.2338  
Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7995 - accuracy:  
0.2425  
Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7973 - accuracy:  
0.2384  
Epoch 183/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7980 - accuracy:  
0.2342  
Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy:  
0.2430  
Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:  
0.2342  
Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:  
0.2352  
Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:  
0.2407  
Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2260  
Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8000 - accuracy:  
0.2310  
Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2384  
Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:

```

0.2352
Epoch 192/200
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:
0.2347
Epoch 193/200
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:
0.2352
Epoch 194/200
68/68 [=====] - 0s 2ms/step - loss: 1.7980 - accuracy:
0.2333
Epoch 195/200
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy:
0.2365
Epoch 196/200
68/68 [=====] - 0s 2ms/step - loss: 1.8005 - accuracy:
0.2287
Epoch 197/200
68/68 [=====] - 0s 2ms/step - loss: 1.7980 - accuracy:
0.2347
Epoch 198/200
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:
0.2342
Epoch 199/200
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:
0.2439
Epoch 200/200
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:
0.2338
17/17 [=====] - 0s 1ms/step
Epoch 1/200
68/68 [=====] - 1s 2ms/step - loss: 2.2682 - accuracy:
0.2646
Epoch 2/200
68/68 [=====] - 0s 2ms/step - loss: 1.8254 - accuracy:
0.2430
Epoch 3/200
68/68 [=====] - 0s 2ms/step - loss: 1.7925 - accuracy:
0.2444
Epoch 4/200
68/68 [=====] - 0s 2ms/step - loss: 1.7921 - accuracy:
0.2296
Epoch 5/200
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy:
0.2292
Epoch 6/200
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy:
0.2352
Epoch 7/200

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68/68 [=====] - 0s 2ms/step - loss: 1.8128 - accuracy:  
0.2425  
Epoch 8/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8068 - accuracy:  
0.2319  
Epoch 9/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2204  
Epoch 10/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy:  
0.2329  
Epoch 11/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7927 - accuracy:  
0.2471  
Epoch 12/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2430  
Epoch 13/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:  
0.2209  
Epoch 14/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8013 - accuracy:  
0.2241  
Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:  
0.2324  
Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8010 - accuracy:  
0.2172  
Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8155 - accuracy:  
0.2448  
Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy:  
0.2402  
Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:  
0.2310  
Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy:  
0.2301  
Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:  
0.2379  
Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7909 - accuracy:  
0.2264  
Epoch 23/200

```

68/68 [=====] - 0s 2ms/step - loss: 1.8000 - accuracy:
0.2301
Epoch 24/200
68/68 [=====] - 0s 2ms/step - loss: 1.7919 - accuracy:
0.2480
Epoch 25/200
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:
0.2398
Epoch 26/200
68/68 [=====] - 0s 3ms/step - loss: 1.7973 - accuracy:
0.2306
Epoch 27/200
68/68 [=====] - 0s 2ms/step - loss: 1.8013 - accuracy:
0.2301
Epoch 28/200
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:
0.2416
Epoch 29/200
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy:
0.2278
Epoch 30/200
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:
0.2361
Epoch 31/200
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:
0.2342
Epoch 32/200
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:
0.2485
Epoch 33/200
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:
0.2490
Epoch 34/200
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:
0.2333
Epoch 35/200
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:
0.2365
Epoch 36/200
68/68 [=====] - 0s 3ms/step - loss: 1.7968 - accuracy:
0.2250
Epoch 37/200
68/68 [=====] - 0s 2ms/step - loss: 1.8016 - accuracy:
0.2296
Epoch 38/200
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:
0.2379
Epoch 39/200

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68/68 [=====] - 0s 2ms/step - loss: 1.7926 - accuracy:  
0.2416  
Epoch 40/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7929 - accuracy:  
0.2388  
Epoch 41/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7945 - accuracy:  
0.2310  
Epoch 42/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7930 - accuracy:  
0.2301  
Epoch 43/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8010 - accuracy:  
0.2232  
Epoch 44/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:  
0.2264  
Epoch 45/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7914 - accuracy:  
0.2393  
Epoch 46/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8033 - accuracy:  
0.2324  
Epoch 47/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7931 - accuracy:  
0.2439  
Epoch 48/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy:  
0.2365  
Epoch 49/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2398  
Epoch 50/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy:  
0.2356  
Epoch 51/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8020 - accuracy:  
0.2158  
Epoch 52/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:  
0.2273  
Epoch 53/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy:  
0.2342  
Epoch 54/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:  
0.2329  
Epoch 55/200

68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:  
0.2296  
Epoch 56/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy:  
0.2411  
Epoch 57/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2342  
Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:  
0.2278  
Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:  
0.2255  
Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:  
0.2278  
Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:  
0.2227  
Epoch 62/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2333  
Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy:  
0.2241  
Epoch 64/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:  
0.2324  
Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7917 - accuracy:  
0.2223  
Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:  
0.2287  
Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:  
0.2273  
Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2384  
Epoch 69/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7917 - accuracy:  
0.2434  
Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:  
0.2338  
Epoch 71/200

68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2269  
Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:  
0.2319  
Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8038 - accuracy:  
0.2140  
Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7994 - accuracy:  
0.2361  
Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7930 - accuracy:  
0.2411  
Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:  
0.2278  
Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7970 - accuracy:  
0.2352  
Epoch 78/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2384  
Epoch 79/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2333  
Epoch 80/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2453  
Epoch 81/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:  
0.2333  
Epoch 82/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:  
0.2319  
Epoch 83/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7921 - accuracy:  
0.2287  
Epoch 84/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8006 - accuracy:  
0.2342  
Epoch 85/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:  
0.2200  
Epoch 86/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:  
0.2421  
Epoch 87/200

68/68 [=====] - 0s 2ms/step - loss: 1.8026 - accuracy:  
0.2329  
Epoch 88/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7967 - accuracy:  
0.2388  
Epoch 89/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:  
0.2407  
Epoch 90/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:  
0.2361  
Epoch 91/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:  
0.2384  
Epoch 92/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:  
0.2324  
Epoch 93/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7900 - accuracy:  
0.2402  
Epoch 94/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2278  
Epoch 95/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:  
0.2490  
Epoch 96/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy:  
0.2329  
Epoch 97/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8005 - accuracy:  
0.2232  
Epoch 98/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7989 - accuracy:  
0.2352  
Epoch 99/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7999 - accuracy:  
0.2402  
Epoch 100/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:  
0.2329  
Epoch 101/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy:  
0.2269  
Epoch 102/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:  
0.2186  
Epoch 103/200

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68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:
0.2269
Epoch 104/200
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:
0.2241
Epoch 105/200
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:
0.2306
Epoch 106/200
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:
0.2296
Epoch 107/200
68/68 [=====] - 0s 2ms/step - loss: 1.7978 - accuracy:
0.2439
Epoch 108/200
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:
0.2333
Epoch 109/200
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:
0.2191
Epoch 110/200
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:
0.2287
Epoch 111/200
68/68 [=====] - 0s 2ms/step - loss: 1.7978 - accuracy:
0.2301
Epoch 112/200
68/68 [=====] - 0s 2ms/step - loss: 1.7983 - accuracy:
0.2227
Epoch 113/200
68/68 [=====] - 0s 2ms/step - loss: 1.9245 - accuracy:
0.2421
Epoch 114/200
68/68 [=====] - 0s 2ms/step - loss: 1.7997 - accuracy:
0.2310
Epoch 115/200
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy:
0.2375
Epoch 116/200
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:
0.2315
Epoch 117/200
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy:
0.2375
Epoch 118/200
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:
0.2490
Epoch 119/200

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68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2384  
Epoch 120/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2329  
Epoch 121/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2310  
Epoch 122/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7978 - accuracy:  
0.2434  
Epoch 123/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7994 - accuracy:  
0.2273  
Epoch 124/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7980 - accuracy:  
0.2342  
Epoch 125/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2379  
Epoch 126/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy:  
0.2310  
Epoch 127/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:  
0.2416  
Epoch 128/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7926 - accuracy:  
0.2255  
Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:  
0.2356  
Epoch 130/200  
68/68 [=====] - 0s 3ms/step - loss: 1.8003 - accuracy:  
0.2214  
Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:  
0.2370  
Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy:  
0.2306  
Epoch 133/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8015 - accuracy:  
0.2370  
Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy:  
0.2232  
Epoch 135/200

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68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:
0.2347
Epoch 136/200
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:
0.2402
Epoch 137/200
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:
0.2310
Epoch 138/200
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:
0.2283
Epoch 139/200
68/68 [=====] - 0s 2ms/step - loss: 1.7923 - accuracy:
0.2531
Epoch 140/200
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy:
0.2356
Epoch 141/200
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:
0.2283
Epoch 142/200
68/68 [=====] - 0s 2ms/step - loss: 1.7915 - accuracy:
0.2255
Epoch 143/200
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:
0.2296
Epoch 144/200
68/68 [=====] - 0s 2ms/step - loss: 1.7995 - accuracy:
0.2223
Epoch 145/200
68/68 [=====] - 0s 2ms/step - loss: 1.8010 - accuracy:
0.2269
Epoch 146/200
68/68 [=====] - 0s 2ms/step - loss: 1.7987 - accuracy:
0.2223
Epoch 147/200
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy:
0.2237
Epoch 148/200
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:
0.2333
Epoch 149/200
68/68 [=====] - 0s 2ms/step - loss: 1.8011 - accuracy:
0.2352
Epoch 150/200
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy:
0.2292
Epoch 151/200

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68/68 [=====] - 0s 2ms/step - loss: 1.7999 - accuracy:
0.2333
Epoch 152/200
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:
0.2329
Epoch 153/200
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy:
0.2333
Epoch 154/200
68/68 [=====] - 0s 2ms/step - loss: 1.7906 - accuracy:
0.2411
Epoch 155/200
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy:
0.2306
Epoch 156/200
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:
0.2375
Epoch 157/200
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:
0.2365
Epoch 158/200
68/68 [=====] - 0s 2ms/step - loss: 1.7997 - accuracy:
0.2306
Epoch 159/200
68/68 [=====] - 0s 2ms/step - loss: 1.7927 - accuracy:
0.2430
Epoch 160/200
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:
0.2246
Epoch 161/200
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy:
0.2352
Epoch 162/200
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy:
0.2439
Epoch 163/200
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy:
0.2352
Epoch 164/200
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:
0.2411
Epoch 165/200
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:
0.2393
Epoch 166/200
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:
0.2485
Epoch 167/200

```



68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2398  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:  
0.2264  
Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2333  
Epoch 170/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7916 - accuracy:  
0.2480  
Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:  
0.2319  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7928 - accuracy:  
0.2306  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7998 - accuracy:  
0.2338  
Epoch 174/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy:  
0.2476  
Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy:  
0.2310  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7927 - accuracy:  
0.2388  
Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7999 - accuracy:  
0.2223  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2379  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:  
0.2296  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy:  
0.2494  
Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:  
0.2319  
Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:  
0.2292  
Epoch 183/200

68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:  
0.2227  
Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7918 - accuracy:  
0.2370  
Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:  
0.2126  
Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:  
0.2338  
Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2324  
Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7931 - accuracy:  
0.2444  
Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7926 - accuracy:  
0.2333  
Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7962 - accuracy:  
0.2407  
Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2411  
Epoch 192/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy:  
0.2333  
Epoch 193/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy:  
0.2218  
Epoch 194/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8064 - accuracy:  
0.2347  
Epoch 195/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:  
0.2375  
Epoch 196/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:  
0.2379  
Epoch 197/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy:  
0.2398  
Epoch 198/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8004 - accuracy:  
0.2310  
Epoch 199/200

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68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:
0.2338
Epoch 200/200
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:
0.2342
17/17 [=====] - 0s 2ms/step
Epoch 1/200
68/68 [=====] - 1s 2ms/step - loss: 2.1492 - accuracy:
0.2559
Epoch 2/200
68/68 [=====] - 0s 2ms/step - loss: 1.8619 - accuracy:
0.2531
Epoch 3/200
68/68 [=====] - 0s 2ms/step - loss: 1.7966 - accuracy:
0.2306
Epoch 4/200
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:
0.2352
Epoch 5/200
68/68 [=====] - 0s 2ms/step - loss: 1.7915 - accuracy:
0.2370
Epoch 6/200
68/68 [=====] - 0s 2ms/step - loss: 1.7918 - accuracy:
0.2310
Epoch 7/200
68/68 [=====] - 0s 2ms/step - loss: 1.8054 - accuracy:
0.2260
Epoch 8/200
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:
0.2388
Epoch 9/200
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy:
0.2356
Epoch 10/200
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:
0.2149
Epoch 11/200
68/68 [=====] - 0s 2ms/step - loss: 1.8062 - accuracy:
0.2232
Epoch 12/200
68/68 [=====] - 0s 3ms/step - loss: 1.7938 - accuracy:
0.2292
Epoch 13/200
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:
0.2232
Epoch 14/200
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:
0.2342

```

Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy: 0.2287

Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy: 0.2434

Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy: 0.2338

Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy: 0.2375

Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy: 0.2444

Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy: 0.2333

Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy: 0.2402

Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy: 0.2388

Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy: 0.2283

Epoch 24/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7984 - accuracy: 0.2315

Epoch 25/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy: 0.2292

Epoch 26/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy: 0.2301

Epoch 27/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy: 0.2223

Epoch 28/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7913 - accuracy: 0.2480

Epoch 29/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy: 0.2260

Epoch 30/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy: 0.2365

Epoch 31/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2191  
Epoch 32/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7941 - accuracy:  
0.2246  
Epoch 33/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2283  
Epoch 34/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7977 - accuracy:  
0.2310  
Epoch 35/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7941 - accuracy:  
0.2425  
Epoch 36/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7927 - accuracy:  
0.2246  
Epoch 37/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:  
0.2365  
Epoch 38/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:  
0.2375  
Epoch 39/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy:  
0.2365  
Epoch 40/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy:  
0.2264  
Epoch 41/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8003 - accuracy:  
0.2306  
Epoch 42/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2310  
Epoch 43/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy:  
0.2333  
Epoch 44/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:  
0.2375  
Epoch 45/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7991 - accuracy:  
0.2145  
Epoch 46/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:  
0.2338

Epoch 47/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy: 0.2416

Epoch 48/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy: 0.2227

Epoch 49/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy: 0.2356

Epoch 50/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy: 0.2319

Epoch 51/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy: 0.2296

Epoch 52/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy: 0.2356

Epoch 53/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy: 0.2324

Epoch 54/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7959 - accuracy: 0.2365

Epoch 55/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy: 0.2411

Epoch 56/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy: 0.2379

Epoch 57/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7981 - accuracy: 0.2296

Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy: 0.2384

Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy: 0.2306

Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy: 0.2131

Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy: 0.2260

Epoch 62/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy: 0.2453

Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7995 - accuracy: 0.2306

Epoch 64/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7995 - accuracy: 0.2237

Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7903 - accuracy: 0.2306

Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7998 - accuracy: 0.2310

Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy: 0.2356

Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy: 0.2319

Epoch 69/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7933 - accuracy: 0.2324

Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy: 0.2315

Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7988 - accuracy: 0.2287

Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy: 0.2457

Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy: 0.2384

Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy: 0.2430

Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7935 - accuracy: 0.2342

Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7947 - accuracy: 0.2356

Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7914 - accuracy: 0.2310

Epoch 78/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy: 0.2434

Epoch 79/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy: 0.2283

Epoch 80/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy: 0.2434

Epoch 81/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy: 0.2370

Epoch 82/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy: 0.2411

Epoch 83/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy: 0.2480

Epoch 84/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy: 0.2315

Epoch 85/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy: 0.2430

Epoch 86/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy: 0.2168

Epoch 87/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7982 - accuracy: 0.2273

Epoch 88/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy: 0.2416

Epoch 89/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy: 0.2342

Epoch 90/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7940 - accuracy: 0.2255

Epoch 91/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy: 0.2218

Epoch 92/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7897 - accuracy: 0.2384

Epoch 93/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy: 0.2338

Epoch 94/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy: 0.2315



Epoch 95/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7989 - accuracy:  
0.2333  
Epoch 96/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7889 - accuracy:  
0.2402  
Epoch 97/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy:  
0.2168  
Epoch 98/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy:  
0.2448  
Epoch 99/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7952 - accuracy:  
0.2283  
Epoch 100/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7917 - accuracy:  
0.2388  
Epoch 101/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7965 - accuracy:  
0.2540  
Epoch 102/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8021 - accuracy:  
0.2310  
Epoch 103/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy:  
0.2342  
Epoch 104/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:  
0.2204  
Epoch 105/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2342  
Epoch 106/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:  
0.2269  
Epoch 107/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:  
0.2342  
Epoch 108/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2301  
Epoch 109/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2342  
Epoch 110/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7980 - accuracy:  
0.2352

Epoch 111/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7985 - accuracy:  
0.2333  
Epoch 112/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7972 - accuracy:  
0.2333  
Epoch 113/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7960 - accuracy:  
0.2338  
Epoch 114/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:  
0.2379  
Epoch 115/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7908 - accuracy:  
0.2490  
Epoch 116/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2411  
Epoch 117/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2186  
Epoch 118/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7946 - accuracy:  
0.2329  
Epoch 119/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2273  
Epoch 120/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy:  
0.2375  
Epoch 121/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2375  
Epoch 122/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7909 - accuracy:  
0.2416  
Epoch 123/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8002 - accuracy:  
0.2421  
Epoch 124/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7969 - accuracy:  
0.2324  
Epoch 125/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7925 - accuracy:  
0.2237  
Epoch 126/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7926 - accuracy:  
0.2352

Epoch 127/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7979 - accuracy: 0.2361

Epoch 128/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy: 0.2269

Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy: 0.2393

Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy: 0.2186

Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7920 - accuracy: 0.2375

Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7986 - accuracy: 0.2246

Epoch 133/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8022 - accuracy: 0.2315

Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7990 - accuracy: 0.2356

Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7968 - accuracy: 0.2273

Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7942 - accuracy: 0.2375

Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy: 0.2296

Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy: 0.2398

Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7939 - accuracy: 0.2407

Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7958 - accuracy: 0.2292

Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7983 - accuracy: 0.2361

Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8033 - accuracy: 0.2324

Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2444  
Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7932 - accuracy:  
0.2402  
Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy:  
0.2283  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8010 - accuracy:  
0.2296  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8015 - accuracy:  
0.2379  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2434  
Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8855 - accuracy:  
0.2375  
Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8095 - accuracy:  
0.2526  
Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7953 - accuracy:  
0.2324  
Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7920 - accuracy:  
0.2269  
Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7918 - accuracy:  
0.2342  
Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7975 - accuracy:  
0.2287  
Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7920 - accuracy:  
0.2149  
Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:  
0.2315  
Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:  
0.2494  
Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:  
0.2379

Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:  
0.2209  
Epoch 160/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:  
0.2324  
Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:  
0.2434  
Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2365  
Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2278  
Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:  
0.2384  
Epoch 165/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:  
0.2356  
Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7915 - accuracy:  
0.2370  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7920 - accuracy:  
0.2361  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 1.8018 - accuracy:  
0.2398  
Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7957 - accuracy:  
0.2273  
Epoch 170/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7955 - accuracy:  
0.2384  
Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7964 - accuracy:  
0.2329  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:  
0.2319  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7944 - accuracy:  
0.2379  
Epoch 174/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:  
0.2177

Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7951 - accuracy:  
0.2287  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:  
0.2425  
Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7943 - accuracy:  
0.2375  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:  
0.2283  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7996 - accuracy:  
0.2195  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7954 - accuracy:  
0.2273  
Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:  
0.2462  
Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7956 - accuracy:  
0.2278  
Epoch 183/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7937 - accuracy:  
0.2356  
Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7936 - accuracy:  
0.2416  
Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7926 - accuracy:  
0.2255  
Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7929 - accuracy:  
0.2352  
Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7950 - accuracy:  
0.2444  
Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2365  
Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7945 - accuracy:  
0.2283  
Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2444

Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7976 - accuracy:  
0.2255

Epoch 192/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:  
0.2388

Epoch 193/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7948 - accuracy:  
0.2375

Epoch 194/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7934 - accuracy:  
0.2246

Epoch 195/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7963 - accuracy:  
0.2365

Epoch 196/200  
68/68 [=====] - 0s 3ms/step - loss: 1.7947 - accuracy:  
0.2273

Epoch 197/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7961 - accuracy:  
0.2241

Epoch 198/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7938 - accuracy:  
0.2218

Epoch 199/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7949 - accuracy:  
0.2306

Epoch 200/200  
68/68 [=====] - 0s 2ms/step - loss: 1.7995 - accuracy:  
0.2209

17/17 [=====] - 0s 1ms/step

Epoch 1/200  
85/85 [=====] - 1s 2ms/step - loss: 1.6805 - accuracy:  
0.3225

Epoch 2/200  
85/85 [=====] - 0s 2ms/step - loss: 1.4563 - accuracy:  
0.3785

Epoch 3/200  
85/85 [=====] - 0s 2ms/step - loss: 1.4135 - accuracy:  
0.3844

Epoch 4/200  
85/85 [=====] - 0s 2ms/step - loss: 1.3825 - accuracy:  
0.4043

Epoch 5/200  
85/85 [=====] - 0s 2ms/step - loss: 1.3544 - accuracy:  
0.4061

Epoch 6/200  
85/85 [=====] - 0s 2ms/step - loss: 1.3391 - accuracy:

0.4138  
Epoch 7/200  
85/85 [=====] - 0s 2ms/step - loss: 1.3271 - accuracy: 0.4190  
Epoch 8/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2984 - accuracy: 0.4473  
Epoch 9/200  
85/85 [=====] - 0s 2ms/step - loss: 1.3000 - accuracy: 0.4359  
Epoch 10/200  
85/85 [=====] - 0s 2ms/step - loss: 1.3005 - accuracy: 0.4260  
Epoch 11/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2895 - accuracy: 0.4595  
Epoch 12/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2826 - accuracy: 0.4514  
Epoch 13/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2709 - accuracy: 0.4415  
Epoch 14/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2726 - accuracy: 0.4348  
Epoch 15/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2562 - accuracy: 0.4543  
Epoch 16/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2468 - accuracy: 0.4580  
Epoch 17/200  
85/85 [=====] - 0s 3ms/step - loss: 1.2464 - accuracy: 0.4558  
Epoch 18/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2475 - accuracy: 0.4584  
Epoch 19/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2372 - accuracy: 0.4628  
Epoch 20/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2211 - accuracy: 0.4705  
Epoch 21/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2239 - accuracy: 0.4705  
Epoch 22/200  
85/85 [=====] - 0s 2ms/step - loss: 1.2130 - accuracy:



0.4797  
 Epoch 23/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.2161 - accuracy:  
 0.4658  
 Epoch 24/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.2004 - accuracy:  
 0.4853  
 Epoch 25/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1989 - accuracy:  
 0.4753  
 Epoch 26/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1750 - accuracy:  
 0.5059  
 Epoch 27/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1835 - accuracy:  
 0.4845  
 Epoch 28/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1759 - accuracy:  
 0.4901  
 Epoch 29/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1629 - accuracy:  
 0.4982  
 Epoch 30/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1738 - accuracy:  
 0.4901  
 Epoch 31/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1710 - accuracy:  
 0.4974  
 Epoch 32/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1706 - accuracy:  
 0.4941  
 Epoch 33/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1590 - accuracy:  
 0.5052  
 Epoch 34/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1454 - accuracy:  
 0.5217  
 Epoch 35/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1527 - accuracy:  
 0.5048  
 Epoch 36/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1580 - accuracy:  
 0.5088  
 Epoch 37/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1431 - accuracy:  
 0.5092  
 Epoch 38/200  
 85/85 [=====] - 0s 2ms/step - loss: 1.1342 - accuracy:

```

0.5169
Epoch 39/200
85/85 [=====] - 0s 2ms/step - loss: 1.1290 - accuracy:
0.5140
Epoch 40/200
85/85 [=====] - 0s 2ms/step - loss: 1.1204 - accuracy:
0.5291
Epoch 41/200
85/85 [=====] - 0s 2ms/step - loss: 1.1146 - accuracy:
0.5147
Epoch 42/200
85/85 [=====] - 0s 2ms/step - loss: 1.1177 - accuracy:
0.5122
Epoch 43/200
85/85 [=====] - 0s 2ms/step - loss: 1.1082 - accuracy:
0.5195
Epoch 44/200
85/85 [=====] - 0s 2ms/step - loss: 1.0994 - accuracy:
0.5361
Epoch 45/200
85/85 [=====] - 0s 2ms/step - loss: 1.0988 - accuracy:
0.5225
Epoch 46/200
85/85 [=====] - 0s 2ms/step - loss: 1.0861 - accuracy:
0.5416
Epoch 47/200
85/85 [=====] - 0s 2ms/step - loss: 1.0953 - accuracy:
0.5346
Epoch 48/200
85/85 [=====] - 0s 2ms/step - loss: 1.0850 - accuracy:
0.5365
Epoch 49/200
85/85 [=====] - 0s 2ms/step - loss: 1.0668 - accuracy:
0.5409
Epoch 50/200
85/85 [=====] - 0s 2ms/step - loss: 1.0696 - accuracy:
0.5423
Epoch 51/200
85/85 [=====] - 0s 2ms/step - loss: 1.0715 - accuracy:
0.5493
Epoch 52/200
85/85 [=====] - 0s 2ms/step - loss: 1.0673 - accuracy:
0.5519
Epoch 53/200
85/85 [=====] - 0s 2ms/step - loss: 1.0633 - accuracy:
0.5475
Epoch 54/200
85/85 [=====] - 0s 2ms/step - loss: 1.0547 - accuracy:

```

```

0.5567
Epoch 55/200
85/85 [=====] - 0s 2ms/step - loss: 1.0559 - accuracy:
0.5482
Epoch 56/200
85/85 [=====] - 0s 2ms/step - loss: 1.0392 - accuracy:
0.5666
Epoch 57/200
85/85 [=====] - 0s 3ms/step - loss: 1.0580 - accuracy:
0.5567
Epoch 58/200
85/85 [=====] - 0s 2ms/step - loss: 1.0370 - accuracy:
0.5659
Epoch 59/200
85/85 [=====] - 0s 2ms/step - loss: 1.0210 - accuracy:
0.5770
Epoch 60/200
85/85 [=====] - 0s 2ms/step - loss: 1.0216 - accuracy:
0.5637
Epoch 61/200
85/85 [=====] - 0s 2ms/step - loss: 1.0363 - accuracy:
0.5633
Epoch 62/200
85/85 [=====] - 0s 2ms/step - loss: 1.0067 - accuracy:
0.5740
Epoch 63/200
85/85 [=====] - 0s 2ms/step - loss: 1.0213 - accuracy:
0.5644
Epoch 64/200
85/85 [=====] - 0s 2ms/step - loss: 1.0127 - accuracy:
0.5692
Epoch 65/200
85/85 [=====] - 0s 2ms/step - loss: 1.0004 - accuracy:
0.5884
Epoch 66/200
85/85 [=====] - 0s 2ms/step - loss: 1.0122 - accuracy:
0.5784
Epoch 67/200
85/85 [=====] - 0s 2ms/step - loss: 1.0018 - accuracy:
0.5810
Epoch 68/200
85/85 [=====] - 0s 2ms/step - loss: 1.0007 - accuracy:
0.5762
Epoch 69/200
85/85 [=====] - 0s 2ms/step - loss: 0.9832 - accuracy:
0.5928
Epoch 70/200
85/85 [=====] - 0s 2ms/step - loss: 0.9912 - accuracy:

```

0.5784  
Epoch 71/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9766 - accuracy: 0.5990  
Epoch 72/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9978 - accuracy: 0.5884  
Epoch 73/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9676 - accuracy: 0.5968  
Epoch 74/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9791 - accuracy: 0.5891  
Epoch 75/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9788 - accuracy: 0.5924  
Epoch 76/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9585 - accuracy: 0.5902  
Epoch 77/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9844 - accuracy: 0.5906  
Epoch 78/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9681 - accuracy: 0.6038  
Epoch 79/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9437 - accuracy: 0.5983  
Epoch 80/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9472 - accuracy: 0.6001  
Epoch 81/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9519 - accuracy: 0.6094  
Epoch 82/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9495 - accuracy: 0.5943  
Epoch 83/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9494 - accuracy: 0.6090  
Epoch 84/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9443 - accuracy: 0.6178  
Epoch 85/200  
85/85 [=====] - 0s 2ms/step - loss: 0.9297 - accuracy: 0.6123  
Epoch 86/200  
85/85 [=====] - 0s 3ms/step - loss: 0.9396 - accuracy:

0.6031  
 Epoch 87/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9290 - accuracy:  
 0.6116  
 Epoch 88/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9243 - accuracy:  
 0.6160  
 Epoch 89/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9012 - accuracy:  
 0.6337  
 Epoch 90/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9159 - accuracy:  
 0.6233  
 Epoch 91/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9231 - accuracy:  
 0.6233  
 Epoch 92/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9166 - accuracy:  
 0.6226  
 Epoch 93/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9147 - accuracy:  
 0.6226  
 Epoch 94/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9140 - accuracy:  
 0.6160  
 Epoch 95/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9063 - accuracy:  
 0.6226  
 Epoch 96/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.9022 - accuracy:  
 0.6182  
 Epoch 97/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8908 - accuracy:  
 0.6204  
 Epoch 98/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8954 - accuracy:  
 0.6333  
 Epoch 99/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8956 - accuracy:  
 0.6381  
 Epoch 100/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8910 - accuracy:  
 0.6436  
 Epoch 101/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8895 - accuracy:  
 0.6270  
 Epoch 102/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8775 - accuracy:

0.6366  
 Epoch 103/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8712 - accuracy:  
 0.6469  
 Epoch 104/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8724 - accuracy:  
 0.6414  
 Epoch 105/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8516 - accuracy:  
 0.6502  
 Epoch 106/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8638 - accuracy:  
 0.6381  
 Epoch 107/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8666 - accuracy:  
 0.6432  
 Epoch 108/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8563 - accuracy:  
 0.6443  
 Epoch 109/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8546 - accuracy:  
 0.6557  
 Epoch 110/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8570 - accuracy:  
 0.6484  
 Epoch 111/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8555 - accuracy:  
 0.6366  
 Epoch 112/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8527 - accuracy:  
 0.6521  
 Epoch 113/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8450 - accuracy:  
 0.6576  
 Epoch 114/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8457 - accuracy:  
 0.6546  
 Epoch 115/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8436 - accuracy:  
 0.6506  
 Epoch 116/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8460 - accuracy:  
 0.6502  
 Epoch 117/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8206 - accuracy:  
 0.6539  
 Epoch 118/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.8538 - accuracy:

0.6532  
Epoch 119/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8359 - accuracy:  
0.6554  
Epoch 120/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8322 - accuracy:  
0.6642  
Epoch 121/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8250 - accuracy:  
0.6550  
Epoch 122/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8269 - accuracy:  
0.6602  
Epoch 123/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8027 - accuracy:  
0.6657  
Epoch 124/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8173 - accuracy:  
0.6598  
Epoch 125/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8294 - accuracy:  
0.6642  
Epoch 126/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8274 - accuracy:  
0.6635  
Epoch 127/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8111 - accuracy:  
0.6730  
Epoch 128/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8173 - accuracy:  
0.6638  
Epoch 129/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8240 - accuracy:  
0.6605  
Epoch 130/200  
85/85 [=====] - 0s 2ms/step - loss: 0.8052 - accuracy:  
0.6719  
Epoch 131/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7868 - accuracy:  
0.6848  
Epoch 132/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7941 - accuracy:  
0.6841  
Epoch 133/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7827 - accuracy:  
0.6848  
Epoch 134/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7961 - accuracy:

0.6823  
Epoch 135/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7861 - accuracy:  
0.6808  
Epoch 136/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7919 - accuracy:  
0.6738  
Epoch 137/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7962 - accuracy:  
0.6708  
Epoch 138/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7932 - accuracy:  
0.6848  
Epoch 139/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7778 - accuracy:  
0.6870  
Epoch 140/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7860 - accuracy:  
0.6753  
Epoch 141/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7850 - accuracy:  
0.6870  
Epoch 142/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7891 - accuracy:  
0.6734  
Epoch 143/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7645 - accuracy:  
0.6881  
Epoch 144/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7757 - accuracy:  
0.6745  
Epoch 145/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7644 - accuracy:  
0.6863  
Epoch 146/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7542 - accuracy:  
0.6951  
Epoch 147/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7691 - accuracy:  
0.6918  
Epoch 148/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7758 - accuracy:  
0.6922  
Epoch 149/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7750 - accuracy:  
0.6944  
Epoch 150/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7701 - accuracy:



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0.6815
Epoch 151/200
85/85 [=====] - 0s 2ms/step - loss: 0.7473 - accuracy:
0.6915
Epoch 152/200
85/85 [=====] - 0s 2ms/step - loss: 0.7566 - accuracy:
0.6900
Epoch 153/200
85/85 [=====] - 0s 2ms/step - loss: 0.7499 - accuracy:
0.6944
Epoch 154/200
85/85 [=====] - 0s 2ms/step - loss: 0.7432 - accuracy:
0.7062
Epoch 155/200
85/85 [=====] - 0s 2ms/step - loss: 0.7610 - accuracy:
0.6922
Epoch 156/200
85/85 [=====] - 0s 2ms/step - loss: 0.7470 - accuracy:
0.6955
Epoch 157/200
85/85 [=====] - 0s 2ms/step - loss: 0.7283 - accuracy:
0.7099
Epoch 158/200
85/85 [=====] - 0s 2ms/step - loss: 0.7424 - accuracy:
0.6981
Epoch 159/200
85/85 [=====] - 0s 3ms/step - loss: 0.7226 - accuracy:
0.7047
Epoch 160/200
85/85 [=====] - 0s 2ms/step - loss: 0.7347 - accuracy:
0.7132
Epoch 161/200
85/85 [=====] - 0s 2ms/step - loss: 0.7331 - accuracy:
0.7095
Epoch 162/200
85/85 [=====] - 0s 2ms/step - loss: 0.7302 - accuracy:
0.7088
Epoch 163/200
85/85 [=====] - 0s 2ms/step - loss: 0.7393 - accuracy:
0.6966
Epoch 164/200
85/85 [=====] - 0s 2ms/step - loss: 0.7257 - accuracy:
0.7058
Epoch 165/200
85/85 [=====] - 0s 3ms/step - loss: 0.7184 - accuracy:
0.7054
Epoch 166/200
85/85 [=====] - 0s 3ms/step - loss: 0.7115 - accuracy:

```

0.7088  
Epoch 167/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7313 - accuracy:  
0.7043  
Epoch 168/200  
85/85 [=====] - 0s 3ms/step - loss: 0.7349 - accuracy:  
0.7021  
Epoch 169/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7254 - accuracy:  
0.7032  
Epoch 170/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7120 - accuracy:  
0.7172  
Epoch 171/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7415 - accuracy:  
0.7003  
Epoch 172/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7370 - accuracy:  
0.7021  
Epoch 173/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7181 - accuracy:  
0.7102  
Epoch 174/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7006 - accuracy:  
0.7117  
Epoch 175/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7271 - accuracy:  
0.6999  
Epoch 176/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7291 - accuracy:  
0.7069  
Epoch 177/200  
85/85 [=====] - 0s 2ms/step - loss: 0.6890 - accuracy:  
0.7316  
Epoch 178/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7212 - accuracy:  
0.7102  
Epoch 179/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7105 - accuracy:  
0.7172  
Epoch 180/200  
85/85 [=====] - 0s 2ms/step - loss: 0.7187 - accuracy:  
0.7117  
Epoch 181/200  
85/85 [=====] - 0s 2ms/step - loss: 0.6872 - accuracy:  
0.7250  
Epoch 182/200  
85/85 [=====] - 0s 2ms/step - loss: 0.6916 - accuracy:

```

0.7202
Epoch 183/200
85/85 [=====] - 0s 2ms/step - loss: 0.6803 - accuracy:
0.7250
Epoch 184/200
85/85 [=====] - 0s 2ms/step - loss: 0.7139 - accuracy:
0.7128
Epoch 185/200
85/85 [=====] - 0s 2ms/step - loss: 0.7017 - accuracy:
0.7135
Epoch 186/200
85/85 [=====] - 0s 2ms/step - loss: 0.7069 - accuracy:
0.7228
Epoch 187/200
85/85 [=====] - 0s 2ms/step - loss: 0.7012 - accuracy:
0.7176
Epoch 188/200
85/85 [=====] - 0s 2ms/step - loss: 0.6905 - accuracy:
0.7158
Epoch 189/200
85/85 [=====] - 0s 2ms/step - loss: 0.6916 - accuracy:
0.7187
Epoch 190/200
85/85 [=====] - 0s 2ms/step - loss: 0.6957 - accuracy:
0.7250
Epoch 191/200
85/85 [=====] - 0s 2ms/step - loss: 0.6832 - accuracy:
0.7309
Epoch 192/200
85/85 [=====] - 0s 2ms/step - loss: 0.6937 - accuracy:
0.7169
Epoch 193/200
85/85 [=====] - 0s 2ms/step - loss: 0.6814 - accuracy:
0.7246
Epoch 194/200
85/85 [=====] - 0s 2ms/step - loss: 0.6776 - accuracy:
0.7371
Epoch 195/200
85/85 [=====] - 0s 2ms/step - loss: 0.6645 - accuracy:
0.7353
Epoch 196/200
85/85 [=====] - 0s 2ms/step - loss: 0.6848 - accuracy:
0.7242
Epoch 197/200
85/85 [=====] - 0s 2ms/step - loss: 0.6790 - accuracy:
0.7338
Epoch 198/200
85/85 [=====] - 0s 2ms/step - loss: 0.6804 - accuracy:

```

0.7246  
 Epoch 199/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.6820 - accuracy:  
 0.7191  
 Epoch 200/200  
 85/85 [=====] - 0s 2ms/step - loss: 0.6671 - accuracy:  
 0.7334  
 85/85 [=====] - 0s 1ms/step  
 Epoch 1/200  
 68/68 [=====] - 1s 2ms/step - loss: 1.7171 - accuracy:  
 0.3052  
 Epoch 2/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.4887 - accuracy:  
 0.3550  
 Epoch 3/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.4207 - accuracy:  
 0.3932  
 Epoch 4/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.3752 - accuracy:  
 0.4139  
 Epoch 5/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.3740 - accuracy:  
 0.4190  
 Epoch 6/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.3521 - accuracy:  
 0.4309  
 Epoch 7/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.3307 - accuracy:  
 0.4222  
 Epoch 8/200  
 68/68 [=====] - 0s 3ms/step - loss: 1.3248 - accuracy:  
 0.4346  
 Epoch 9/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.3242 - accuracy:  
 0.4208  
 Epoch 10/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.3068 - accuracy:  
 0.4471  
 Epoch 11/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.2941 - accuracy:  
 0.4420  
 Epoch 12/200  
 68/68 [=====] - 0s 3ms/step - loss: 1.2911 - accuracy:  
 0.4268  
 Epoch 13/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.2663 - accuracy:  
 0.4576  
 Epoch 14/200

68/68 [=====] - 0s 2ms/step - loss: 1.2724 - accuracy:  
0.4540  
Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2585 - accuracy:  
0.4645  
Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2375 - accuracy:  
0.4793  
Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2408 - accuracy:  
0.4586  
Epoch 18/200  
68/68 [=====] - 0s 3ms/step - loss: 1.2371 - accuracy:  
0.4765  
Epoch 19/200  
68/68 [=====] - 0s 3ms/step - loss: 1.2317 - accuracy:  
0.4816  
Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2235 - accuracy:  
0.4728  
Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2072 - accuracy:  
0.4954  
Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2162 - accuracy:  
0.4949  
Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1992 - accuracy:  
0.4945  
Epoch 24/200  
68/68 [=====] - 0s 3ms/step - loss: 1.1950 - accuracy:  
0.4936  
Epoch 25/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2077 - accuracy:  
0.4982  
Epoch 26/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1808 - accuracy:  
0.4995  
Epoch 27/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1778 - accuracy:  
0.5041  
Epoch 28/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1750 - accuracy:  
0.5000  
Epoch 29/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1659 - accuracy:  
0.5060  
Epoch 30/200

68/68 [=====] - 0s 2ms/step - loss: 1.1576 - accuracy:  
0.5074  
Epoch 31/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1490 - accuracy:  
0.5110  
Epoch 32/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1481 - accuracy:  
0.5134  
Epoch 33/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1383 - accuracy:  
0.5345  
Epoch 34/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1425 - accuracy:  
0.5313  
Epoch 35/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1353 - accuracy:  
0.5308  
Epoch 36/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1243 - accuracy:  
0.5318  
Epoch 37/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1226 - accuracy:  
0.5267  
Epoch 38/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1212 - accuracy:  
0.5212  
Epoch 39/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1014 - accuracy:  
0.5442  
Epoch 40/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1115 - accuracy:  
0.5327  
Epoch 41/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0967 - accuracy:  
0.5534  
Epoch 42/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1023 - accuracy:  
0.5387  
Epoch 43/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0991 - accuracy:  
0.5599  
Epoch 44/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0813 - accuracy:  
0.5493  
Epoch 45/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0825 - accuracy:  
0.5520  
Epoch 46/200

68/68 [=====] - 0s 2ms/step - loss: 1.0916 - accuracy:  
0.5497  
Epoch 47/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0620 - accuracy:  
0.5668  
Epoch 48/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0529 - accuracy:  
0.5640  
Epoch 49/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0526 - accuracy:  
0.5506  
Epoch 50/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0603 - accuracy:  
0.5488  
Epoch 51/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0537 - accuracy:  
0.5709  
Epoch 52/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0377 - accuracy:  
0.5806  
Epoch 53/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0329 - accuracy:  
0.5755  
Epoch 54/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0412 - accuracy:  
0.5737  
Epoch 55/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0464 - accuracy:  
0.5594  
Epoch 56/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0291 - accuracy:  
0.5912  
Epoch 57/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0322 - accuracy:  
0.5727  
Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0215 - accuracy:  
0.5824  
Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0070 - accuracy:  
0.5902  
Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0171 - accuracy:  
0.5686  
Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0000 - accuracy:  
0.5852  
Epoch 62/200

68/68 [=====] - 0s 2ms/step - loss: 1.0026 - accuracy:  
0.5778  
Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9992 - accuracy:  
0.5760  
Epoch 64/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9972 - accuracy:  
0.5852  
Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9910 - accuracy:  
0.5948  
Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9761 - accuracy:  
0.6064  
Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9593 - accuracy:  
0.6054  
Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9785 - accuracy:  
0.5985  
Epoch 69/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9672 - accuracy:  
0.5994  
Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9684 - accuracy:  
0.6119  
Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9675 - accuracy:  
0.6013  
Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9590 - accuracy:  
0.5962  
Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9498 - accuracy:  
0.6096  
Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9377 - accuracy:  
0.6197  
Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9338 - accuracy:  
0.6156  
Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9300 - accuracy:  
0.6123  
Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9455 - accuracy:  
0.6160  
Epoch 78/200



68/68 [=====] - 0s 2ms/step - loss: 0.9190 - accuracy:  
0.6123  
Epoch 79/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9168 - accuracy:  
0.6257  
Epoch 80/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9150 - accuracy:  
0.6225  
Epoch 81/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9088 - accuracy:  
0.6225  
Epoch 82/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9086 - accuracy:  
0.6248  
Epoch 83/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9357 - accuracy:  
0.6146  
Epoch 84/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9200 - accuracy:  
0.6312  
Epoch 85/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8933 - accuracy:  
0.6363  
Epoch 86/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9031 - accuracy:  
0.6354  
Epoch 87/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8869 - accuracy:  
0.6436  
Epoch 88/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8874 - accuracy:  
0.6390  
Epoch 89/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8982 - accuracy:  
0.6404  
Epoch 90/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8787 - accuracy:  
0.6317  
Epoch 91/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8984 - accuracy:  
0.6238  
Epoch 92/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8903 - accuracy:  
0.6266  
Epoch 93/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8686 - accuracy:  
0.6547  
Epoch 94/200

```

68/68 [=====] - 0s 2ms/step - loss: 0.8905 - accuracy:
0.6363
Epoch 95/200
68/68 [=====] - 0s 2ms/step - loss: 0.8670 - accuracy:
0.6395
Epoch 96/200
68/68 [=====] - 0s 2ms/step - loss: 0.8560 - accuracy:
0.6363
Epoch 97/200
68/68 [=====] - 0s 2ms/step - loss: 0.8638 - accuracy:
0.6510
Epoch 98/200
68/68 [=====] - 0s 2ms/step - loss: 0.8795 - accuracy:
0.6404
Epoch 99/200
68/68 [=====] - 0s 2ms/step - loss: 0.8577 - accuracy:
0.6496
Epoch 100/200
68/68 [=====] - 0s 2ms/step - loss: 0.8579 - accuracy:
0.6593
Epoch 101/200
68/68 [=====] - 0s 2ms/step - loss: 0.8451 - accuracy:
0.6492
Epoch 102/200
68/68 [=====] - 0s 2ms/step - loss: 0.8329 - accuracy:
0.6653
Epoch 103/200
68/68 [=====] - 0s 3ms/step - loss: 0.8641 - accuracy:
0.6450
Epoch 104/200
68/68 [=====] - 0s 3ms/step - loss: 0.8418 - accuracy:
0.6625
Epoch 105/200
68/68 [=====] - 0s 3ms/step - loss: 0.8323 - accuracy:
0.6561
Epoch 106/200
68/68 [=====] - 0s 2ms/step - loss: 0.8292 - accuracy:
0.6616
Epoch 107/200
68/68 [=====] - 0s 3ms/step - loss: 0.8421 - accuracy:
0.6496
Epoch 108/200
68/68 [=====] - 0s 2ms/step - loss: 0.8254 - accuracy:
0.6745
Epoch 109/200
68/68 [=====] - 0s 2ms/step - loss: 0.8100 - accuracy:
0.6869
Epoch 110/200

```

68/68 [=====] - 0s 2ms/step - loss: 0.8269 - accuracy:  
0.6644  
Epoch 111/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8190 - accuracy:  
0.6685  
Epoch 112/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8225 - accuracy:  
0.6773  
Epoch 113/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8078 - accuracy:  
0.6763  
Epoch 114/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8144 - accuracy:  
0.6782  
Epoch 115/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7858 - accuracy:  
0.6823  
Epoch 116/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8166 - accuracy:  
0.6736  
Epoch 117/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7691 - accuracy:  
0.6851  
Epoch 118/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7777 - accuracy:  
0.6846  
Epoch 119/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7778 - accuracy:  
0.6901  
Epoch 120/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8041 - accuracy:  
0.6800  
Epoch 121/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7800 - accuracy:  
0.6915  
Epoch 122/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7981 - accuracy:  
0.6699  
Epoch 123/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7615 - accuracy:  
0.6966  
Epoch 124/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7785 - accuracy:  
0.6892  
Epoch 125/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7729 - accuracy:  
0.6828  
Epoch 126/200

68/68 [=====] - 0s 2ms/step - loss: 0.7696 - accuracy:  
0.6888  
Epoch 127/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7695 - accuracy:  
0.6998  
Epoch 128/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7569 - accuracy:  
0.7003  
Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7770 - accuracy:  
0.6842  
Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7622 - accuracy:  
0.7030  
Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7573 - accuracy:  
0.6961  
Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7715 - accuracy:  
0.6911  
Epoch 133/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7439 - accuracy:  
0.7090  
Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7616 - accuracy:  
0.6943  
Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7534 - accuracy:  
0.7012  
Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7429 - accuracy:  
0.6948  
Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7629 - accuracy:  
0.6800  
Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7607 - accuracy:  
0.6971  
Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7342 - accuracy:  
0.7095  
Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7487 - accuracy:  
0.6874  
Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7306 - accuracy:  
0.7238  
Epoch 142/200

68/68 [=====] - 0s 2ms/step - loss: 0.7218 - accuracy:  
0.7173  
Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7108 - accuracy:  
0.7155  
Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7381 - accuracy:  
0.7113  
Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7496 - accuracy:  
0.7081  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7058 - accuracy:  
0.7150  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7330 - accuracy:  
0.7058  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7204 - accuracy:  
0.7150  
Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7263 - accuracy:  
0.7063  
Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7200 - accuracy:  
0.7053  
Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7117 - accuracy:  
0.7224  
Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7090 - accuracy:  
0.7210  
Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6938 - accuracy:  
0.7302  
Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6851 - accuracy:  
0.7274  
Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7040 - accuracy:  
0.7215  
Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7239 - accuracy:  
0.7067  
Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7074 - accuracy:  
0.7044  
Epoch 158/200

68/68 [=====] - 0s 2ms/step - loss: 0.7039 - accuracy:  
0.7233  
Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6998 - accuracy:  
0.7196  
Epoch 160/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6818 - accuracy:  
0.7334  
Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7084 - accuracy:  
0.7205  
Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6942 - accuracy:  
0.7201  
Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6959 - accuracy:  
0.7150  
Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6846 - accuracy:  
0.7293  
Epoch 165/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7009 - accuracy:  
0.7224  
Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6766 - accuracy:  
0.7325  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6696 - accuracy:  
0.7353  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6710 - accuracy:  
0.7353  
Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6470 - accuracy:  
0.7509  
Epoch 170/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6806 - accuracy:  
0.7293  
Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6762 - accuracy:  
0.7380  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6623 - accuracy:  
0.7385  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6600 - accuracy:  
0.7343  
Epoch 174/200

68/68 [=====] - 0s 2ms/step - loss: 0.6485 - accuracy:  
0.7537  
Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6565 - accuracy:  
0.7468  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6555 - accuracy:  
0.7399  
Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6470 - accuracy:  
0.7413  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6640 - accuracy:  
0.7413  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6472 - accuracy:  
0.7551  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6379 - accuracy:  
0.7440  
Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6559 - accuracy:  
0.7445  
Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6505 - accuracy:  
0.7463  
Epoch 183/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6415 - accuracy:  
0.7380  
Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6421 - accuracy:  
0.7436  
Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6494 - accuracy:  
0.7353  
Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6706 - accuracy:  
0.7339  
Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6400 - accuracy:  
0.7546  
Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6596 - accuracy:  
0.7380  
Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6292 - accuracy:  
0.7624  
Epoch 190/200

```

68/68 [=====] - 0s 2ms/step - loss: 0.6183 - accuracy:
0.7606
Epoch 191/200
68/68 [=====] - 0s 2ms/step - loss: 0.6379 - accuracy:
0.7463
Epoch 192/200
68/68 [=====] - 0s 2ms/step - loss: 0.6248 - accuracy:
0.7514
Epoch 193/200
68/68 [=====] - 0s 2ms/step - loss: 0.6221 - accuracy:
0.7541
Epoch 194/200
68/68 [=====] - 0s 2ms/step - loss: 0.6359 - accuracy:
0.7454
Epoch 195/200
68/68 [=====] - 0s 2ms/step - loss: 0.6465 - accuracy:
0.7449
Epoch 196/200
68/68 [=====] - 0s 3ms/step - loss: 0.5997 - accuracy:
0.7799
Epoch 197/200
68/68 [=====] - 0s 2ms/step - loss: 0.6413 - accuracy:
0.7468
Epoch 198/200
68/68 [=====] - 0s 2ms/step - loss: 0.6076 - accuracy:
0.7730
Epoch 199/200
68/68 [=====] - 0s 2ms/step - loss: 0.6369 - accuracy:
0.7463
Epoch 200/200
68/68 [=====] - 0s 2ms/step - loss: 0.6042 - accuracy:
0.7629
17/17 [=====] - 0s 2ms/step
Epoch 1/200
68/68 [=====] - 1s 2ms/step - loss: 1.7148 - accuracy:
0.3116
Epoch 2/200
68/68 [=====] - 0s 3ms/step - loss: 1.4775 - accuracy:
0.3728
Epoch 3/200
68/68 [=====] - 0s 2ms/step - loss: 1.4318 - accuracy:
0.3691
Epoch 4/200
68/68 [=====] - 0s 2ms/step - loss: 1.3953 - accuracy:
0.3884
Epoch 5/200
68/68 [=====] - 0s 2ms/step - loss: 1.3647 - accuracy:
0.4059

```



Epoch 6/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3441 - accuracy:  
0.4040  
Epoch 7/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3331 - accuracy:  
0.4156  
Epoch 8/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3106 - accuracy:  
0.4192  
Epoch 9/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3035 - accuracy:  
0.4261  
Epoch 10/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2923 - accuracy:  
0.4284  
Epoch 11/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3004 - accuracy:  
0.4294  
Epoch 12/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2764 - accuracy:  
0.4367  
Epoch 13/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2569 - accuracy:  
0.4450  
Epoch 14/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2526 - accuracy:  
0.4464  
Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2530 - accuracy:  
0.4556  
Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2502 - accuracy:  
0.4648  
Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2434 - accuracy:  
0.4671  
Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2377 - accuracy:  
0.4699  
Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2363 - accuracy:  
0.4639  
Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2107 - accuracy:  
0.4694  
Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2007 - accuracy:  
0.4814

Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2075 - accuracy: 0.4814

Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1978 - accuracy: 0.4837

Epoch 24/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1882 - accuracy: 0.4809

Epoch 25/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1884 - accuracy: 0.4942

Epoch 26/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1818 - accuracy: 0.4864

Epoch 27/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1745 - accuracy: 0.4938

Epoch 28/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1549 - accuracy: 0.4998

Epoch 29/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1496 - accuracy: 0.5053

Epoch 30/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1529 - accuracy: 0.5205

Epoch 31/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1451 - accuracy: 0.5071

Epoch 32/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1226 - accuracy: 0.5320

Epoch 33/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1394 - accuracy: 0.5223

Epoch 34/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1303 - accuracy: 0.5177

Epoch 35/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1231 - accuracy: 0.5324

Epoch 36/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1106 - accuracy: 0.5136

Epoch 37/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1066 - accuracy: 0.5389

Epoch 38/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1003 - accuracy: 0.5357

Epoch 39/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1044 - accuracy: 0.5361

Epoch 40/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1005 - accuracy: 0.5324

Epoch 41/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0944 - accuracy: 0.5255

Epoch 42/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0860 - accuracy: 0.5366

Epoch 43/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0799 - accuracy: 0.5518

Epoch 44/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0739 - accuracy: 0.5439

Epoch 45/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0747 - accuracy: 0.5430

Epoch 46/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0821 - accuracy: 0.5486

Epoch 47/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0746 - accuracy: 0.5324

Epoch 48/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0479 - accuracy: 0.5614

Epoch 49/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0471 - accuracy: 0.5536

Epoch 50/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0287 - accuracy: 0.5656

Epoch 51/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0423 - accuracy: 0.5527

Epoch 52/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0350 - accuracy: 0.5522

Epoch 53/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0430 - accuracy: 0.5490

Epoch 54/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0220 - accuracy:  
0.5780  
Epoch 55/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0302 - accuracy:  
0.5683  
Epoch 56/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0105 - accuracy:  
0.5771  
Epoch 57/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9990 - accuracy:  
0.5789  
Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0004 - accuracy:  
0.5808  
Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9869 - accuracy:  
0.5794  
Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0024 - accuracy:  
0.5817  
Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0086 - accuracy:  
0.5688  
Epoch 62/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9836 - accuracy:  
0.5849  
Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9841 - accuracy:  
0.5881  
Epoch 64/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9764 - accuracy:  
0.5955  
Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9809 - accuracy:  
0.5739  
Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9704 - accuracy:  
0.6006  
Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9765 - accuracy:  
0.5978  
Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9837 - accuracy:  
0.5840  
Epoch 69/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9589 - accuracy:  
0.5978

Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9474 - accuracy: 0.6019

Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9431 - accuracy: 0.6033

Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9318 - accuracy: 0.6157

Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9521 - accuracy: 0.5996

Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9484 - accuracy: 0.6033

Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9193 - accuracy: 0.6176

Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9283 - accuracy: 0.6029

Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9229 - accuracy: 0.6116

Epoch 78/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9183 - accuracy: 0.6231

Epoch 79/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9205 - accuracy: 0.6157

Epoch 80/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9199 - accuracy: 0.6121

Epoch 81/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9019 - accuracy: 0.6332

Epoch 82/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9123 - accuracy: 0.6185

Epoch 83/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9119 - accuracy: 0.6171

Epoch 84/200  
68/68 [=====] - 0s 3ms/step - loss: 0.9008 - accuracy: 0.6282

Epoch 85/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8944 - accuracy: 0.6236

Epoch 86/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8785 - accuracy:  
0.6503  
Epoch 87/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8800 - accuracy:  
0.6392  
Epoch 88/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8889 - accuracy:  
0.6346  
Epoch 89/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8730 - accuracy:  
0.6549  
Epoch 90/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8736 - accuracy:  
0.6475  
Epoch 91/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8656 - accuracy:  
0.6424  
Epoch 92/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8795 - accuracy:  
0.6300  
Epoch 93/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8858 - accuracy:  
0.6415  
Epoch 94/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8772 - accuracy:  
0.6461  
Epoch 95/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8306 - accuracy:  
0.6613  
Epoch 96/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8563 - accuracy:  
0.6526  
Epoch 97/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8663 - accuracy:  
0.6397  
Epoch 98/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8502 - accuracy:  
0.6539  
Epoch 99/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8485 - accuracy:  
0.6470  
Epoch 100/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8649 - accuracy:  
0.6484  
Epoch 101/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8420 - accuracy:  
0.6576

Epoch 102/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8528 - accuracy: 0.6461

Epoch 103/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8426 - accuracy: 0.6493

Epoch 104/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8097 - accuracy: 0.6733

Epoch 105/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8272 - accuracy: 0.6733

Epoch 106/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8293 - accuracy: 0.6567

Epoch 107/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8277 - accuracy: 0.6696

Epoch 108/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8106 - accuracy: 0.6710

Epoch 109/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8056 - accuracy: 0.6737

Epoch 110/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7945 - accuracy: 0.6797

Epoch 111/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8134 - accuracy: 0.6737

Epoch 112/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8026 - accuracy: 0.6728

Epoch 113/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8041 - accuracy: 0.6696

Epoch 114/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8175 - accuracy: 0.6774

Epoch 115/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7986 - accuracy: 0.6861

Epoch 116/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8144 - accuracy: 0.6756

Epoch 117/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8056 - accuracy: 0.6622

Epoch 118/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8056 - accuracy:  
0.6696  
Epoch 119/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8194 - accuracy:  
0.6723  
Epoch 120/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7971 - accuracy:  
0.6783  
Epoch 121/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8015 - accuracy:  
0.6834  
Epoch 122/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7777 - accuracy:  
0.6760  
Epoch 123/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7880 - accuracy:  
0.6838  
Epoch 124/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7984 - accuracy:  
0.6710  
Epoch 125/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7587 - accuracy:  
0.6875  
Epoch 126/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7849 - accuracy:  
0.6700  
Epoch 127/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7703 - accuracy:  
0.6866  
Epoch 128/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7603 - accuracy:  
0.6921  
Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7638 - accuracy:  
0.6986  
Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7708 - accuracy:  
0.6802  
Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7509 - accuracy:  
0.6967  
Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7606 - accuracy:  
0.6931  
Epoch 133/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7639 - accuracy:  
0.6898



Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7211 - accuracy:  
0.7064  
Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7688 - accuracy:  
0.6917  
Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7291 - accuracy:  
0.7073  
Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7608 - accuracy:  
0.6958  
Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7542 - accuracy:  
0.7069  
Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7432 - accuracy:  
0.6954  
Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7299 - accuracy:  
0.7096  
Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7270 - accuracy:  
0.7087  
Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7148 - accuracy:  
0.7119  
Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7333 - accuracy:  
0.7101  
Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7323 - accuracy:  
0.7082  
Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7343 - accuracy:  
0.7000  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7096 - accuracy:  
0.7055  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7212 - accuracy:  
0.7151  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7045 - accuracy:  
0.7207  
Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7232 - accuracy:  
0.7119

Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7230 - accuracy: 0.7142

Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7001 - accuracy: 0.7179

Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7135 - accuracy: 0.7184

Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7468 - accuracy: 0.7046

Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6823 - accuracy: 0.7271

Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6995 - accuracy: 0.7289

Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7136 - accuracy: 0.7142

Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6788 - accuracy: 0.7211

Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6969 - accuracy: 0.7257

Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6843 - accuracy: 0.7202

Epoch 160/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6728 - accuracy: 0.7312

Epoch 161/200  
68/68 [=====] - 0s 3ms/step - loss: 0.6639 - accuracy: 0.7368

Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6681 - accuracy: 0.7428

Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6932 - accuracy: 0.7202

Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6859 - accuracy: 0.7266

Epoch 165/200  
68/68 [=====] - 0s 3ms/step - loss: 0.6779 - accuracy: 0.7349

Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6904 - accuracy:  
0.7197  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6524 - accuracy:  
0.7451  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6688 - accuracy:  
0.7335  
Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6725 - accuracy:  
0.7335  
Epoch 170/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7027 - accuracy:  
0.7174  
Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6758 - accuracy:  
0.7340  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6726 - accuracy:  
0.7345  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6571 - accuracy:  
0.7340  
Epoch 174/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6725 - accuracy:  
0.7317  
Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6629 - accuracy:  
0.7497  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6750 - accuracy:  
0.7372  
Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6766 - accuracy:  
0.7358  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6364 - accuracy:  
0.7598  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6843 - accuracy:  
0.7230  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6474 - accuracy:  
0.7469  
Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6468 - accuracy:  
0.7317

Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6665 - accuracy: 0.7294

Epoch 183/200  
68/68 [=====] - 0s 3ms/step - loss: 0.6589 - accuracy: 0.7372

Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6360 - accuracy: 0.7441

Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6666 - accuracy: 0.7395

Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6583 - accuracy: 0.7418

Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6235 - accuracy: 0.7570

Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6207 - accuracy: 0.7644

Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6322 - accuracy: 0.7538

Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6383 - accuracy: 0.7556

Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6215 - accuracy: 0.7662

Epoch 192/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6375 - accuracy: 0.7533

Epoch 193/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6201 - accuracy: 0.7437

Epoch 194/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6280 - accuracy: 0.7575

Epoch 195/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6342 - accuracy: 0.7556

Epoch 196/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6320 - accuracy: 0.7432

Epoch 197/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6250 - accuracy: 0.7736

Epoch 198/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6343 - accuracy:  
0.7432  
Epoch 199/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6095 - accuracy:  
0.7570  
Epoch 200/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6181 - accuracy:  
0.7561  
17/17 [=====] - 0s 2ms/step  
Epoch 1/200  
68/68 [=====] - 1s 3ms/step - loss: 1.6902 - accuracy:  
0.3148  
Epoch 2/200  
68/68 [=====] - 0s 2ms/step - loss: 1.4637 - accuracy:  
0.3801  
Epoch 3/200  
68/68 [=====] - 0s 2ms/step - loss: 1.4108 - accuracy:  
0.3847  
Epoch 4/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3727 - accuracy:  
0.3999  
Epoch 5/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3557 - accuracy:  
0.4087  
Epoch 6/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3363 - accuracy:  
0.4160  
Epoch 7/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3202 - accuracy:  
0.4271  
Epoch 8/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3211 - accuracy:  
0.4330  
Epoch 9/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3171 - accuracy:  
0.4266  
Epoch 10/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2981 - accuracy:  
0.4395  
Epoch 11/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2897 - accuracy:  
0.4413  
Epoch 12/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2784 - accuracy:  
0.4510  
Epoch 13/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2735 - accuracy:

0.4491  
Epoch 14/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2478 - accuracy:  
0.4510  
Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2482 - accuracy:  
0.4749  
Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2404 - accuracy:  
0.4487  
Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2478 - accuracy:  
0.4565  
Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2287 - accuracy:  
0.4768  
Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2180 - accuracy:  
0.4795  
Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2089 - accuracy:  
0.4749  
Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2165 - accuracy:  
0.4758  
Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2170 - accuracy:  
0.4846  
Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1919 - accuracy:  
0.5002  
Epoch 24/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1807 - accuracy:  
0.5085  
Epoch 25/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1919 - accuracy:  
0.4910  
Epoch 26/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1800 - accuracy:  
0.5071  
Epoch 27/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1765 - accuracy:  
0.5044  
Epoch 28/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1780 - accuracy:  
0.4938  
Epoch 29/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1560 - accuracy:

0.5062  
 Epoch 30/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1522 - accuracy:  
 0.4993  
 Epoch 31/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1501 - accuracy:  
 0.5039  
 Epoch 32/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1538 - accuracy:  
 0.5081  
 Epoch 33/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1400 - accuracy:  
 0.5113  
 Epoch 34/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1379 - accuracy:  
 0.5283  
 Epoch 35/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1240 - accuracy:  
 0.5283  
 Epoch 36/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1123 - accuracy:  
 0.5398  
 Epoch 37/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1174 - accuracy:  
 0.5389  
 Epoch 38/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0974 - accuracy:  
 0.5334  
 Epoch 39/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1197 - accuracy:  
 0.5214  
 Epoch 40/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1088 - accuracy:  
 0.5297  
 Epoch 41/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.1013 - accuracy:  
 0.5476  
 Epoch 42/200  
 68/68 [=====] - 0s 3ms/step - loss: 1.1048 - accuracy:  
 0.5389  
 Epoch 43/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0927 - accuracy:  
 0.5338  
 Epoch 44/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0746 - accuracy:  
 0.5458  
 Epoch 45/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.0719 - accuracy:

```

0.5573
Epoch 46/200
68/68 [=====] - 0s 2ms/step - loss: 1.0784 - accuracy:
0.5426
Epoch 47/200
68/68 [=====] - 0s 2ms/step - loss: 1.0503 - accuracy:
0.5541
Epoch 48/200
68/68 [=====] - 0s 2ms/step - loss: 1.0617 - accuracy:
0.5582
Epoch 49/200
68/68 [=====] - 0s 2ms/step - loss: 1.0608 - accuracy:
0.5545
Epoch 50/200
68/68 [=====] - 0s 2ms/step - loss: 1.0528 - accuracy:
0.5559
Epoch 51/200
68/68 [=====] - 0s 2ms/step - loss: 1.0459 - accuracy:
0.5706
Epoch 52/200
68/68 [=====] - 0s 2ms/step - loss: 1.0277 - accuracy:
0.5720
Epoch 53/200
68/68 [=====] - 0s 3ms/step - loss: 1.0549 - accuracy:
0.5555
Epoch 54/200
68/68 [=====] - 0s 3ms/step - loss: 1.0285 - accuracy:
0.5716
Epoch 55/200
68/68 [=====] - 0s 2ms/step - loss: 1.0356 - accuracy:
0.5651
Epoch 56/200
68/68 [=====] - 0s 2ms/step - loss: 1.0160 - accuracy:
0.5794
Epoch 57/200
68/68 [=====] - 0s 2ms/step - loss: 1.0190 - accuracy:
0.5748
Epoch 58/200
68/68 [=====] - 0s 2ms/step - loss: 1.0090 - accuracy:
0.5803
Epoch 59/200
68/68 [=====] - 0s 2ms/step - loss: 1.0303 - accuracy:
0.5716
Epoch 60/200
68/68 [=====] - 0s 2ms/step - loss: 1.0047 - accuracy:
0.5771
Epoch 61/200
68/68 [=====] - 0s 2ms/step - loss: 0.9868 - accuracy:

```



0.5757  
Epoch 62/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0013 - accuracy: 0.5739  
Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9785 - accuracy: 0.5987  
Epoch 64/200  
68/68 [=====] - 0s 3ms/step - loss: 0.9797 - accuracy: 0.5987  
Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9968 - accuracy: 0.5895  
Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9719 - accuracy: 0.5987  
Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9736 - accuracy: 0.6061  
Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9666 - accuracy: 0.5904  
Epoch 69/200  
68/68 [=====] - 0s 3ms/step - loss: 0.9905 - accuracy: 0.5808  
Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9523 - accuracy: 0.6052  
Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9373 - accuracy: 0.6185  
Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9482 - accuracy: 0.6107  
Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9339 - accuracy: 0.6295  
Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9386 - accuracy: 0.6194  
Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9249 - accuracy: 0.6295  
Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9296 - accuracy: 0.6208  
Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9400 - accuracy:

```

0.6029
Epoch 78/200
68/68 [=====] - 0s 2ms/step - loss: 0.9312 - accuracy:
0.6263
Epoch 79/200
68/68 [=====] - 0s 2ms/step - loss: 0.9286 - accuracy:
0.6226
Epoch 80/200
68/68 [=====] - 0s 2ms/step - loss: 0.9255 - accuracy:
0.6162
Epoch 81/200
68/68 [=====] - 0s 2ms/step - loss: 0.9064 - accuracy:
0.6300
Epoch 82/200
68/68 [=====] - 0s 2ms/step - loss: 0.9109 - accuracy:
0.6364
Epoch 83/200
68/68 [=====] - 0s 2ms/step - loss: 0.9054 - accuracy:
0.6341
Epoch 84/200
68/68 [=====] - 0s 2ms/step - loss: 0.8885 - accuracy:
0.6438
Epoch 85/200
68/68 [=====] - 0s 2ms/step - loss: 0.8977 - accuracy:
0.6217
Epoch 86/200
68/68 [=====] - 0s 2ms/step - loss: 0.8855 - accuracy:
0.6291
Epoch 87/200
68/68 [=====] - 0s 2ms/step - loss: 0.9086 - accuracy:
0.6249
Epoch 88/200
68/68 [=====] - 0s 2ms/step - loss: 0.9087 - accuracy:
0.6291
Epoch 89/200
68/68 [=====] - 0s 2ms/step - loss: 0.8793 - accuracy:
0.6590
Epoch 90/200
68/68 [=====] - 0s 2ms/step - loss: 0.8771 - accuracy:
0.6438
Epoch 91/200
68/68 [=====] - 0s 2ms/step - loss: 0.8834 - accuracy:
0.6493
Epoch 92/200
68/68 [=====] - 0s 2ms/step - loss: 0.8662 - accuracy:
0.6457
Epoch 93/200
68/68 [=====] - 0s 2ms/step - loss: 0.8701 - accuracy:

```

0.6489  
Epoch 94/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8522 - accuracy:  
0.6535  
Epoch 95/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8573 - accuracy:  
0.6604  
Epoch 96/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8552 - accuracy:  
0.6461  
Epoch 97/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8586 - accuracy:  
0.6553  
Epoch 98/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8438 - accuracy:  
0.6461  
Epoch 99/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8365 - accuracy:  
0.6636  
Epoch 100/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8531 - accuracy:  
0.6604  
Epoch 101/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8422 - accuracy:  
0.6434  
Epoch 102/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8595 - accuracy:  
0.6434  
Epoch 103/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8284 - accuracy:  
0.6590  
Epoch 104/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8334 - accuracy:  
0.6664  
Epoch 105/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8394 - accuracy:  
0.6503  
Epoch 106/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8218 - accuracy:  
0.6636  
Epoch 107/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8034 - accuracy:  
0.6811  
Epoch 108/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8031 - accuracy:  
0.6691  
Epoch 109/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8281 - accuracy:

0.6664  
 Epoch 110/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.8139 - accuracy:  
 0.6673  
 Epoch 111/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7974 - accuracy:  
 0.6636  
 Epoch 112/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.8066 - accuracy:  
 0.6779  
 Epoch 113/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.8143 - accuracy:  
 0.6710  
 Epoch 114/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7957 - accuracy:  
 0.6760  
 Epoch 115/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7914 - accuracy:  
 0.6783  
 Epoch 116/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.8146 - accuracy:  
 0.6687  
 Epoch 117/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7924 - accuracy:  
 0.6668  
 Epoch 118/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7805 - accuracy:  
 0.6931  
 Epoch 119/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7839 - accuracy:  
 0.6931  
 Epoch 120/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7773 - accuracy:  
 0.6889  
 Epoch 121/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7762 - accuracy:  
 0.6825  
 Epoch 122/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7607 - accuracy:  
 0.7004  
 Epoch 123/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7601 - accuracy:  
 0.6981  
 Epoch 124/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7631 - accuracy:  
 0.7078  
 Epoch 125/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7619 - accuracy:

0.7027  
 Epoch 126/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7750 - accuracy:  
 0.6908  
 Epoch 127/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7632 - accuracy:  
 0.7036  
 Epoch 128/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7747 - accuracy:  
 0.6825  
 Epoch 129/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7664 - accuracy:  
 0.7036  
 Epoch 130/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7526 - accuracy:  
 0.7046  
 Epoch 131/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7524 - accuracy:  
 0.6958  
 Epoch 132/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7430 - accuracy:  
 0.7105  
 Epoch 133/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7227 - accuracy:  
 0.7138  
 Epoch 134/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7451 - accuracy:  
 0.7059  
 Epoch 135/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7270 - accuracy:  
 0.6981  
 Epoch 136/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7443 - accuracy:  
 0.6972  
 Epoch 137/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7349 - accuracy:  
 0.7082  
 Epoch 138/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7386 - accuracy:  
 0.7147  
 Epoch 139/200  
 68/68 [=====] - 0s 3ms/step - loss: 0.6984 - accuracy:  
 0.7184  
 Epoch 140/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7136 - accuracy:  
 0.7197  
 Epoch 141/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7241 - accuracy:

0.7105  
 Epoch 142/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7019 - accuracy:  
 0.7207  
 Epoch 143/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7076 - accuracy:  
 0.7133  
 Epoch 144/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7311 - accuracy:  
 0.7036  
 Epoch 145/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7385 - accuracy:  
 0.7073  
 Epoch 146/200  
 68/68 [=====] - 0s 3ms/step - loss: 0.7202 - accuracy:  
 0.7142  
 Epoch 147/200  
 68/68 [=====] - 0s 3ms/step - loss: 0.7043 - accuracy:  
 0.7184  
 Epoch 148/200  
 68/68 [=====] - 0s 3ms/step - loss: 0.7013 - accuracy:  
 0.7064  
 Epoch 149/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7246 - accuracy:  
 0.7023  
 Epoch 150/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7131 - accuracy:  
 0.7184  
 Epoch 151/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6890 - accuracy:  
 0.7308  
 Epoch 152/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6786 - accuracy:  
 0.7317  
 Epoch 153/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6890 - accuracy:  
 0.7230  
 Epoch 154/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.7116 - accuracy:  
 0.7096  
 Epoch 155/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6956 - accuracy:  
 0.7220  
 Epoch 156/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6693 - accuracy:  
 0.7437  
 Epoch 157/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6945 - accuracy:

0.7211  
Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6939 - accuracy:  
0.7391  
Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6820 - accuracy:  
0.7331  
Epoch 160/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6747 - accuracy:  
0.7239  
Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6747 - accuracy:  
0.7294  
Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6747 - accuracy:  
0.7455  
Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6838 - accuracy:  
0.7184  
Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6693 - accuracy:  
0.7349  
Epoch 165/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6619 - accuracy:  
0.7469  
Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6663 - accuracy:  
0.7372  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6881 - accuracy:  
0.7193  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6736 - accuracy:  
0.7418  
Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6453 - accuracy:  
0.7400  
Epoch 170/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6414 - accuracy:  
0.7506  
Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6619 - accuracy:  
0.7391  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6549 - accuracy:  
0.7501  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6492 - accuracy:

0.7561  
 Epoch 174/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6623 - accuracy:  
 0.7345  
 Epoch 175/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6531 - accuracy:  
 0.7395  
 Epoch 176/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6404 - accuracy:  
 0.7520  
 Epoch 177/200  
 68/68 [=====] - 0s 3ms/step - loss: 0.6290 - accuracy:  
 0.7529  
 Epoch 178/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6274 - accuracy:  
 0.7556  
 Epoch 179/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6348 - accuracy:  
 0.7529  
 Epoch 180/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6233 - accuracy:  
 0.7561  
 Epoch 181/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6337 - accuracy:  
 0.7437  
 Epoch 182/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6515 - accuracy:  
 0.7414  
 Epoch 183/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6327 - accuracy:  
 0.7469  
 Epoch 184/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6213 - accuracy:  
 0.7543  
 Epoch 185/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6319 - accuracy:  
 0.7487  
 Epoch 186/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6405 - accuracy:  
 0.7483  
 Epoch 187/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6444 - accuracy:  
 0.7529  
 Epoch 188/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6421 - accuracy:  
 0.7538  
 Epoch 189/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6084 - accuracy:



0.7538  
 Epoch 190/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6154 - accuracy:  
 0.7570  
 Epoch 191/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6143 - accuracy:  
 0.7575  
 Epoch 192/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6213 - accuracy:  
 0.7556  
 Epoch 193/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6133 - accuracy:  
 0.7520  
 Epoch 194/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6175 - accuracy:  
 0.7625  
 Epoch 195/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6020 - accuracy:  
 0.7625  
 Epoch 196/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.5948 - accuracy:  
 0.7685  
 Epoch 197/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6064 - accuracy:  
 0.7579  
 Epoch 198/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6279 - accuracy:  
 0.7487  
 Epoch 199/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6292 - accuracy:  
 0.7547  
 Epoch 200/200  
 68/68 [=====] - 0s 2ms/step - loss: 0.6172 - accuracy:  
 0.7616  
 17/17 [=====] - 0s 1ms/step  
 Epoch 1/200  
 68/68 [=====] - 1s 2ms/step - loss: 1.7072 - accuracy:  
 0.2968  
 Epoch 2/200  
 68/68 [=====] - 0s 3ms/step - loss: 1.4675 - accuracy:  
 0.3654  
 Epoch 3/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.4087 - accuracy:  
 0.3935  
 Epoch 4/200  
 68/68 [=====] - 0s 2ms/step - loss: 1.3886 - accuracy:  
 0.4054  
 Epoch 5/200

68/68 [=====] - 0s 2ms/step - loss: 1.3485 - accuracy:  
0.4188  
Epoch 6/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3394 - accuracy:  
0.4169  
Epoch 7/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3204 - accuracy:  
0.4307  
Epoch 8/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3216 - accuracy:  
0.4110  
Epoch 9/200  
68/68 [=====] - 0s 2ms/step - loss: 1.3045 - accuracy:  
0.4330  
Epoch 10/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2958 - accuracy:  
0.4413  
Epoch 11/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2773 - accuracy:  
0.4381  
Epoch 12/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2763 - accuracy:  
0.4450  
Epoch 13/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2636 - accuracy:  
0.4514  
Epoch 14/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2514 - accuracy:  
0.4574  
Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2502 - accuracy:  
0.4524  
Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2459 - accuracy:  
0.4685  
Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2344 - accuracy:  
0.4666  
Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2369 - accuracy:  
0.4643  
Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2185 - accuracy:  
0.4823  
Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2134 - accuracy:  
0.4827  
Epoch 21/200

68/68 [=====] - 0s 3ms/step - loss: 1.2085 - accuracy:  
0.4938  
Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1988 - accuracy:  
0.4850  
Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1963 - accuracy:  
0.4841  
Epoch 24/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1905 - accuracy:  
0.5012  
Epoch 25/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1853 - accuracy:  
0.4869  
Epoch 26/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1761 - accuracy:  
0.4984  
Epoch 27/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1696 - accuracy:  
0.5048  
Epoch 28/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1663 - accuracy:  
0.4988  
Epoch 29/200  
68/68 [=====] - 0s 3ms/step - loss: 1.1592 - accuracy:  
0.5030  
Epoch 30/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1446 - accuracy:  
0.5113  
Epoch 31/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1585 - accuracy:  
0.5067  
Epoch 32/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1325 - accuracy:  
0.5145  
Epoch 33/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1424 - accuracy:  
0.5154  
Epoch 34/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1361 - accuracy:  
0.5269  
Epoch 35/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1319 - accuracy:  
0.5196  
Epoch 36/200  
68/68 [=====] - 0s 3ms/step - loss: 1.1170 - accuracy:  
0.5255  
Epoch 37/200

68/68 [=====] - 0s 2ms/step - loss: 1.1161 - accuracy: 0.5246  
Epoch 38/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1223 - accuracy: 0.5200  
Epoch 39/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1091 - accuracy: 0.5407  
Epoch 40/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0909 - accuracy: 0.5430  
Epoch 41/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0926 - accuracy: 0.5338  
Epoch 42/200  
68/68 [=====] - 0s 3ms/step - loss: 1.0846 - accuracy: 0.5536  
Epoch 43/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0768 - accuracy: 0.5430  
Epoch 44/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0820 - accuracy: 0.5393  
Epoch 45/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0705 - accuracy: 0.5518  
Epoch 46/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0591 - accuracy: 0.5421  
Epoch 47/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0619 - accuracy: 0.5550  
Epoch 48/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0680 - accuracy: 0.5495  
Epoch 49/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0482 - accuracy: 0.5582  
Epoch 50/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0464 - accuracy: 0.5601  
Epoch 51/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0391 - accuracy: 0.5628  
Epoch 52/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0417 - accuracy: 0.5564  
Epoch 53/200

68/68 [=====] - 0s 2ms/step - loss: 1.0221 - accuracy:  
0.5679  
Epoch 54/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0314 - accuracy:  
0.5789  
Epoch 55/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0323 - accuracy:  
0.5679  
Epoch 56/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0234 - accuracy:  
0.5633  
Epoch 57/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0202 - accuracy:  
0.5835  
Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0165 - accuracy:  
0.5821  
Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0254 - accuracy:  
0.5739  
Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9918 - accuracy:  
0.6029  
Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9934 - accuracy:  
0.6033  
Epoch 62/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9965 - accuracy:  
0.5886  
Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9815 - accuracy:  
0.5895  
Epoch 64/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0026 - accuracy:  
0.5780  
Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9846 - accuracy:  
0.5923  
Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9724 - accuracy:  
0.5946  
Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9851 - accuracy:  
0.5955  
Epoch 68/200  
68/68 [=====] - 0s 3ms/step - loss: 0.9531 - accuracy:  
0.6033  
Epoch 69/200

68/68 [=====] - 0s 2ms/step - loss: 0.9769 - accuracy:  
0.5955  
Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9697 - accuracy:  
0.6056  
Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9419 - accuracy:  
0.6088  
Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9525 - accuracy:  
0.6102  
Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9418 - accuracy:  
0.6162  
Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9406 - accuracy:  
0.6139  
Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9390 - accuracy:  
0.6084  
Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9330 - accuracy:  
0.6093  
Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9195 - accuracy:  
0.6180  
Epoch 78/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9128 - accuracy:  
0.6309  
Epoch 79/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9308 - accuracy:  
0.6203  
Epoch 80/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8987 - accuracy:  
0.6300  
Epoch 81/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9130 - accuracy:  
0.6194  
Epoch 82/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9237 - accuracy:  
0.6222  
Epoch 83/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9142 - accuracy:  
0.6180  
Epoch 84/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8957 - accuracy:  
0.6314  
Epoch 85/200

68/68 [=====] - 0s 2ms/step - loss: 0.8995 - accuracy:  
0.6337  
Epoch 86/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9058 - accuracy:  
0.6272  
Epoch 87/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8842 - accuracy:  
0.6475  
Epoch 88/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8836 - accuracy:  
0.6489  
Epoch 89/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8809 - accuracy:  
0.6443  
Epoch 90/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8672 - accuracy:  
0.6558  
Epoch 91/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8791 - accuracy:  
0.6378  
Epoch 92/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8641 - accuracy:  
0.6549  
Epoch 93/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8594 - accuracy:  
0.6475  
Epoch 94/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8677 - accuracy:  
0.6572  
Epoch 95/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8670 - accuracy:  
0.6475  
Epoch 96/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8529 - accuracy:  
0.6521  
Epoch 97/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8677 - accuracy:  
0.6420  
Epoch 98/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8491 - accuracy:  
0.6654  
Epoch 99/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8322 - accuracy:  
0.6733  
Epoch 100/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8267 - accuracy:  
0.6742  
Epoch 101/200

68/68 [=====] - 0s 2ms/step - loss: 0.8302 - accuracy:  
0.6728  
Epoch 102/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8431 - accuracy:  
0.6613  
Epoch 103/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8163 - accuracy:  
0.6811  
Epoch 104/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8575 - accuracy:  
0.6581  
Epoch 105/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8326 - accuracy:  
0.6581  
Epoch 106/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8158 - accuracy:  
0.6792  
Epoch 107/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8187 - accuracy:  
0.6792  
Epoch 108/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8229 - accuracy:  
0.6723  
Epoch 109/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8184 - accuracy:  
0.6728  
Epoch 110/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8165 - accuracy:  
0.6760  
Epoch 111/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8300 - accuracy:  
0.6576  
Epoch 112/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8105 - accuracy:  
0.6737  
Epoch 113/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8063 - accuracy:  
0.6756  
Epoch 114/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7907 - accuracy:  
0.6880  
Epoch 115/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7994 - accuracy:  
0.6861  
Epoch 116/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8027 - accuracy:  
0.6677  
Epoch 117/200



68/68 [=====] - 0s 2ms/step - loss: 0.7862 - accuracy:  
0.6779  
Epoch 118/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7872 - accuracy:  
0.6779  
Epoch 119/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7880 - accuracy:  
0.6829  
Epoch 120/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7618 - accuracy:  
0.6884  
Epoch 121/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7698 - accuracy:  
0.6908  
Epoch 122/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7704 - accuracy:  
0.6912  
Epoch 123/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7783 - accuracy:  
0.6769  
Epoch 124/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7714 - accuracy:  
0.6857  
Epoch 125/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7618 - accuracy:  
0.6917  
Epoch 126/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7725 - accuracy:  
0.6972  
Epoch 127/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7906 - accuracy:  
0.6995  
Epoch 128/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7560 - accuracy:  
0.6940  
Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7552 - accuracy:  
0.6986  
Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7515 - accuracy:  
0.6986  
Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7453 - accuracy:  
0.7124  
Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7544 - accuracy:  
0.6967  
Epoch 133/200

68/68 [=====] - 0s 3ms/step - loss: 0.7405 - accuracy:  
0.7101  
Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7604 - accuracy:  
0.7092  
Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7402 - accuracy:  
0.7041  
Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7347 - accuracy:  
0.6926  
Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7260 - accuracy:  
0.7170  
Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7323 - accuracy:  
0.7188  
Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7289 - accuracy:  
0.7013  
Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7307 - accuracy:  
0.7128  
Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7188 - accuracy:  
0.7179  
Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7113 - accuracy:  
0.7128  
Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7183 - accuracy:  
0.7161  
Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7030 - accuracy:  
0.7188  
Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7261 - accuracy:  
0.7207  
Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7084 - accuracy:  
0.7156  
Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7017 - accuracy:  
0.7303  
Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7177 - accuracy:  
0.7138  
Epoch 149/200

68/68 [=====] - 0s 2ms/step - loss: 0.7058 - accuracy:  
0.7216  
Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7056 - accuracy:  
0.7197  
Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7253 - accuracy:  
0.7202  
Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6869 - accuracy:  
0.7280  
Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7196 - accuracy:  
0.7119  
Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6790 - accuracy:  
0.7289  
Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6997 - accuracy:  
0.7239  
Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6939 - accuracy:  
0.7257  
Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6896 - accuracy:  
0.7285  
Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6754 - accuracy:  
0.7312  
Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6599 - accuracy:  
0.7391  
Epoch 160/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6742 - accuracy:  
0.7289  
Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6838 - accuracy:  
0.7262  
Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6718 - accuracy:  
0.7391  
Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6980 - accuracy:  
0.7303  
Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6867 - accuracy:  
0.7386  
Epoch 165/200

68/68 [=====] - 0s 2ms/step - loss: 0.6714 - accuracy:  
0.7340  
Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6682 - accuracy:  
0.7294  
Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6899 - accuracy:  
0.7147  
Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6704 - accuracy:  
0.7368  
Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6597 - accuracy:  
0.7409  
Epoch 170/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6704 - accuracy:  
0.7487  
Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6662 - accuracy:  
0.7515  
Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6507 - accuracy:  
0.7395  
Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6422 - accuracy:  
0.7497  
Epoch 174/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6459 - accuracy:  
0.7487  
Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6476 - accuracy:  
0.7395  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6753 - accuracy:  
0.7400  
Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6543 - accuracy:  
0.7299  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6286 - accuracy:  
0.7538  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6414 - accuracy:  
0.7437  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6601 - accuracy:  
0.7497  
Epoch 181/200

68/68 [=====] - 0s 2ms/step - loss: 0.6419 - accuracy:  
0.7483  
Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6384 - accuracy:  
0.7561  
Epoch 183/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6472 - accuracy:  
0.7432  
Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6376 - accuracy:  
0.7520  
Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6201 - accuracy:  
0.7593  
Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6384 - accuracy:  
0.7437  
Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6561 - accuracy:  
0.7349  
Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6247 - accuracy:  
0.7552  
Epoch 189/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6290 - accuracy:  
0.7501  
Epoch 190/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6108 - accuracy:  
0.7630  
Epoch 191/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6244 - accuracy:  
0.7579  
Epoch 192/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6274 - accuracy:  
0.7515  
Epoch 193/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6135 - accuracy:  
0.7713  
Epoch 194/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6030 - accuracy:  
0.7690  
Epoch 195/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6340 - accuracy:  
0.7409  
Epoch 196/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6171 - accuracy:  
0.7533  
Epoch 197/200

```

68/68 [=====] - 0s 2ms/step - loss: 0.6239 - accuracy:
0.7598
Epoch 198/200
68/68 [=====] - 0s 2ms/step - loss: 0.6081 - accuracy:
0.7658
Epoch 199/200
68/68 [=====] - 0s 2ms/step - loss: 0.6203 - accuracy:
0.7547
Epoch 200/200
68/68 [=====] - 0s 2ms/step - loss: 0.6097 - accuracy:
0.7653
17/17 [=====] - 0s 1ms/step
Epoch 1/200
68/68 [=====] - 1s 2ms/step - loss: 1.7134 - accuracy:
0.3125
Epoch 2/200
68/68 [=====] - 0s 2ms/step - loss: 1.4891 - accuracy:
0.3783
Epoch 3/200
68/68 [=====] - 0s 2ms/step - loss: 1.4291 - accuracy:
0.3797
Epoch 4/200
68/68 [=====] - 0s 2ms/step - loss: 1.3943 - accuracy:
0.3962
Epoch 5/200
68/68 [=====] - 0s 3ms/step - loss: 1.3666 - accuracy:
0.4073
Epoch 6/200
68/68 [=====] - 0s 2ms/step - loss: 1.3443 - accuracy:
0.4114
Epoch 7/200
68/68 [=====] - 0s 2ms/step - loss: 1.3429 - accuracy:
0.4119
Epoch 8/200
68/68 [=====] - 0s 2ms/step - loss: 1.3142 - accuracy:
0.4289
Epoch 9/200
68/68 [=====] - 0s 2ms/step - loss: 1.3004 - accuracy:
0.4344
Epoch 10/200
68/68 [=====] - 0s 2ms/step - loss: 1.2955 - accuracy:
0.4321
Epoch 11/200
68/68 [=====] - 0s 2ms/step - loss: 1.2951 - accuracy:
0.4344
Epoch 12/200
68/68 [=====] - 0s 3ms/step - loss: 1.2774 - accuracy:
0.4570

```

Epoch 13/200  
68/68 [=====] - 0s 3ms/step - loss: 1.2720 - accuracy:  
0.4344  
Epoch 14/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2647 - accuracy:  
0.4538  
Epoch 15/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2485 - accuracy:  
0.4561  
Epoch 16/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2501 - accuracy:  
0.4561  
Epoch 17/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2457 - accuracy:  
0.4689  
Epoch 18/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2469 - accuracy:  
0.4625  
Epoch 19/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2439 - accuracy:  
0.4611  
Epoch 20/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2248 - accuracy:  
0.4754  
Epoch 21/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2130 - accuracy:  
0.4754  
Epoch 22/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2102 - accuracy:  
0.4648  
Epoch 23/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2031 - accuracy:  
0.4735  
Epoch 24/200  
68/68 [=====] - 0s 2ms/step - loss: 1.2045 - accuracy:  
0.4883  
Epoch 25/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1864 - accuracy:  
0.4993  
Epoch 26/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1739 - accuracy:  
0.5113  
Epoch 27/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1858 - accuracy:  
0.5012  
Epoch 28/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1679 - accuracy:  
0.4988

Epoch 29/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1584 - accuracy:  
0.5012  
Epoch 30/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1623 - accuracy:  
0.5007  
Epoch 31/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1539 - accuracy:  
0.5085  
Epoch 32/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1574 - accuracy:  
0.5085  
Epoch 33/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1318 - accuracy:  
0.5145  
Epoch 34/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1395 - accuracy:  
0.5062  
Epoch 35/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1241 - accuracy:  
0.5186  
Epoch 36/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1030 - accuracy:  
0.5407  
Epoch 37/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1127 - accuracy:  
0.5269  
Epoch 38/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1123 - accuracy:  
0.5297  
Epoch 39/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0973 - accuracy:  
0.5306  
Epoch 40/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0989 - accuracy:  
0.5366  
Epoch 41/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1018 - accuracy:  
0.5315  
Epoch 42/200  
68/68 [=====] - 0s 2ms/step - loss: 1.1031 - accuracy:  
0.5384  
Epoch 43/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0896 - accuracy:  
0.5421  
Epoch 44/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0865 - accuracy:  
0.5486



Epoch 45/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0678 - accuracy:  
0.5582  
Epoch 46/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0583 - accuracy:  
0.5536  
Epoch 47/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0572 - accuracy:  
0.5499  
Epoch 48/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0515 - accuracy:  
0.5527  
Epoch 49/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0510 - accuracy:  
0.5587  
Epoch 50/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0556 - accuracy:  
0.5509  
Epoch 51/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0594 - accuracy:  
0.5587  
Epoch 52/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0255 - accuracy:  
0.5660  
Epoch 53/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0178 - accuracy:  
0.5835  
Epoch 54/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0105 - accuracy:  
0.5757  
Epoch 55/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0248 - accuracy:  
0.5734  
Epoch 56/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0069 - accuracy:  
0.5748  
Epoch 57/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9964 - accuracy:  
0.5785  
Epoch 58/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0163 - accuracy:  
0.5734  
Epoch 59/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0046 - accuracy:  
0.5808  
Epoch 60/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9936 - accuracy:  
0.5798

Epoch 61/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9902 - accuracy: 0.5817

Epoch 62/200  
68/68 [=====] - 0s 2ms/step - loss: 1.0042 - accuracy: 0.5771

Epoch 63/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9858 - accuracy: 0.5808

Epoch 64/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9785 - accuracy: 0.5923

Epoch 65/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9768 - accuracy: 0.5932

Epoch 66/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9648 - accuracy: 0.6010

Epoch 67/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9635 - accuracy: 0.5913

Epoch 68/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9688 - accuracy: 0.5932

Epoch 69/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9517 - accuracy: 0.6098

Epoch 70/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9613 - accuracy: 0.5973

Epoch 71/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9422 - accuracy: 0.6029

Epoch 72/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9487 - accuracy: 0.6024

Epoch 73/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9321 - accuracy: 0.6075

Epoch 74/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9470 - accuracy: 0.6148

Epoch 75/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9298 - accuracy: 0.6240

Epoch 76/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9320 - accuracy: 0.6102

Epoch 77/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9373 - accuracy: 0.6176  
Epoch 78/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9244 - accuracy: 0.6263  
Epoch 79/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9122 - accuracy: 0.6208  
Epoch 80/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9262 - accuracy: 0.6065  
Epoch 81/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9149 - accuracy: 0.6190  
Epoch 82/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8938 - accuracy: 0.6397  
Epoch 83/200  
68/68 [=====] - 0s 2ms/step - loss: 0.9195 - accuracy: 0.6125  
Epoch 84/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8957 - accuracy: 0.6240  
Epoch 85/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8747 - accuracy: 0.6410  
Epoch 86/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8750 - accuracy: 0.6337  
Epoch 87/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8689 - accuracy: 0.6424  
Epoch 88/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8786 - accuracy: 0.6378  
Epoch 89/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8684 - accuracy: 0.6507  
Epoch 90/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8808 - accuracy: 0.6355  
Epoch 91/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8861 - accuracy: 0.6222  
Epoch 92/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8740 - accuracy: 0.6452

Epoch 93/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8710 - accuracy: 0.6484  
Epoch 94/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8367 - accuracy: 0.6746  
Epoch 95/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8656 - accuracy: 0.6410  
Epoch 96/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8447 - accuracy: 0.6604  
Epoch 97/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8542 - accuracy: 0.6507  
Epoch 98/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8457 - accuracy: 0.6590  
Epoch 99/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8246 - accuracy: 0.6599  
Epoch 100/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8422 - accuracy: 0.6544  
Epoch 101/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8293 - accuracy: 0.6627  
Epoch 102/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8151 - accuracy: 0.6723  
Epoch 103/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8063 - accuracy: 0.6792  
Epoch 104/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8355 - accuracy: 0.6544  
Epoch 105/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8096 - accuracy: 0.6645  
Epoch 106/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8300 - accuracy: 0.6604  
Epoch 107/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8003 - accuracy: 0.6723  
Epoch 108/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8242 - accuracy: 0.6756

Epoch 109/200  
68/68 [=====] - 0s 3ms/step - loss: 0.8091 - accuracy:  
0.6788  
Epoch 110/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8087 - accuracy:  
0.6746  
Epoch 111/200  
68/68 [=====] - 0s 2ms/step - loss: 0.8103 - accuracy:  
0.6774  
Epoch 112/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7984 - accuracy:  
0.6751  
Epoch 113/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7995 - accuracy:  
0.6792  
Epoch 114/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7780 - accuracy:  
0.6884  
Epoch 115/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7677 - accuracy:  
0.6829  
Epoch 116/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7847 - accuracy:  
0.6756  
Epoch 117/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7935 - accuracy:  
0.6765  
Epoch 118/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7572 - accuracy:  
0.6963  
Epoch 119/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7664 - accuracy:  
0.6926  
Epoch 120/200  
68/68 [=====] - 0s 3ms/step - loss: 0.7658 - accuracy:  
0.6861  
Epoch 121/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7860 - accuracy:  
0.6820  
Epoch 122/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7727 - accuracy:  
0.6889  
Epoch 123/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7751 - accuracy:  
0.6742  
Epoch 124/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7710 - accuracy:  
0.6894

Epoch 125/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7672 - accuracy: 0.6967

Epoch 126/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7347 - accuracy: 0.6972

Epoch 127/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7531 - accuracy: 0.6954

Epoch 128/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7356 - accuracy: 0.7069

Epoch 129/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7539 - accuracy: 0.6963

Epoch 130/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7536 - accuracy: 0.6898

Epoch 131/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7568 - accuracy: 0.6963

Epoch 132/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7600 - accuracy: 0.6926

Epoch 133/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7421 - accuracy: 0.7000

Epoch 134/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7514 - accuracy: 0.6954

Epoch 135/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7266 - accuracy: 0.7013

Epoch 136/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7282 - accuracy: 0.7004

Epoch 137/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7208 - accuracy: 0.7188

Epoch 138/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7340 - accuracy: 0.7124

Epoch 139/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7322 - accuracy: 0.7055

Epoch 140/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7290 - accuracy: 0.7032

Epoch 141/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7140 - accuracy: 0.7156

Epoch 142/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7143 - accuracy: 0.7266

Epoch 143/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6971 - accuracy: 0.7230

Epoch 144/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7223 - accuracy: 0.7156

Epoch 145/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7288 - accuracy: 0.6981

Epoch 146/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7083 - accuracy: 0.7276

Epoch 147/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7134 - accuracy: 0.7087

Epoch 148/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7049 - accuracy: 0.7262

Epoch 149/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6950 - accuracy: 0.7220

Epoch 150/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6937 - accuracy: 0.7197

Epoch 151/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7005 - accuracy: 0.7312

Epoch 152/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6718 - accuracy: 0.7363

Epoch 153/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6887 - accuracy: 0.7202

Epoch 154/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6875 - accuracy: 0.7322

Epoch 155/200  
68/68 [=====] - 0s 2ms/step - loss: 0.7009 - accuracy: 0.7230

Epoch 156/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6805 - accuracy: 0.7294

Epoch 157/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6732 - accuracy:  
0.7382

Epoch 158/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6859 - accuracy:  
0.7326

Epoch 159/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6832 - accuracy:  
0.7368

Epoch 160/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6882 - accuracy:  
0.7184

Epoch 161/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6802 - accuracy:  
0.7280

Epoch 162/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6480 - accuracy:  
0.7474

Epoch 163/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6646 - accuracy:  
0.7386

Epoch 164/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6612 - accuracy:  
0.7377

Epoch 165/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6588 - accuracy:  
0.7414

Epoch 166/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6810 - accuracy:  
0.7312

Epoch 167/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6746 - accuracy:  
0.7386

Epoch 168/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6651 - accuracy:  
0.7345

Epoch 169/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6373 - accuracy:  
0.7520

Epoch 170/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6609 - accuracy:  
0.7349

Epoch 171/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6681 - accuracy:  
0.7349

Epoch 172/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6694 - accuracy:  
0.7386



Epoch 173/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6648 - accuracy:  
0.7437  
Epoch 174/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6491 - accuracy:  
0.7533  
Epoch 175/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6650 - accuracy:  
0.7322  
Epoch 176/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6411 - accuracy:  
0.7469  
Epoch 177/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6477 - accuracy:  
0.7460  
Epoch 178/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6573 - accuracy:  
0.7483  
Epoch 179/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6380 - accuracy:  
0.7428  
Epoch 180/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6401 - accuracy:  
0.7625  
Epoch 181/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6311 - accuracy:  
0.7483  
Epoch 182/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6340 - accuracy:  
0.7460  
Epoch 183/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6475 - accuracy:  
0.7432  
Epoch 184/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6192 - accuracy:  
0.7621  
Epoch 185/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6426 - accuracy:  
0.7552  
Epoch 186/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6415 - accuracy:  
0.7451  
Epoch 187/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6040 - accuracy:  
0.7621  
Epoch 188/200  
68/68 [=====] - 0s 2ms/step - loss: 0.6435 - accuracy:  
0.7469

```

Epoch 189/200
68/68 [=====] - 0s 3ms/step - loss: 0.6144 - accuracy:
0.7520
Epoch 190/200
68/68 [=====] - 0s 2ms/step - loss: 0.6128 - accuracy:
0.7630
Epoch 191/200
68/68 [=====] - 0s 2ms/step - loss: 0.6178 - accuracy:
0.7635
Epoch 192/200
68/68 [=====] - 0s 2ms/step - loss: 0.6112 - accuracy:
0.7612
Epoch 193/200
68/68 [=====] - 0s 2ms/step - loss: 0.6266 - accuracy:
0.7478
Epoch 194/200
68/68 [=====] - 0s 2ms/step - loss: 0.6098 - accuracy:
0.7621
Epoch 195/200
68/68 [=====] - 0s 2ms/step - loss: 0.5880 - accuracy:
0.7759
Epoch 196/200
68/68 [=====] - 0s 2ms/step - loss: 0.6252 - accuracy:
0.7566
Epoch 197/200
68/68 [=====] - 0s 2ms/step - loss: 0.6166 - accuracy:
0.7639
Epoch 198/200
68/68 [=====] - 0s 2ms/step - loss: 0.6108 - accuracy:
0.7635
Epoch 199/200
68/68 [=====] - 0s 2ms/step - loss: 0.6016 - accuracy:
0.7676
Epoch 200/200
68/68 [=====] - 0s 3ms/step - loss: 0.6040 - accuracy:
0.7722
17/17 [=====] - 0s 1ms/step
Cross-Validation Scores:
[0.38769338 0.3581837 0.38846864 0.36907976 0.38542935]
Training F1 Score: 0.3777709634221348
95% CI: (0.36098846524415357, 0.394553461600116)
22/22 [=====] - 0s 1ms/step
Test Set Evaluation:

```

	precision	recall	f1-score	support
1	0.45	0.30	0.36	33
2	0.26	0.34	0.29	70
3	0.30	0.30	0.30	111

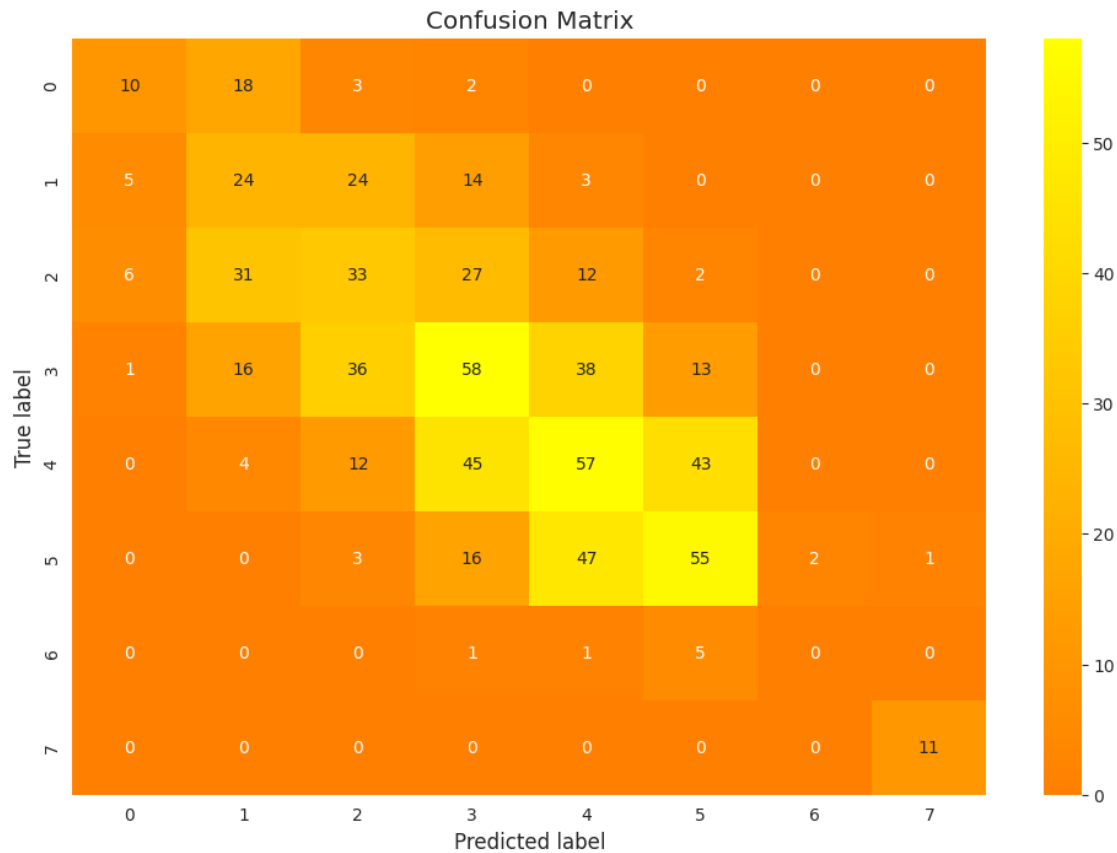
4	0.36	0.36	0.36	162
5	0.36	0.35	0.36	161
6	0.47	0.44	0.45	124
7	0.00	0.00	0.00	7
8	0.92	1.00	0.96	11
accuracy			0.37	679
macro avg	0.39	0.39	0.39	679
weighted avg	0.37	0.37	0.37	679

The neural network using synthetic data has the best 95% confidence interval metrics when compared to other models on the training data, but the accuracy on testing data is not better.

```
[82]: def draw_confusion_matrix(cm):
    plt.figure(figsize=(12,8))
    sns.heatmap(cm,annot=True,fmt="d", center=0, cmap='autumn')
    plt.title("Confusion Matrix")
    plt.ylabel('True label')
    plt.xlabel('Predicted label')
    plt.show()

y_test_model_rf = best_estimator.predict(test_set_prepared)
print('Test Set Evaluation:')
cm_rf = confusion_matrix(y_test, y_test_model_rf)
draw_confusion_matrix(cm_rf)
```

```
22/22 [=====] - 0s 942us/step
Test Set Evaluation:
```



```
[83]: def calculate_accuracy(true_labels, predicted_labels):
    total_predictions = len(true_labels)
    correct_predictions = 0

    for true_label, predicted_label in zip(true_labels, predicted_labels):
        if abs(predicted_label - true_label) <= 1:
            correct_predictions += 1

    accuracy = (correct_predictions / total_predictions) * 100
    return accuracy
calculate_accuracy(y_test, y_test_model_rf)
```

[83]: 83.79970544918999

[ ]: