

main

April 30, 2021

1 MATH 3660 Final Project

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1.1 Introduction

- We got the dataset from <https://www.kaggle.com/cnic92/200-financial-indicators-of-us-stocks-20142018>
- It contains about ~200 columns for each of the companies from their 10-K filings.

1.1.1 Question

Can we use machine learning and deep learning techniques to predict the price variance of a company?

1.2 Data Processing

- Here, we're processing the data to clean up some issues where there are NaN (Not a Number) and crazy outliers.

```
[1]: import pandas as pd
import numpy as np
from sklearn.impute import KNNImputer
from sklearn.preprocessing import RobustScaler
from sklearn.svm import SVC
from sklearn.metrics import classification_report
from sklearn.model_selection import train_test_split
import seaborn as sns
import matplotlib.pyplot as plt
sns.set_theme(style="whitegrid")
```

```
[2]: # Import the data as dataframes for each year
df1=pd.read_csv("./data/2014_Financial_Data.csv")
df2=pd.read_csv("./data/2015_Financial_Data.csv")
df3=pd.read_csv("./data/2016_Financial_Data.csv")
df4=pd.read_csv("./data/2017_Financial_Data.csv")
df5=pd.read_csv("./data/2018_Financial_Data.csv")
```

```
[3]: df1.head()
```

```

[3]: Unnamed: 0      Revenue  Revenue Growth  Cost of Revenue  Gross Profit  \
0      PG  7.440100e+10      -0.0713      3.903000e+10  3.537100e+10
1      VIPS  3.734148e+09      1.1737      2.805625e+09  9.285226e+08
2      KR  9.837500e+10      0.0182      7.813800e+10  2.023700e+10
3      RAD  2.552641e+10      0.0053      1.820268e+10  7.323734e+09
4      GIS  1.790960e+10      0.0076      1.153980e+10  6.369800e+09

      R&D Expenses  SG&A Expense  Operating Expenses  Operating Income  \
0  0.000000e+00  2.146100e+10      2.146100e+10      1.391000e+10
1  1.083303e+08  3.441414e+08      7.939267e+08      1.345959e+08
2  0.000000e+00  1.519600e+10      1.751200e+10      2.725000e+09
3  0.000000e+00  6.561162e+09      6.586482e+09      7.372520e+08
4  0.000000e+00  3.474300e+09      3.412400e+09      2.957400e+09

      Interest Expense  ...  Receivables growth  Inventory Growth  Asset Growth  \
0      7.090000e+08  ...      -0.0187      -0.0217      0.0359
1      1.214869e+07  ...      NaN      NaN      NaN
2      4.430000e+08  ...      0.0618      0.0981      0.1886
3      4.245910e+08  ...      0.0211      -0.0510      -0.0189
4      3.024000e+08  ...      0.0257      0.0090      0.0215

      Book Value per Share Growth  Debt Growth  R&D Expense Growth  \
0      0.0316      0.1228      0.0000
1      NaN      NaN      1.6484
2      0.3268      0.2738      0.0000
3      0.1963      -0.0458      0.0000
4      0.0274      0.1025      0.0000

      SG&A Expenses Growth      Sector  2015 PRICE VAR [%]  Class
0      -0.1746  Consumer Defensive      -9.323276      0
1      1.7313  Consumer Defensive     -25.512193      0
2      0.0234  Consumer Defensive     33.118297      1
3     -0.0060  Consumer Defensive      2.752291      1
4     -0.0220  Consumer Defensive     12.897715      1

```

[5 rows x 225 columns]

```

[4]: def better_than_spy(row, spy_var):
      price_var = row['PRICE_VAR']
      if price_var > spy_var:
          return 1
      else:
          return 0

```

```

[5]: # Drop stock ticker and sector
df1 = df1.drop(df1.columns[0], axis = 1)
df2 = df2.drop(df2.columns[0], axis = 1)

```

```

df3 = df3.drop(df3.columns[0], axis = 1)
df4 = df4.drop(df4.columns[0], axis = 1)
df5 = df5.drop(df5.columns[0], axis = 1)

df1.drop(['Sector'], axis=1, inplace=True)
df2.drop(['Sector'], axis=1, inplace=True)
df3.drop(['Sector'], axis=1, inplace=True)
df4.drop(['Sector'], axis=1, inplace=True)
df5.drop(['Sector'], axis=1, inplace=True)

df1['Year'] = 2014
df2['Year'] = 2015
df3['Year'] = 2016
df4['Year'] = 2017
df5['Year'] = 2018

# Rename price var
df1.rename(columns={"2015 PRICE VAR [%]": "PRICE_VAR"},inplace=True)
df2.rename(columns={"2016 PRICE VAR [%]": "PRICE_VAR"},inplace=True)
df3.rename(columns={"2017 PRICE VAR [%]": "PRICE_VAR"},inplace=True)
df4.rename(columns={"2018 PRICE VAR [%]": "PRICE_VAR"},inplace=True)
df5.rename(columns={"2019 PRICE VAR [%]": "PRICE_VAR"},inplace=True)

```

```

[6]: # Use SPY as another metric for looking at change, calculating its variance,
      ↪ over the course of the year and using that.
spy = pd.read_csv("./data/SPY.csv", index_col='Date')

spy2015 = spy.loc[['2015-01-01', '2016-01-01']]['Open']
spy2015change = ((spy2015[1] - spy2015[0]) / spy2015[0]) * 100
df1['Beat Spy'] = df1.apply(lambda row: better_than_spy(row, spy2015change),
    ↪ axis=1)

spy2016 = spy.loc[['2016-01-01', '2017-01-01']]['Open']
spy2016change = ((spy2016[1] - spy2016[0]) / spy2016[0]) * 100
df2['Beat Spy'] = df2.apply(lambda row: better_than_spy(row, spy2016change),
    ↪ axis=1)

spy2017 = spy.loc[['2017-01-01', '2018-01-01']]['Open']
spy2017change = ((spy2017[1] - spy2017[0]) / spy2017[0]) * 100
df3['Beat Spy'] = df3.apply(lambda row: better_than_spy(row, spy2017change),
    ↪ axis=1)

spy2018 = spy.loc[['2018-01-01', '2019-01-01']]['Open']
spy2018change = ((spy2018[1] - spy2018[0]) / spy2018[0]) * 100
df4['Beat Spy'] = df4.apply(lambda row: better_than_spy(row, spy2018change),
    ↪ axis=1)

```

```

spy2019 = spy.loc[['2019-01-01', '2020-01-01']]['Open']
spy2019change = ((spy2019[1] - spy2019[0]) / spy2019[0]) * 100
df5['Beat Spy'] = df5.apply(lambda row: better_than_spy(row, spy2019change),
    ↪axis=1)

```

[7]: *# Clean the data, removing NaN values and completing them using KNN*

```

imputer = KNNImputer(n_neighbors=10, weights='distance',
    ↪metric='nan_euclidean', copy=True)

```

```

df1_clean = imputer.fit_transform(df1)
df1_clean = pd.DataFrame(df1_clean)
df1_clean.columns = list(df1)

```

```

df2_clean = imputer.fit_transform(df2)
df2_clean = pd.DataFrame(df2_clean)
df2_clean.columns = list(df2)

```

```

df3_clean = imputer.fit_transform(df3)
df3_clean = pd.DataFrame(df3_clean)
df3_clean.columns = list(df3)

```

```

df4_clean = imputer.fit_transform(df4)
df4_clean = pd.DataFrame(df4_clean)
df4_clean.columns = list(df4)

```

```

df5_clean = imputer.fit_transform(df5)
df5_clean = pd.DataFrame(df5_clean)
df5_clean.columns = list(df5)

```

[8]: *# Then, concatenate all of the dataframes into a single dataframe*

```

data = pd.concat([df1_clean, df2_clean, df3_clean, df4_clean, df5_clean])

```

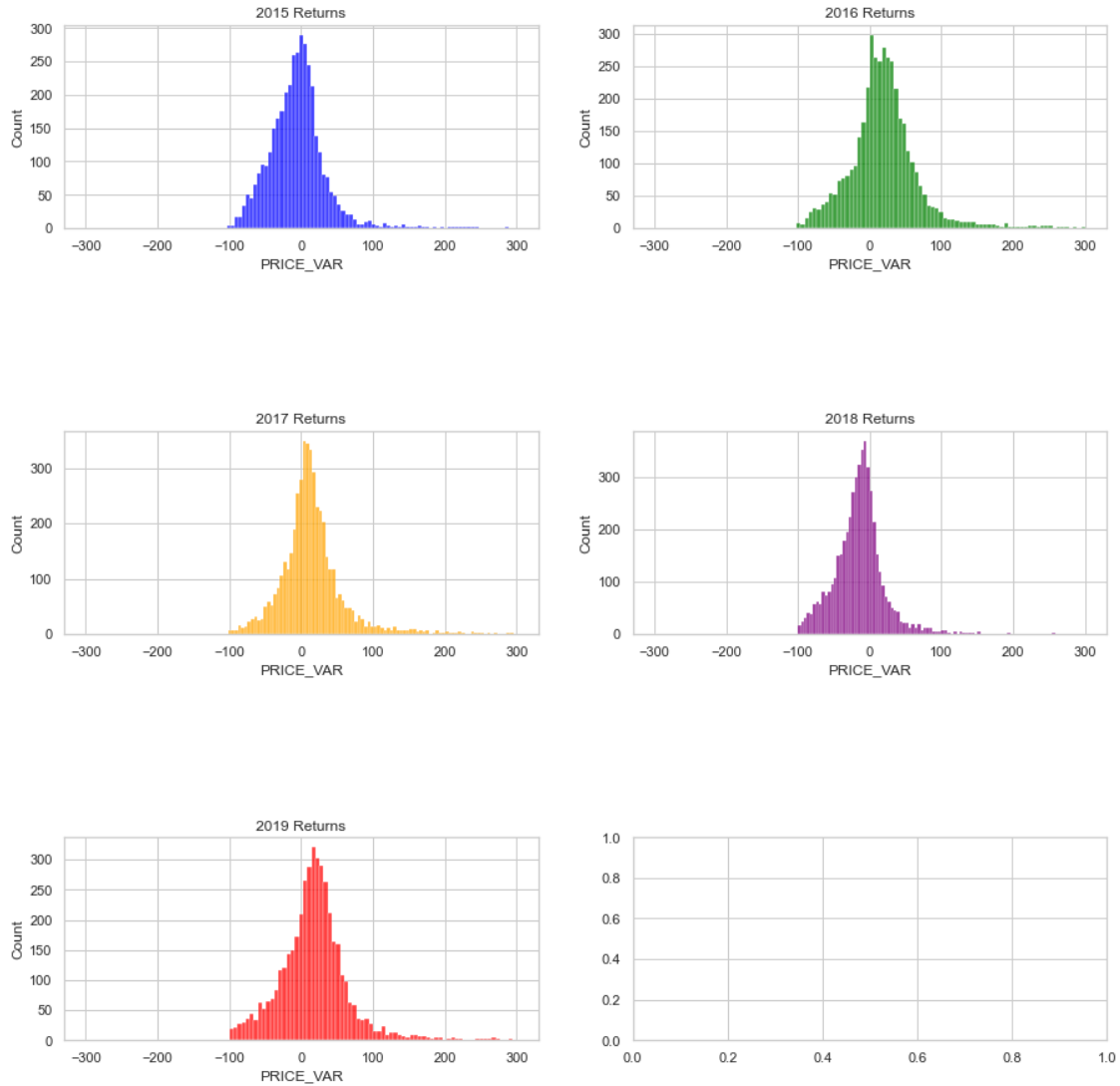
[9]: *# Plot the data for each of the years, graphing what their returns are like*

```

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
fig.subplots_adjust(hspace=1)
sns.histplot(data=df1['PRICE_VAR'], binrange=(-300,300), stat='count',
    ↪color='blue', ax=axs[0, 0]).set_title('2015 Returns')
sns.histplot(data=df2['PRICE_VAR'], binrange=(-300,300), stat='count',
    ↪color='green', ax=axs[0, 1]).set_title('2016 Returns')
sns.histplot(data=df3['PRICE_VAR'], binrange=(-300,300), stat='count',
    ↪color='orange', ax=axs[1, 0]).set_title('2017 Returns')
sns.histplot(data=df4['PRICE_VAR'], binrange=(-300,300), stat='count',
    ↪color='purple', ax=axs[1, 1]).set_title('2018 Returns')
sns.histplot(data=df5['PRICE_VAR'], binrange=(-300,300), stat='count',
    ↪color='red', ax=axs[2, 0]).set_title('2019 Returns')

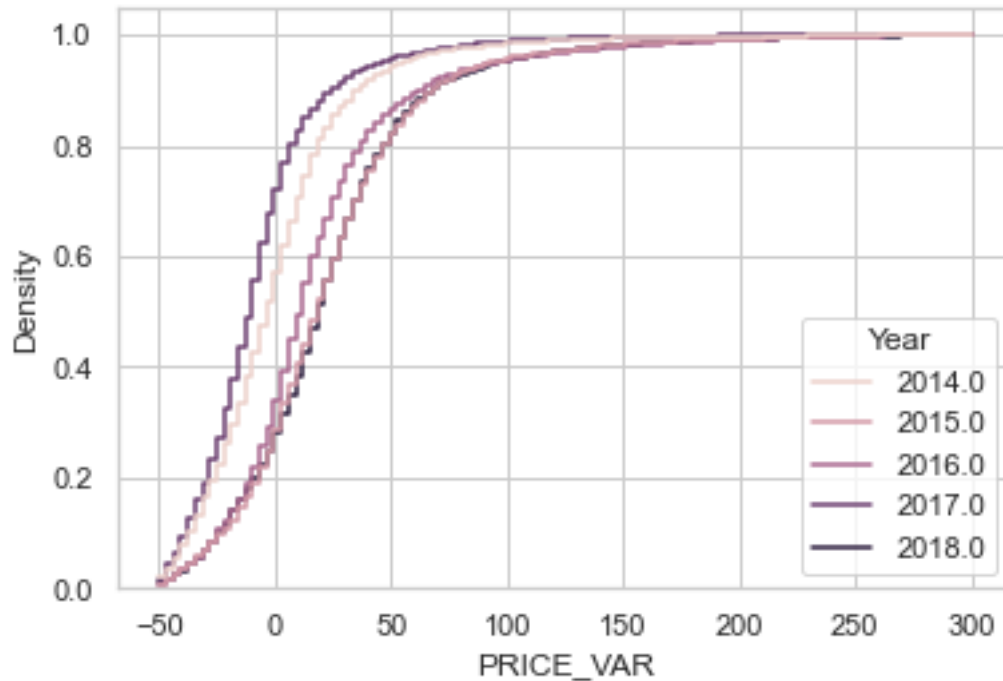
```

[9]: Text(0.5, 1.0, '2019 Returns')



```
[10]: sns.histplot(  
    data=data, x='PRICE_VAR', hue='Year',  
    element="step", fill=False,  
    cumulative=True, common_norm=False, stat='density', binrange=(-50,300)  
)
```

[10]: <AxesSubplot:xlabel='PRICE_VAR', ylabel='Density'>



```
[11]: print(f"In {df1['Year'][0]}, {(sum(df1['Beat Spy'] == 1) / len(df1['Beat_
    ↳Spy']))*100}% of companies had better performance than SPY the next year")
print(f"In {df2['Year'][0]}, {(sum(df2['Beat Spy'] == 1) / len(df2['Beat_
    ↳Spy']))*100}% of companies had better performance than SPY the next year")
print(f"In {df3['Year'][0]}, {(sum(df3['Beat Spy'] == 1) / len(df3['Beat_
    ↳Spy']))*100}% of companies had better performance than SPY the next year")
print(f"In {df4['Year'][0]}, {(sum(df4['Beat Spy'] == 1) / len(df4['Beat_
    ↳Spy']))*100}% of companies had better performance than SPY the next year")
print(f"In {df5['Year'][0]}, {(sum(df5['Beat Spy'] == 1) / len(df5['Beat_
    ↳Spy']))*100}% of companies had better performance than SPY the next year")
print("\n")
print(f"In {df1['Year'][0]}, {(sum(df1['Class'] == 1) /
    ↳len(df1['Class']))*100}% of companies had positive stock growth the next_
    ↳year")
print(f"In {df2['Year'][0]}, {(sum(df2['Class'] == 1) /
    ↳len(df2['Class']))*100}% of companies had positive stock growth the next_
    ↳year")
print(f"In {df3['Year'][0]}, {(sum(df3['Class'] == 1) /
    ↳len(df3['Class']))*100}% of companies had positive stock growth the next_
    ↳year")
print(f"In {df4['Year'][0]}, {(sum(df4['Class'] == 1) /
    ↳len(df4['Class']))*100}% of companies had positive stock growth the next_
    ↳year")
```

```
print(f"In {df5['Year'][0]}, {(sum(df5['Class'] == 1) /
↳len(df5['Class']))*100}% of companies had positive stock growth the next_
↳year")
```

In 2014, 46.58613445378151% of companies had better performance than SPY the next year

In 2015, 55.46116504854369% of companies had better performance than SPY the next year

In 2016, 37.91953304148426% of companies had better performance than SPY the next year

In 2017, 41.16935483870968% of companies had better performance than SPY the next year

In 2018, 33.310564663023676% of companies had better performance than SPY the next year

In 2014, 42.909663865546214% of companies had positive stock growth the next year

In 2015, 70.16990291262137% of companies had positive stock growth the next year

In 2016, 67.0835939128622% of companies had positive stock growth the next year

In 2017, 27.62096774193548% of companies had positive stock growth the next year

In 2018, 69.35336976320583% of companies had positive stock growth the next year

2 Training a Model with SVC

- C-Support Vector Classification, as seen in this: <https://scikit-learn.org/stable/modules/generated/sklearn.svm.SVC.html>

```
[12]: # We get all but the last four rows of the data
X = data.iloc[:, :-4]
y = data.iloc[:, -1:] # for spy, do -3 for buy/sell
y = np.array([int(i[0]) for i in y.values.tolist()])

# Look at dimensions for sanity check
print("Dimensions of X: {}".format(X.shape))
print("Dimensions of y: {}".format(y.shape))
```

Dimensions of X: (22077, 221)

Dimensions of y: (22077,)

```
[13]: # Canonical 80/20% train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y)
```

```
[14]: transformer = RobustScaler(unit_variance=True).fit(X_train)
X_train = transformer.transform(X_train)
X_test = transformer.transform(X_test)
```

```
[15]: svm = SVC(C=1, gamma='scale', class_weight='balanced')
      svm.fit(X_train, y_train)
      y_pred = svm.predict(X_test)
```

```
[16]: target_names = ['Doesnt Beat SPY', 'Beats SPY']
      print(classification_report(y_test, y_pred, target_names=target_names))
```

| | precision | recall | f1-score | support |
|-----------------|-----------|--------|----------|---------|
| Doesnt Beat SPY | 0.58 | 0.98 | 0.73 | 3192 |
| Beats SPY | 0.42 | 0.02 | 0.04 | 2328 |
| accuracy | | | 0.57 | 5520 |
| macro avg | 0.50 | 0.50 | 0.39 | 5520 |
| weighted avg | 0.51 | 0.57 | 0.44 | 5520 |

3 Deep Learning Approach

- Next, we take a look at using a deep learning model.

```
[41]: import tensorflow as tf

      from tensorflow import keras
      from tensorflow.keras import layers
      from tensorflow import feature_column
```

```
[81]: # We gotta grab out the feature columns as well
      feature_columns = X.columns.values
      feature_columns = [feature_column.numeric_column(val) for val in
      ↪feature_columns]

      feature_layer = tf.keras.layers.DenseFeatures(feature_columns)

      model = keras.Sequential([
          keras.layers.Dense(16, activation=tf.nn.relu),
          keras.layers.Dense(16, activation=tf.nn.relu),
          keras.layers.Dense(1, activation=tf.nn.sigmoid),
      ])
```

```
[82]: y_train.shape
```

```
[82]: (16557,)
```

```
[83]: model.compile(optimizer='adam',
                    loss=tf.keras.losses.BinaryCrossentropy(from_logits=True),
```



```
metrics=['accuracy'])

history = model.fit(X_train, y_train, epochs=50)
test_loss, test_acc = model.evaluate(X_test, y_test)
```

Train on 16557 samples

Epoch 1/50

16557/16557 [=====] - 2s 103us/sample - loss: 0.6980 -
accuracy: 0.5698

Epoch 2/50

16557/16557 [=====] - 1s 58us/sample - loss: 0.6933 -
accuracy: 0.5738

Epoch 3/50

16557/16557 [=====] - 1s 58us/sample - loss: 0.6930 -
accuracy: 0.5740

Epoch 4/50

16557/16557 [=====] - 1s 59us/sample - loss: 0.6925 -
accuracy: 0.5748

Epoch 5/50

16557/16557 [=====] - 1s 58us/sample - loss: 0.6923 -
accuracy: 0.5762

Epoch 6/50

16557/16557 [=====] - 1s 60us/sample - loss: 0.6917 -
accuracy: 0.5788

Epoch 7/50

16557/16557 [=====] - 1s 59us/sample - loss: 0.6915 -
accuracy: 0.5810

Epoch 8/50

16557/16557 [=====] - 1s 57us/sample - loss: 0.6908 -
accuracy: 0.5828

Epoch 9/50

16557/16557 [=====] - 1s 59us/sample - loss: 0.6896 -
accuracy: 0.5856

Epoch 10/50

16557/16557 [=====] - 1s 59us/sample - loss: 0.6886 -
accuracy: 0.5912

Epoch 11/50

16557/16557 [=====] - 1s 60us/sample - loss: 0.6871 -
accuracy: 0.5948

Epoch 12/50

16557/16557 [=====] - 1s 60us/sample - loss: 0.6860 -
accuracy: 0.5980

Epoch 13/50

16557/16557 [=====] - 1s 59us/sample - loss: 0.6857 -
accuracy: 0.5999

Epoch 14/50

16557/16557 [=====] - 1s 58us/sample - loss: 0.6850 -

```

accuracy: 0.6015
Epoch 15/50
16557/16557 [=====] - 1s 58us/sample - loss: 0.6844 -
accuracy: 0.6032
Epoch 16/50
16557/16557 [=====] - 1s 58us/sample - loss: 0.6831 -
accuracy: 0.6063
Epoch 17/50
16557/16557 [=====] - 1s 59us/sample - loss: 0.6822 -
accuracy: 0.6088
Epoch 18/50
16557/16557 [=====] - 1s 59us/sample - loss: 0.6825 -
accuracy: 0.6080
Epoch 19/50
16557/16557 [=====] - 1s 60us/sample - loss: 0.6812 -
accuracy: 0.6112
Epoch 20/50
16557/16557 [=====] - 1s 59us/sample - loss: 0.6806 -
accuracy: 0.6124
Epoch 21/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6803 -
accuracy: 0.6134
Epoch 22/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6803 -
accuracy: 0.6146
Epoch 23/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6803 -
accuracy: 0.6132
Epoch 24/50
16557/16557 [=====] - 1s 56us/sample - loss: 0.6791 -
accuracy: 0.6170
Epoch 25/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6788 -
accuracy: 0.6179
Epoch 26/50
16557/16557 [=====] - 1s 58us/sample - loss: 0.6789 -
accuracy: 0.6173
Epoch 27/50
16557/16557 [=====] - 1s 61us/sample - loss: 0.6781 -
accuracy: 0.6190
Epoch 28/50
16557/16557 [=====] - 1s 58us/sample - loss: 0.6780 -
accuracy: 0.6197
Epoch 29/50
16557/16557 [=====] - 1s 58us/sample - loss: 0.6781 -
accuracy: 0.6196
Epoch 30/50
16557/16557 [=====] - 1s 58us/sample - loss: 0.6789 -

```

```

accuracy: 0.6185
Epoch 31/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6777 -
accuracy: 0.6201
Epoch 32/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6773 -
accuracy: 0.6212
Epoch 33/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6776 -
accuracy: 0.6195
Epoch 34/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6767 -
accuracy: 0.6223
Epoch 35/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6762 -
accuracy: 0.6230
Epoch 36/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6759 -
accuracy: 0.6241
Epoch 37/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6774 -
accuracy: 0.6215
Epoch 38/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6769 -
accuracy: 0.6225
Epoch 39/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6765 -
accuracy: 0.6232
Epoch 40/50
16557/16557 [=====] - 1s 56us/sample - loss: 0.6760 -
accuracy: 0.6251
Epoch 41/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6760 -
accuracy: 0.6253
Epoch 42/50
16557/16557 [=====] - 1s 58us/sample - loss: 0.6760 -
accuracy: 0.6245
Epoch 43/50
16557/16557 [=====] - 1s 56us/sample - loss: 0.6763 -
accuracy: 0.6241
Epoch 44/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6759 -
accuracy: 0.6247
Epoch 45/50
16557/16557 [=====] - 1s 57us/sample - loss: 0.6754 -
accuracy: 0.6261
Epoch 46/50
16557/16557 [=====] - 1s 56us/sample - loss: 0.6750 -

```


Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

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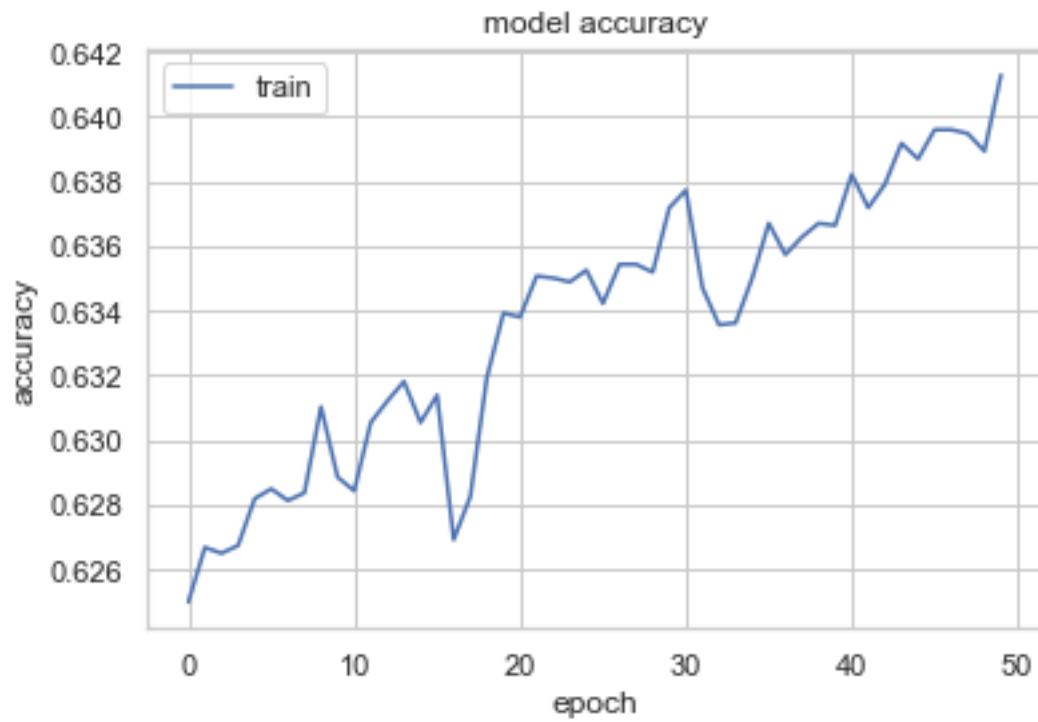
Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

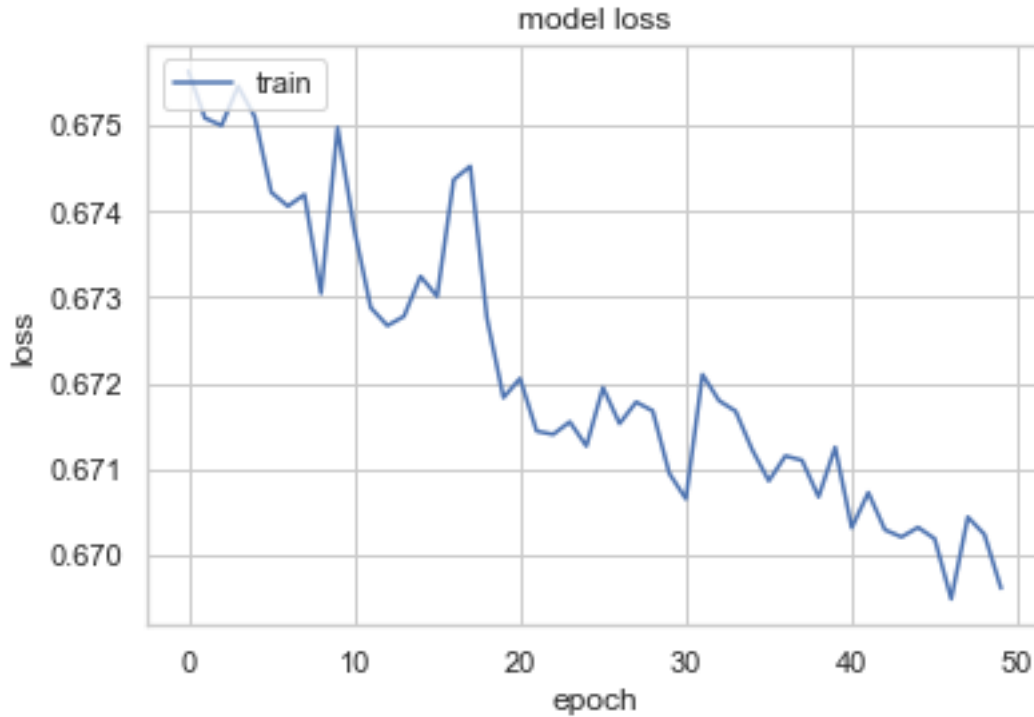
Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.


```
[57]: plot_history(history)
```

```
dict_keys(['loss', 'accuracy'])
```





3.0.1 Results

- This isn't awful, but it isn't great. The total training time isn't great, but it could be worse. We see that the loss graph isn't bottoming out – maybe our naive model has more space to go?

3.1 Make a more complex model and try it again

```
[59]: model = keras.Sequential([
    keras.layers.Dense(128, activation=tf.nn.relu),
    keras.layers.Dense(128, activation=tf.nn.relu),
    keras.layers.Dense(1, activation=tf.nn.sigmoid),
])
```

```
[60]: model.compile(optimizer='adam',
    loss=tf.keras.losses.BinaryCrossentropy(from_logits=True),
    metrics=['accuracy'])

history = model.fit(X_train, y_train, epochs=50)
test_loss, test_acc = model.evaluate(X_test, y_test)
```

Train on 16557 samples
Epoch 1/50

16557/16557 [=====] - 2s 143us/sample - loss: 0.6946 -
 accuracy: 0.5732
 Epoch 2/50
 16557/16557 [=====] - 2s 106us/sample - loss: 0.6931 -
 accuracy: 0.5739
 Epoch 3/50
 16557/16557 [=====] - 2s 106us/sample - loss: 0.6933 -
 accuracy: 0.5743
 Epoch 4/50
 16557/16557 [=====] - 2s 107us/sample - loss: 0.6934 -
 accuracy: 0.5751
 Epoch 5/50
 16557/16557 [=====] - 2s 109us/sample - loss: 0.6924 -
 accuracy: 0.5776
 Epoch 6/50
 16557/16557 [=====] - 2s 109us/sample - loss: 0.6917 -
 accuracy: 0.5790
 Epoch 7/50
 16557/16557 [=====] - 2s 110us/sample - loss: 0.6910 -
 accuracy: 0.5805
 Epoch 8/50
 16557/16557 [=====] - 2s 108us/sample - loss: 0.6906 -
 accuracy: 0.5817
 Epoch 9/50
 16557/16557 [=====] - 2s 107us/sample - loss: 0.6911 -
 accuracy: 0.5818
 Epoch 10/50
 16557/16557 [=====] - 2s 107us/sample - loss: 0.6908 -
 accuracy: 0.5819
 Epoch 11/50
 16557/16557 [=====] - 2s 107us/sample - loss: 0.6901 -
 accuracy: 0.5830
 Epoch 12/50
 16557/16557 [=====] - 2s 111us/sample - loss: 0.6901 -
 accuracy: 0.5831
 Epoch 13/50
 16557/16557 [=====] - 2s 111us/sample - loss: 0.6899 -
 accuracy: 0.5834
 Epoch 14/50
 16557/16557 [=====] - 2s 110us/sample - loss: 0.6898 -
 accuracy: 0.5837
 Epoch 15/50
 16557/16557 [=====] - 2s 117us/sample - loss: 0.6896 -
 accuracy: 0.5843
 Epoch 16/50
 16557/16557 [=====] - 2s 113us/sample - loss: 0.6898 -
 accuracy: 0.5858
 Epoch 17/50

16557/16557 [=====] - 2s 115us/sample - loss: 0.6898 -
 accuracy: 0.5860
 Epoch 18/50
 16557/16557 [=====] - 2s 105us/sample - loss: 0.6895 -
 accuracy: 0.5865
 Epoch 19/50
 16557/16557 [=====] - 2s 106us/sample - loss: 0.6892 -
 accuracy: 0.5850
 Epoch 20/50
 16557/16557 [=====] - 2s 105us/sample - loss: 0.6892 -
 accuracy: 0.5860
 Epoch 21/50
 16557/16557 [=====] - 2s 104us/sample - loss: 0.6887 -
 accuracy: 0.5867
 Epoch 22/50
 16557/16557 [=====] - 2s 104us/sample - loss: 0.6932 -
 accuracy: 0.5851
 Epoch 23/50
 16557/16557 [=====] - 2s 105us/sample - loss: 0.6885 -
 accuracy: 0.5876
 Epoch 24/50
 16557/16557 [=====] - 2s 104us/sample - loss: 0.6886 -
 accuracy: 0.5874
 Epoch 25/50
 16557/16557 [=====] - 2s 108us/sample - loss: 0.6886 -
 accuracy: 0.5900
 Epoch 26/50
 16557/16557 [=====] - 2s 112us/sample - loss: 0.6882 -
 accuracy: 0.5882
 Epoch 27/50
 16557/16557 [=====] - 2s 112us/sample - loss: 0.6885 -
 accuracy: 0.5865
 Epoch 28/50
 16557/16557 [=====] - 2s 113us/sample - loss: 0.6880 -
 accuracy: 0.5884
 Epoch 29/50
 16557/16557 [=====] - 2s 109us/sample - loss: 0.6876 -
 accuracy: 0.5891
 Epoch 30/50
 16557/16557 [=====] - 2s 109us/sample - loss: 0.6876 -
 accuracy: 0.5893
 Epoch 31/50
 16557/16557 [=====] - 2s 105us/sample - loss: 0.6873 -
 accuracy: 0.5899
 Epoch 32/50
 16557/16557 [=====] - 2s 108us/sample - loss: 0.6880 -
 accuracy: 0.5886
 Epoch 33/50

16557/16557 [=====] - 2s 110us/sample - loss: 0.6872 -
 accuracy: 0.5903
 Epoch 34/50
 16557/16557 [=====] - 2s 121us/sample - loss: 0.6874 -
 accuracy: 0.5900
 Epoch 35/50
 16557/16557 [=====] - 2s 132us/sample - loss: 0.6878 -
 accuracy: 0.5897
 Epoch 36/50
 16557/16557 [=====] - 2s 141us/sample - loss: 0.6875 -
 accuracy: 0.5903
 Epoch 37/50
 16557/16557 [=====] - 2s 132us/sample - loss: 0.6887 -
 accuracy: 0.5903
 Epoch 38/50
 16557/16557 [=====] - 2s 138us/sample - loss: 0.6875 -
 accuracy: 0.5917
 Epoch 39/50
 16557/16557 [=====] - 2s 141us/sample - loss: 0.6884 -
 accuracy: 0.5921
 Epoch 40/50
 16557/16557 [=====] - 2s 140us/sample - loss: 0.6911 -
 accuracy: 0.5893
 Epoch 41/50
 16557/16557 [=====] - 2s 142us/sample - loss: 0.6906 -
 accuracy: 0.5892
 Epoch 42/50
 16557/16557 [=====] - 2s 140us/sample - loss: 0.6886 -
 accuracy: 0.5924
 Epoch 43/50
 16557/16557 [=====] - 2s 143us/sample - loss: 0.6896 -
 accuracy: 0.5936
 Epoch 44/50
 16557/16557 [=====] - 2s 143us/sample - loss: 0.6892 -
 accuracy: 0.5935
 Epoch 45/50
 16557/16557 [=====] - 2s 150us/sample - loss: 0.6901 -
 accuracy: 0.5944
 Epoch 46/50
 16557/16557 [=====] - 2s 141us/sample - loss: 0.6913 -
 accuracy: 0.5933
 Epoch 47/50
 16557/16557 [=====] - 2s 145us/sample - loss: 0.6920 -
 accuracy: 0.5935
 Epoch 48/50
 16557/16557 [=====] - 2s 139us/sample - loss: 0.6903 -
 accuracy: 0.5940
 Epoch 49/50

16557/16557 [=====] - 2s 137us/sample - loss: 0.6873 -
accuracy: 0.5975

Epoch 50/50

16557/16557 [=====] - 2s 136us/sample - loss: 0.6876 -
accuracy: 0.5978

5520/1 [=====]
=====

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

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Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

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Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

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Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

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Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

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Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

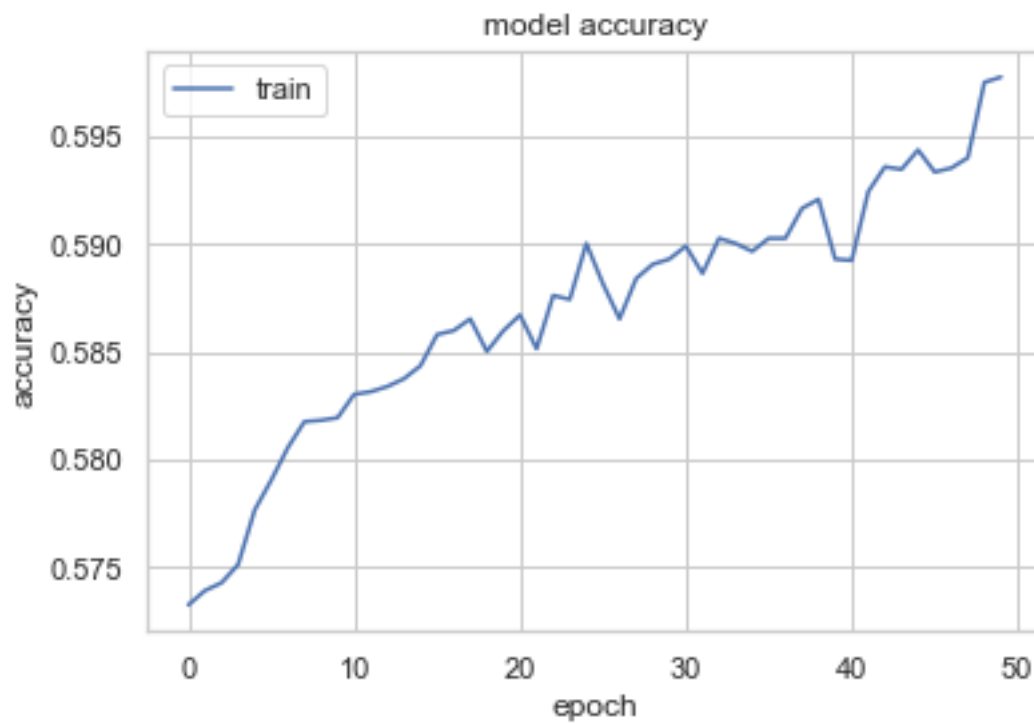
Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

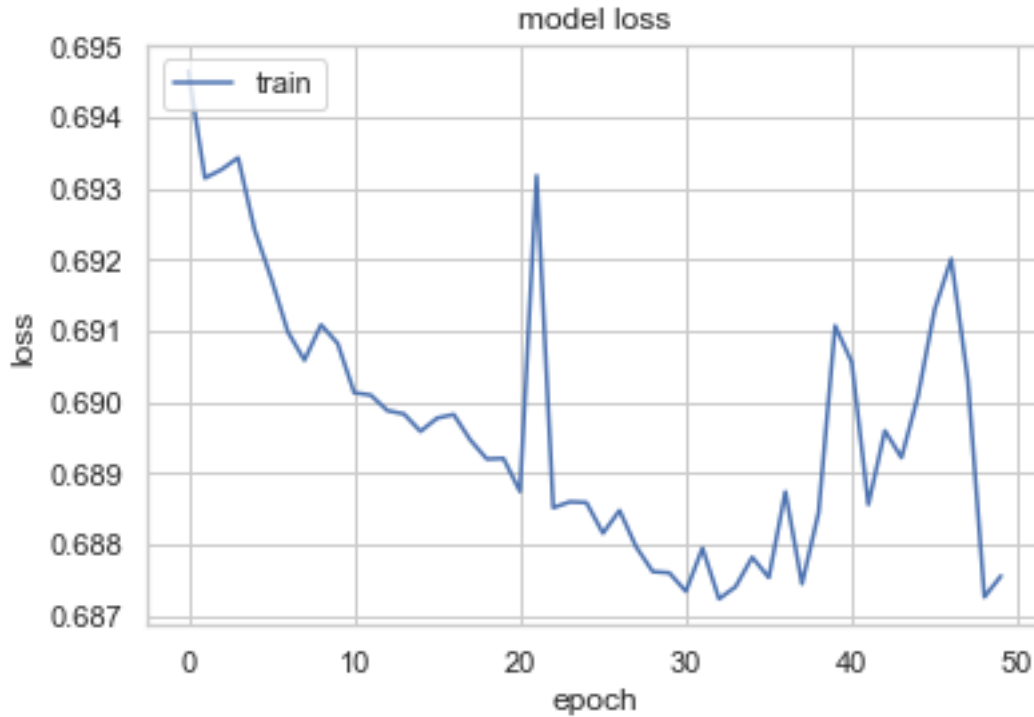
Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

```
=====
=====
=====
=====
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=====
=====
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=====
=====
=====
=====
=====
=====] - 1s 96us/sample - loss: 0.6964 - accuracy: 0.5862
```

```
[61]: print("Accuracy: {}".format(test_acc))
      plot_history(history)
```

Accuracy: 0.5862318873405457





3.2 Results

- Making it a more “complex” model just make it train very slowly and work poorly. Let’s just try using more epochs.

4 1000 Epochs, 16 RELU / 16 RELU

```
[62]: model = keras.Sequential([
    keras.layers.Dense(16, activation=tf.nn.relu),
    keras.layers.Dense(16, activation=tf.nn.relu),
    keras.layers.Dense(1, activation=tf.nn.sigmoid),
])
```

```
[63]: model.compile(optimizer='adam',
    loss=tf.keras.losses.BinaryCrossentropy(from_logits=True),
    metrics=['accuracy'])

history = model.fit(X_train, y_train, epochs=1000)
test_loss, test_acc = model.evaluate(X_test, y_test)
```

Train on 16557 samples
Epoch 1/1000

16557/16557 [=====] - 2s 101us/sample - loss: 0.6938 -
 accuracy: 0.5740
 Epoch 2/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6930 -
 accuracy: 0.5741
 Epoch 3/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6928 -
 accuracy: 0.5748
 Epoch 4/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6919 -
 accuracy: 0.5775
 Epoch 5/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6916 -
 accuracy: 0.5798
 Epoch 6/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6911 -
 accuracy: 0.5824
 Epoch 7/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6899 -
 accuracy: 0.5871
 Epoch 8/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6895 -
 accuracy: 0.5887
 Epoch 9/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6882 -
 accuracy: 0.5913
 Epoch 10/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6872 -
 accuracy: 0.5947
 Epoch 11/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6863 -
 accuracy: 0.5976
 Epoch 12/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6851 -
 accuracy: 0.5995
 Epoch 13/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6846 -
 accuracy: 0.6025
 Epoch 14/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6843 -
 accuracy: 0.6030
 Epoch 15/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6836 -
 accuracy: 0.6047
 Epoch 16/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6837 -
 accuracy: 0.6046
 Epoch 17/1000

16557/16557 [=====] - 1s 51us/sample - loss: 0.6838 -
 accuracy: 0.6053
 Epoch 18/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6833 -
 accuracy: 0.6054
 Epoch 19/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6820 -
 accuracy: 0.6095
 Epoch 20/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6818 -
 accuracy: 0.6092
 Epoch 21/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6812 -
 accuracy: 0.6106
 Epoch 22/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6808 -
 accuracy: 0.6110
 Epoch 23/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6804 -
 accuracy: 0.6116
 Epoch 24/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6801 -
 accuracy: 0.6128
 Epoch 25/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6812 -
 accuracy: 0.6126
 Epoch 26/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6810 -
 accuracy: 0.6104
 Epoch 27/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6796 -
 accuracy: 0.6145
 Epoch 28/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6797 -
 accuracy: 0.6156
 Epoch 29/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6789 -
 accuracy: 0.6164
 Epoch 30/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6793 -
 accuracy: 0.6151
 Epoch 31/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6782 -
 accuracy: 0.6171
 Epoch 32/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6779 -
 accuracy: 0.6183
 Epoch 33/1000

16557/16557 [=====] - 1s 51us/sample - loss: 0.6790 -
 accuracy: 0.6164
 Epoch 34/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6781 -
 accuracy: 0.6187
 Epoch 35/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6777 -
 accuracy: 0.6191
 Epoch 36/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6773 -
 accuracy: 0.6205
 Epoch 37/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6775 -
 accuracy: 0.6199
 Epoch 38/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6772 -
 accuracy: 0.6206
 Epoch 39/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6775 -
 accuracy: 0.6191
 Epoch 40/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6788 -
 accuracy: 0.6157
 Epoch 41/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6771 -
 accuracy: 0.6200
 Epoch 42/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6770 -
 accuracy: 0.6202
 Epoch 43/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6764 -
 accuracy: 0.6213
 Epoch 44/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6760 -
 accuracy: 0.6223
 Epoch 45/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6765 -
 accuracy: 0.6220
 Epoch 46/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6764 -
 accuracy: 0.6227
 Epoch 47/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6757 -
 accuracy: 0.6225
 Epoch 48/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6759 -
 accuracy: 0.6232
 Epoch 49/1000

16557/16557 [=====] - 1s 51us/sample - loss: 0.6762 -
 accuracy: 0.6226
 Epoch 50/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6760 -
 accuracy: 0.6228
 Epoch 51/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6763 -
 accuracy: 0.6220
 Epoch 52/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6760 -
 accuracy: 0.6223
 Epoch 53/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6755 -
 accuracy: 0.6238
 Epoch 54/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6755 -
 accuracy: 0.6234
 Epoch 55/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6754 -
 accuracy: 0.6240
 Epoch 56/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6750 -
 accuracy: 0.6248
 Epoch 57/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6749 -
 accuracy: 0.6246
 Epoch 58/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6748 -
 accuracy: 0.6253
 Epoch 59/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6753 -
 accuracy: 0.6234
 Epoch 60/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6754 -
 accuracy: 0.6231
 Epoch 61/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6756 -
 accuracy: 0.6234
 Epoch 62/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6747 -
 accuracy: 0.6258
 Epoch 63/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6754 -
 accuracy: 0.6239
 Epoch 64/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6744 -
 accuracy: 0.6258
 Epoch 65/1000

16557/16557 [=====] - 1s 54us/sample - loss: 0.6741 -
 accuracy: 0.6272
 Epoch 66/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6744 -
 accuracy: 0.6257
 Epoch 67/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6757 -
 accuracy: 0.6240
 Epoch 68/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6750 -
 accuracy: 0.6252
 Epoch 69/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6746 -
 accuracy: 0.6260
 Epoch 70/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6749 -
 accuracy: 0.6252
 Epoch 71/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6746 -
 accuracy: 0.6257
 Epoch 72/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6743 -
 accuracy: 0.6262
 Epoch 73/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6740 -
 accuracy: 0.6272
 Epoch 74/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6743 -
 accuracy: 0.6263
 Epoch 75/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6748 -
 accuracy: 0.6251
 Epoch 76/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6746 -
 accuracy: 0.6252
 Epoch 77/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6746 -
 accuracy: 0.6258
 Epoch 78/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6742 -
 accuracy: 0.6264
 Epoch 79/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6746 -
 accuracy: 0.6256
 Epoch 80/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6742 -
 accuracy: 0.6265
 Epoch 81/1000

16557/16557 [=====] - 1s 51us/sample - loss: 0.6745 -
 accuracy: 0.6263
 Epoch 82/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6745 -
 accuracy: 0.6258
 Epoch 83/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6743 -
 accuracy: 0.6264
 Epoch 84/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6736 -
 accuracy: 0.6283
 Epoch 85/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6733 -
 accuracy: 0.6287
 Epoch 86/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6735 -
 accuracy: 0.6278
 Epoch 87/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6736 -
 accuracy: 0.6277
 Epoch 88/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6741 -
 accuracy: 0.6267
 Epoch 89/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6737 -
 accuracy: 0.6284
 Epoch 90/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6734 -
 accuracy: 0.6286
 Epoch 91/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6729 -
 accuracy: 0.6302
 Epoch 92/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6731 -
 accuracy: 0.6294
 Epoch 93/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6733 -
 accuracy: 0.6290
 Epoch 94/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6741 -
 accuracy: 0.6281
 Epoch 95/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6740 -
 accuracy: 0.6273
 Epoch 96/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6730 -
 accuracy: 0.6293
 Epoch 97/1000

16557/16557 [=====] - 1s 50us/sample - loss: 0.6730 -
 accuracy: 0.6296
 Epoch 98/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6736 -
 accuracy: 0.6281
 Epoch 99/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6742 -
 accuracy: 0.6270
 Epoch 100/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6740 -
 accuracy: 0.6281
 Epoch 101/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6732 -
 accuracy: 0.6302
 Epoch 102/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6733 -
 accuracy: 0.6293
 Epoch 103/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6734 -
 accuracy: 0.6304
 Epoch 104/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6746 -
 accuracy: 0.6275
 Epoch 105/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6734 -
 accuracy: 0.6285
 Epoch 106/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6731 -
 accuracy: 0.6296
 Epoch 107/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6730 -
 accuracy: 0.6302
 Epoch 108/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6737 -
 accuracy: 0.6288
 Epoch 109/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6736 -
 accuracy: 0.6288
 Epoch 110/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6726 -
 accuracy: 0.6308
 Epoch 111/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6726 -
 accuracy: 0.6310
 Epoch 112/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6728 -
 accuracy: 0.6310
 Epoch 113/1000

16557/16557 [=====] - 1s 58us/sample - loss: 0.6729 -
 accuracy: 0.6302
 Epoch 114/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6729 -
 accuracy: 0.6302
 Epoch 115/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6730 -
 accuracy: 0.6301
 Epoch 116/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6732 -
 accuracy: 0.6298
 Epoch 117/1000
 16557/16557 [=====] - 1s 47us/sample - loss: 0.6729 -
 accuracy: 0.6305
 Epoch 118/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6726 -
 accuracy: 0.6316
 Epoch 119/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6728 -
 accuracy: 0.6309
 Epoch 120/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6730 -
 accuracy: 0.6300
 Epoch 121/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6725 -
 accuracy: 0.6312
 Epoch 122/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6722 -
 accuracy: 0.6313
 Epoch 123/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6724 -
 accuracy: 0.6306
 Epoch 124/1000
 16557/16557 [=====] - 1s 48us/sample - loss: 0.6722 -
 accuracy: 0.6311
 Epoch 125/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6717 -
 accuracy: 0.6328
 Epoch 126/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6716 -
 accuracy: 0.6328
 Epoch 127/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6715 -
 accuracy: 0.6331
 Epoch 128/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6722 -
 accuracy: 0.6322
 Epoch 129/1000

16557/16557 [=====] - 1s 49us/sample - loss: 0.6724 -
 accuracy: 0.6316
 Epoch 130/1000
 16557/16557 [=====] - 1s 49us/sample - loss: 0.6732 -
 accuracy: 0.6304
 Epoch 131/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6729 -
 accuracy: 0.6309
 Epoch 132/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6729 -
 accuracy: 0.6309
 Epoch 133/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6722 -
 accuracy: 0.6317
 Epoch 134/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6719 -
 accuracy: 0.6334
 Epoch 135/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6715 -
 accuracy: 0.6339
 Epoch 136/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6727 -
 accuracy: 0.6312
 Epoch 137/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6727 -
 accuracy: 0.6319
 Epoch 138/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6734 -
 accuracy: 0.6301
 Epoch 139/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6724 -
 accuracy: 0.6319
 Epoch 140/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6790 -
 accuracy: 0.6305
 Epoch 141/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6717 -
 accuracy: 0.6337
 Epoch 142/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6725 -
 accuracy: 0.6323
 Epoch 143/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6793 -
 accuracy: 0.6324
 Epoch 144/1000
 16557/16557 [=====] - 1s 50us/sample - loss: 0.6720 -
 accuracy: 0.6330
 Epoch 145/1000

16557/16557 [=====] - 1s 50us/sample - loss: 0.6720 -
 accuracy: 0.6335
 Epoch 146/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6721 -
 accuracy: 0.6325
 Epoch 147/1000
 16557/16557 [=====] - 1s 51us/sample - loss: 0.6714 -
 accuracy: 0.6342
 Epoch 148/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6715 -
 accuracy: 0.6344
 Epoch 149/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6724 -
 accuracy: 0.6313
 Epoch 150/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6726 -
 accuracy: 0.6307
 Epoch 151/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6719 -
 accuracy: 0.6325
 Epoch 152/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6715 -
 accuracy: 0.6339
 Epoch 153/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6712 -
 accuracy: 0.6347
 Epoch 154/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6715 -
 accuracy: 0.6339
 Epoch 155/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6710 -
 accuracy: 0.6353
 Epoch 156/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6711 -
 accuracy: 0.6346
 Epoch 157/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6715 -
 accuracy: 0.6340
 Epoch 158/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6716 -
 accuracy: 0.6341
 Epoch 159/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6716 -
 accuracy: 0.6344
 Epoch 160/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6711 -
 accuracy: 0.6352
 Epoch 161/1000

16557/16557 [=====] - 1s 53us/sample - loss: 0.6716 -
 accuracy: 0.6334
 Epoch 162/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6704 -
 accuracy: 0.6365
 Epoch 163/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6705 -
 accuracy: 0.6362
 Epoch 164/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6707 -
 accuracy: 0.6366
 Epoch 165/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6702 -
 accuracy: 0.6369
 Epoch 166/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6703 -
 accuracy: 0.6370
 Epoch 167/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6707 -
 accuracy: 0.6359
 Epoch 168/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6707 -
 accuracy: 0.6364
 Epoch 169/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6711 -
 accuracy: 0.6360
 Epoch 170/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6715 -
 accuracy: 0.6347
 Epoch 171/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6712 -
 accuracy: 0.6359
 Epoch 172/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6716 -
 accuracy: 0.6350
 Epoch 173/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6724 -
 accuracy: 0.6330
 Epoch 174/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6724 -
 accuracy: 0.6316
 Epoch 175/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6714 -
 accuracy: 0.6346
 Epoch 176/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6717 -
 accuracy: 0.6352
 Epoch 177/1000

16557/16557 [=====] - 1s 54us/sample - loss: 0.6711 -
 accuracy: 0.6363
 Epoch 178/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6720 -
 accuracy: 0.6346
 Epoch 179/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6718 -
 accuracy: 0.6341
 Epoch 180/1000
 16557/16557 [=====] - 1s 52us/sample - loss: 0.6715 -
 accuracy: 0.6343
 Epoch 181/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6712 -
 accuracy: 0.6348
 Epoch 182/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6706 -
 accuracy: 0.6368
 Epoch 183/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6702 -
 accuracy: 0.6376
 Epoch 184/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6712 -
 accuracy: 0.6359
 Epoch 185/1000
 16557/16557 [=====] - 1s 53us/sample - loss: 0.6719 -
 accuracy: 0.6351
 Epoch 186/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6724 -
 accuracy: 0.6344
 Epoch 187/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6732 -
 accuracy: 0.6326
 Epoch 188/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6721 -
 accuracy: 0.6345
 Epoch 189/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6715 -
 accuracy: 0.6356
 Epoch 190/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6712 -
 accuracy: 0.6356
 Epoch 191/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6715 -
 accuracy: 0.6344
 Epoch 192/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6716 -
 accuracy: 0.6343
 Epoch 193/1000

16557/16557 [=====] - 1s 56us/sample - loss: 0.6717 -
 accuracy: 0.6348
 Epoch 194/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6710 -
 accuracy: 0.6357
 Epoch 195/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6705 -
 accuracy: 0.6369
 Epoch 196/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6703 -
 accuracy: 0.6374
 Epoch 197/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6703 -
 accuracy: 0.6374
 Epoch 198/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6700 -
 accuracy: 0.6380
 Epoch 199/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6697 -
 accuracy: 0.6385
 Epoch 200/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6698 -
 accuracy: 0.6380
 Epoch 201/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6709 -
 accuracy: 0.6361
 Epoch 202/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6716 -
 accuracy: 0.6340
 Epoch 203/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6703 -
 accuracy: 0.6376
 Epoch 204/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6703 -
 accuracy: 0.6377
 Epoch 205/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6708 -
 accuracy: 0.6371
 Epoch 206/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6703 -
 accuracy: 0.6376
 Epoch 207/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6697 -
 accuracy: 0.6387
 Epoch 208/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6696 -
 accuracy: 0.6391
 Epoch 209/1000

16557/16557 [=====] - 1s 58us/sample - loss: 0.6703 -
 accuracy: 0.6370
 Epoch 210/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6698 -
 accuracy: 0.6388
 Epoch 211/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6696 -
 accuracy: 0.6386
 Epoch 212/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6699 -
 accuracy: 0.6383
 Epoch 213/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6708 -
 accuracy: 0.6366
 Epoch 214/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6697 -
 accuracy: 0.6388
 Epoch 215/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6703 -
 accuracy: 0.6379
 Epoch 216/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6698 -
 accuracy: 0.6385
 Epoch 217/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6699 -
 accuracy: 0.6389
 Epoch 218/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6707 -
 accuracy: 0.6372
 Epoch 219/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6703 -
 accuracy: 0.6381
 Epoch 220/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6695 -
 accuracy: 0.6396
 Epoch 221/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6700 -
 accuracy: 0.6385
 Epoch 222/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6702 -
 accuracy: 0.6384
 Epoch 223/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6697 -
 accuracy: 0.6391
 Epoch 224/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6700 -
 accuracy: 0.6388
 Epoch 225/1000

16557/16557 [=====] - 1s 56us/sample - loss: 0.6699 -
 accuracy: 0.6391
 Epoch 226/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6695 -
 accuracy: 0.6399
 Epoch 227/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6692 -
 accuracy: 0.6405
 Epoch 228/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6693 -
 accuracy: 0.6400
 Epoch 229/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6697 -
 accuracy: 0.6395
 Epoch 230/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6699 -
 accuracy: 0.6389
 Epoch 231/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6707 -
 accuracy: 0.6369
 Epoch 232/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6703 -
 accuracy: 0.6380
 Epoch 233/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6700 -
 accuracy: 0.6389
 Epoch 234/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6698 -
 accuracy: 0.6391
 Epoch 235/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6702 -
 accuracy: 0.6380
 Epoch 236/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6694 -
 accuracy: 0.6400
 Epoch 237/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6710 -
 accuracy: 0.6365
 Epoch 238/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6706 -
 accuracy: 0.6366
 Epoch 239/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6697 -
 accuracy: 0.6389
 Epoch 240/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6697 -
 accuracy: 0.6389
 Epoch 241/1000

16557/16557 [=====] - 1s 54us/sample - loss: 0.6694 -
 accuracy: 0.6399
 Epoch 242/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6694 -
 accuracy: 0.6395
 Epoch 243/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6689 -
 accuracy: 0.6405
 Epoch 244/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6694 -
 accuracy: 0.6397
 Epoch 245/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6695 -
 accuracy: 0.6395
 Epoch 246/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6696 -
 accuracy: 0.6393
 Epoch 247/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6697 -
 accuracy: 0.6390
 Epoch 248/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6702 -
 accuracy: 0.6373
 Epoch 249/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6702 -
 accuracy: 0.6378
 Epoch 250/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6700 -
 accuracy: 0.6388
 Epoch 251/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6699 -
 accuracy: 0.6385
 Epoch 252/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6693 -
 accuracy: 0.6399
 Epoch 253/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6692 -
 accuracy: 0.6399
 Epoch 254/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6693 -
 accuracy: 0.6399
 Epoch 255/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6691 -
 accuracy: 0.6405
 Epoch 256/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6697 -
 accuracy: 0.6392
 Epoch 257/1000

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16557/16557 [=====] - 1s 55us/sample - loss: 0.6695 -
accuracy: 0.6395
Epoch 258/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6694 -
accuracy: 0.6398
Epoch 259/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6693 -
accuracy: 0.6401
Epoch 260/1000
16557/16557 [=====] - 1s 54us/sample - loss: 0.6692 -
accuracy: 0.6404
Epoch 261/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6693 -
accuracy: 0.6405
Epoch 262/1000
16557/16557 [=====] - 1s 54us/sample - loss: 0.6698 -
accuracy: 0.6386
Epoch 263/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6698 -
accuracy: 0.6386
Epoch 264/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6694 -
accuracy: 0.6402
Epoch 265/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6690 -
accuracy: 0.6409
Epoch 266/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6688 -
accuracy: 0.6415
Epoch 267/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6687 -
accuracy: 0.6415
Epoch 268/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6693 -
accuracy: 0.6406
Epoch 269/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6695 -
accuracy: 0.6403
Epoch 270/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6701 -
accuracy: 0.6388
Epoch 271/1000
16557/16557 [=====] - 1s 54us/sample - loss: 0.6699 -
accuracy: 0.6381
Epoch 272/1000
16557/16557 [=====] - 1s 61us/sample - loss: 0.6693 -
accuracy: 0.6399
Epoch 273/1000

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16557/16557 [=====] - 1s 54us/sample - loss: 0.6702 -
 accuracy: 0.6376
 Epoch 274/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6690 -
 accuracy: 0.6409
 Epoch 275/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6686 -
 accuracy: 0.6417
 Epoch 276/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6687 -
 accuracy: 0.6418
 Epoch 277/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6689 -
 accuracy: 0.6409
 Epoch 278/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6693 -
 accuracy: 0.6407
 Epoch 279/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6690 -
 accuracy: 0.6406
 Epoch 280/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6686 -
 accuracy: 0.6417
 Epoch 281/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6690 -
 accuracy: 0.6408
 Epoch 282/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6687 -
 accuracy: 0.6414
 Epoch 283/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6690 -
 accuracy: 0.6404
 Epoch 284/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6691 -
 accuracy: 0.6410
 Epoch 285/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6691 -
 accuracy: 0.6410
 Epoch 286/1000
 16557/16557 [=====] - 1s 54us/sample - loss: 0.6694 -
 accuracy: 0.6399
 Epoch 287/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6699 -
 accuracy: 0.6389
 Epoch 288/1000
 16557/16557 [=====] - 1s 55us/sample - loss: 0.6694 -
 accuracy: 0.6397
 Epoch 289/1000

```

16557/16557 [=====] - 1s 54us/sample - loss: 0.6698 -
accuracy: 0.6392
Epoch 290/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6695 -
accuracy: 0.6400
Epoch 291/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6689 -
accuracy: 0.6409
Epoch 292/1000
16557/16557 [=====] - 1s 54us/sample - loss: 0.6690 -
accuracy: 0.6407
Epoch 293/1000
16557/16557 [=====] - 1s 55us/sample - loss: 0.6688 -
accuracy: 0.6408
Epoch 294/1000
16557/16557 [=====] - 1s 54us/sample - loss: 0.6681 -
accuracy: 0.6431
Epoch 295/1000
16557/16557 [=====] - 1s 54us/sample - loss: 0.6682 -
accuracy: 0.6425
Epoch 296/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6684 -
accuracy: 0.6420
Epoch 297/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6687 -
accuracy: 0.6410
Epoch 298/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6678 -
accuracy: 0.6432
Epoch 299/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6686 -
accuracy: 0.6418
Epoch 300/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6684 -
accuracy: 0.6423
Epoch 301/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6682 -
accuracy: 0.6425
Epoch 302/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6683 -
accuracy: 0.6422
Epoch 303/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6686 -
accuracy: 0.6413
Epoch 304/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6692 -
accuracy: 0.6402
Epoch 305/1000

```

16557/16557 [=====] - 1s 56us/sample - loss: 0.6683 -
accuracy: 0.6420
Epoch 306/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6685 -
accuracy: 0.6417
Epoch 307/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6678 -
accuracy: 0.6433
Epoch 308/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6683 -
accuracy: 0.6419
Epoch 309/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6686 -
accuracy: 0.6412
Epoch 310/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6690 -
accuracy: 0.6402
Epoch 311/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6684 -
accuracy: 0.6418
Epoch 312/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6680 -
accuracy: 0.6426
Epoch 313/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6682 -
accuracy: 0.6423
Epoch 314/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6683 -
accuracy: 0.6421
Epoch 315/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6682 -
accuracy: 0.6421
Epoch 316/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6686 -
accuracy: 0.6417
Epoch 317/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6682 -
accuracy: 0.6424
Epoch 318/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6684 -
accuracy: 0.6419
Epoch 319/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6686 -
accuracy: 0.6409
Epoch 320/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6681 -
accuracy: 0.6423
Epoch 321/1000

16557/16557 [=====] - 1s 56us/sample - loss: 0.6676 -
 accuracy: 0.6437
 Epoch 322/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6683 -
 accuracy: 0.6418
 Epoch 323/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6687 -
 accuracy: 0.6413
 Epoch 324/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6683 -
 accuracy: 0.6420
 Epoch 325/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6682 -
 accuracy: 0.6424
 Epoch 326/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6685 -
 accuracy: 0.6415
 Epoch 327/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6690 -
 accuracy: 0.6403
 Epoch 328/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6686 -
 accuracy: 0.6409
 Epoch 329/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6679 -
 accuracy: 0.6426
 Epoch 330/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6682 -
 accuracy: 0.6415
 Epoch 331/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6685 -
 accuracy: 0.6411
 Epoch 332/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6680 -
 accuracy: 0.6423
 Epoch 333/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6671 -
 accuracy: 0.6443
 Epoch 334/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6679 -
 accuracy: 0.6427
 Epoch 335/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6677 -
 accuracy: 0.6432
 Epoch 336/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6675 -
 accuracy: 0.6437
 Epoch 337/1000

16557/16557 [=====] - 1s 56us/sample - loss: 0.6679 -
 accuracy: 0.6429
 Epoch 338/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6675 -
 accuracy: 0.6438
 Epoch 339/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6681 -
 accuracy: 0.6428
 Epoch 340/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6681 -
 accuracy: 0.6429
 Epoch 341/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6676 -
 accuracy: 0.6440
 Epoch 342/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6679 -
 accuracy: 0.6429
 Epoch 343/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6682 -
 accuracy: 0.6433
 Epoch 344/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6675 -
 accuracy: 0.6438
 Epoch 345/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6671 -
 accuracy: 0.6444
 Epoch 346/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6677 -
 accuracy: 0.6434
 Epoch 347/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6693 -
 accuracy: 0.6412
 Epoch 348/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6683 -
 accuracy: 0.6429
 Epoch 349/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6677 -
 accuracy: 0.6441
 Epoch 350/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6683 -
 accuracy: 0.6431
 Epoch 351/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6679 -
 accuracy: 0.6436
 Epoch 352/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6677 -
 accuracy: 0.6436
 Epoch 353/1000

16557/16557 [=====] - 1s 59us/sample - loss: 0.6674 -
 accuracy: 0.6440
 Epoch 354/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6676 -
 accuracy: 0.6437
 Epoch 355/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6676 -
 accuracy: 0.6439
 Epoch 356/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6692 -
 accuracy: 0.6414
 Epoch 357/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6679 -
 accuracy: 0.6438
 Epoch 358/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6677 -
 accuracy: 0.6435
 Epoch 359/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6681 -
 accuracy: 0.6433
 Epoch 360/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6673 -
 accuracy: 0.6444
 Epoch 361/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6672 -
 accuracy: 0.6446
 Epoch 362/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6671 -
 accuracy: 0.6449
 Epoch 363/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6673 -
 accuracy: 0.6447
 Epoch 364/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6676 -
 accuracy: 0.6440
 Epoch 365/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6679 -
 accuracy: 0.6430
 Epoch 366/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6687 -
 accuracy: 0.6419
 Epoch 367/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6695 -
 accuracy: 0.6400
 Epoch 368/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6684 -
 accuracy: 0.6429
 Epoch 369/1000

16557/16557 [=====] - 1s 58us/sample - loss: 0.6683 -
 accuracy: 0.6426
 Epoch 370/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6678 -
 accuracy: 0.6435
 Epoch 371/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6673 -
 accuracy: 0.6445
 Epoch 372/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6672 -
 accuracy: 0.6448
 Epoch 373/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6672 -
 accuracy: 0.6446
 Epoch 374/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6670 -
 accuracy: 0.6451
 Epoch 375/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6667 -
 accuracy: 0.6458
 Epoch 376/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6669 -
 accuracy: 0.6450
 Epoch 377/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6678 -
 accuracy: 0.6432
 Epoch 378/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6675 -
 accuracy: 0.6446
 Epoch 379/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6673 -
 accuracy: 0.6450
 Epoch 380/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6673 -
 accuracy: 0.6450
 Epoch 381/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6668 -
 accuracy: 0.6458
 Epoch 382/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6667 -
 accuracy: 0.6460
 Epoch 383/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6673 -
 accuracy: 0.6445
 Epoch 384/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6676 -
 accuracy: 0.6444
 Epoch 385/1000

```

16557/16557 [=====] - 1s 57us/sample - loss: 0.6682 -
accuracy: 0.6427
Epoch 386/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6677 -
accuracy: 0.6444
Epoch 387/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6666 -
accuracy: 0.6466
Epoch 388/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6676 -
accuracy: 0.6443
Epoch 389/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6680 -
accuracy: 0.6438
Epoch 390/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6674 -
accuracy: 0.6445
Epoch 391/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6670 -
accuracy: 0.6458
Epoch 392/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6674 -
accuracy: 0.6447
Epoch 393/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6668 -
accuracy: 0.6462
Epoch 394/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6669 -
accuracy: 0.6461
Epoch 395/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6670 -
accuracy: 0.6455
Epoch 396/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6688 -
accuracy: 0.6408
Epoch 397/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6686 -
accuracy: 0.6407
Epoch 398/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6674 -
accuracy: 0.6442
Epoch 399/1000
16557/16557 [=====] - 1s 57us/sample - loss: 0.6674 -
accuracy: 0.6437
Epoch 400/1000
16557/16557 [=====] - 1s 56us/sample - loss: 0.6667 -
accuracy: 0.6455
Epoch 401/1000

```

16557/16557 [=====] - 1s 57us/sample - loss: 0.6667 -
 accuracy: 0.6463
 Epoch 402/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6667 -
 accuracy: 0.6458
 Epoch 403/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6663 -
 accuracy: 0.6469
 Epoch 404/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6666 -
 accuracy: 0.6466
 Epoch 405/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6671 -
 accuracy: 0.6455
 Epoch 406/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6666 -
 accuracy: 0.6459
 Epoch 407/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6665 -
 accuracy: 0.6466
 Epoch 408/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6672 -
 accuracy: 0.6456
 Epoch 409/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6672 -
 accuracy: 0.6452
 Epoch 410/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6667 -
 accuracy: 0.6465
 Epoch 411/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6668 -
 accuracy: 0.6460
 Epoch 412/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6665 -
 accuracy: 0.6463
 Epoch 413/1000
 16557/16557 [=====] - 1s 56us/sample - loss: 0.6663 -
 accuracy: 0.6471
 Epoch 414/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6666 -
 accuracy: 0.6466
 Epoch 415/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6668 -
 accuracy: 0.6457
 Epoch 416/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6670 -
 accuracy: 0.6456
 Epoch 417/1000

16557/16557 [=====] - 1s 57us/sample - loss: 0.6673 -
 accuracy: 0.6452
 Epoch 418/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6666 -
 accuracy: 0.6463
 Epoch 419/1000
 16557/16557 [=====] - 1s 57us/sample - loss: 0.6668 -
 accuracy: 0.6461
 Epoch 420/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6673 -
 accuracy: 0.6450
 Epoch 421/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6671 -
 accuracy: 0.6454
 Epoch 422/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6667 -
 accuracy: 0.6461
 Epoch 423/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6666 -
 accuracy: 0.6460
 Epoch 424/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6663 -
 accuracy: 0.6472
 Epoch 425/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6670 -
 accuracy: 0.6458
 Epoch 426/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6675 -
 accuracy: 0.6441
 Epoch 427/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6665 -
 accuracy: 0.6464
 Epoch 428/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6663 -
 accuracy: 0.6469
 Epoch 429/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6668 -
 accuracy: 0.6458
 Epoch 430/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6666 -
 accuracy: 0.6463
 Epoch 431/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6664 -
 accuracy: 0.6470
 Epoch 432/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6660 -
 accuracy: 0.6475
 Epoch 433/1000

```

16557/16557 [=====] - 1s 58us/sample - loss: 0.6667 -
accuracy: 0.6461
Epoch 434/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6662 -
accuracy: 0.6468
Epoch 435/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6662 -
accuracy: 0.6476
Epoch 436/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6659 -
accuracy: 0.6481
Epoch 437/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6670 -
accuracy: 0.6452
Epoch 438/1000
16557/16557 [=====] - 1s 60us/sample - loss: 0.6675 -
accuracy: 0.6439
Epoch 439/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6669 -
accuracy: 0.6457
Epoch 440/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6659 -
accuracy: 0.6479
Epoch 441/1000
16557/16557 [=====] - 1s 59us/sample - loss: 0.6659 -
accuracy: 0.6478
Epoch 442/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6656 -
accuracy: 0.6487
Epoch 443/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6661 -
accuracy: 0.6475
Epoch 444/1000
16557/16557 [=====] - 1s 60us/sample - loss: 0.6668 -
accuracy: 0.6456
Epoch 445/1000
16557/16557 [=====] - 1s 66us/sample - loss: 0.6670 -
accuracy: 0.6449
Epoch 446/1000
16557/16557 [=====] - 1s 61us/sample - loss: 0.6668 -
accuracy: 0.6459
Epoch 447/1000
16557/16557 [=====] - 1s 60us/sample - loss: 0.6665 -
accuracy: 0.6472
Epoch 448/1000
16557/16557 [=====] - 1s 63us/sample - loss: 0.6669 -
accuracy: 0.6463
Epoch 449/1000

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16557/16557 [=====] - 1s 62us/sample - loss: 0.6666 -
accuracy: 0.6467
Epoch 450/1000
16557/16557 [=====] - 1s 62us/sample - loss: 0.6664 -
accuracy: 0.6475
Epoch 451/1000
16557/16557 [=====] - 1s 62us/sample - loss: 0.6664 -
accuracy: 0.6472
Epoch 452/1000
16557/16557 [=====] - 1s 62us/sample - loss: 0.6656 -
accuracy: 0.6485
Epoch 453/1000
16557/16557 [=====] - 1s 62us/sample - loss: 0.6655 -
accuracy: 0.6488
Epoch 454/1000
16557/16557 [=====] - 1s 62us/sample - loss: 0.6655 -
accuracy: 0.6490
Epoch 455/1000
16557/16557 [=====] - 1s 61us/sample - loss: 0.6660 -
accuracy: 0.6479
Epoch 456/1000
16557/16557 [=====] - 1s 62us/sample - loss: 0.6664 -
accuracy: 0.6476
Epoch 457/1000
16557/16557 [=====] - 1s 71us/sample - loss: 0.6660 -
accuracy: 0.6483
Epoch 458/1000
16557/16557 [=====] - 1s 64us/sample - loss: 0.6657 -
accuracy: 0.6490
Epoch 459/1000
16557/16557 [=====] - 1s 63us/sample - loss: 0.6660 -
accuracy: 0.6482
Epoch 460/1000
16557/16557 [=====] - 1s 63us/sample - loss: 0.6656 -
accuracy: 0.6488
Epoch 461/1000
16557/16557 [=====] - 1s 62us/sample - loss: 0.6660 -
accuracy: 0.6482
Epoch 462/1000
16557/16557 [=====] - 1s 63us/sample - loss: 0.6664 -
accuracy: 0.6469
Epoch 463/1000
16557/16557 [=====] - 1s 63us/sample - loss: 0.6664 -
accuracy: 0.6473
Epoch 464/1000
16557/16557 [=====] - 1s 63us/sample - loss: 0.6660 -
accuracy: 0.6484
Epoch 465/1000

```

16557/16557 [=====] - 1s 63us/sample - loss: 0.6662 -
 accuracy: 0.6477
 Epoch 466/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6663 -
 accuracy: 0.6476
 Epoch 467/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6663 -
 accuracy: 0.6478
 Epoch 468/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6659 -
 accuracy: 0.6484
 Epoch 469/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6670 -
 accuracy: 0.6460
 Epoch 470/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6665 -
 accuracy: 0.6470
 Epoch 471/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6656 -
 accuracy: 0.6490
 Epoch 472/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6655 -
 accuracy: 0.6492
 Epoch 473/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6665 -
 accuracy: 0.6477
 Epoch 474/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6666 -
 accuracy: 0.6470
 Epoch 475/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6670 -
 accuracy: 0.6458
 Epoch 476/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6665 -
 accuracy: 0.6472
 Epoch 477/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6667 -
 accuracy: 0.6472
 Epoch 478/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6667 -
 accuracy: 0.6471
 Epoch 479/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6663 -
 accuracy: 0.6481
 Epoch 480/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6660 -
 accuracy: 0.6482
 Epoch 481/1000

16557/16557 [=====] - 1s 63us/sample - loss: 0.6656 -
 accuracy: 0.6493
 Epoch 482/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6650 -
 accuracy: 0.6503
 Epoch 483/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6654 -
 accuracy: 0.6497
 Epoch 484/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6652 -
 accuracy: 0.6503
 Epoch 485/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6653 -
 accuracy: 0.6496
 Epoch 486/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6661 -
 accuracy: 0.6481
 Epoch 487/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6657 -
 accuracy: 0.6493
 Epoch 488/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6659 -
 accuracy: 0.6487
 Epoch 489/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6655 -
 accuracy: 0.6494
 Epoch 490/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6655 -
 accuracy: 0.6498
 Epoch 491/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6660 -
 accuracy: 0.6490
 Epoch 492/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6663 -
 accuracy: 0.6492
 Epoch 493/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6656 -
 accuracy: 0.6492
 Epoch 494/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6657 -
 accuracy: 0.6494
 Epoch 495/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6653 -
 accuracy: 0.6498
 Epoch 496/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6649 -
 accuracy: 0.6502
 Epoch 497/1000

16557/16557 [=====] - 1s 62us/sample - loss: 0.6653 -
 accuracy: 0.6495
 Epoch 498/1000
 16557/16557 [=====] - 1s 72us/sample - loss: 0.6657 -
 accuracy: 0.6495
 Epoch 499/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6655 -
 accuracy: 0.6499
 Epoch 500/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6648 -
 accuracy: 0.6510
 Epoch 501/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6655 -
 accuracy: 0.6493
 Epoch 502/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6660 -
 accuracy: 0.6485
 Epoch 503/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6654 -
 accuracy: 0.6496
 Epoch 504/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6652 -
 accuracy: 0.6505
 Epoch 505/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6652 -
 accuracy: 0.6504
 Epoch 506/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6661 -
 accuracy: 0.6479
 Epoch 507/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6660 -
 accuracy: 0.6484
 Epoch 508/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6654 -
 accuracy: 0.6496
 Epoch 509/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6654 -
 accuracy: 0.6497
 Epoch 510/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6650 -
 accuracy: 0.6504
 Epoch 511/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6650 -
 accuracy: 0.6505
 Epoch 512/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6649 -
 accuracy: 0.6504
 Epoch 513/1000

16557/16557 [=====] - 1s 62us/sample - loss: 0.6654 -
 accuracy: 0.6499
 Epoch 514/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6667 -
 accuracy: 0.6467
 Epoch 515/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6663 -
 accuracy: 0.6477
 Epoch 516/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6661 -
 accuracy: 0.6479
 Epoch 517/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6657 -
 accuracy: 0.6490
 Epoch 518/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6661 -
 accuracy: 0.6484
 Epoch 519/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6660 -
 accuracy: 0.6498
 Epoch 520/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6655 -
 accuracy: 0.6504
 Epoch 521/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6653 -
 accuracy: 0.6500
 Epoch 522/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6655 -
 accuracy: 0.6504
 Epoch 523/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6656 -
 accuracy: 0.6500
 Epoch 524/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6661 -
 accuracy: 0.6500
 Epoch 525/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6666 -
 accuracy: 0.6486
 Epoch 526/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6661 -
 accuracy: 0.6493
 Epoch 527/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6648 -
 accuracy: 0.6508
 Epoch 528/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6654 -
 accuracy: 0.6502
 Epoch 529/1000

16557/16557 [=====] - 1s 60us/sample - loss: 0.6661 -
 accuracy: 0.6498
 Epoch 530/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6658 -
 accuracy: 0.6494
 Epoch 531/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6655 -
 accuracy: 0.6494
 Epoch 532/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6651 -
 accuracy: 0.6506
 Epoch 533/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6653 -
 accuracy: 0.6505
 Epoch 534/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6644 -
 accuracy: 0.6522
 Epoch 535/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6653 -
 accuracy: 0.6506
 Epoch 536/1000
 16557/16557 [=====] - 1s 58us/sample - loss: 0.6652 -
 accuracy: 0.6509
 Epoch 537/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6654 -
 accuracy: 0.6499
 Epoch 538/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6658 -
 accuracy: 0.6497
 Epoch 539/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6648 -
 accuracy: 0.6517
 Epoch 540/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6653 -
 accuracy: 0.6498
 Epoch 541/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6663 -
 accuracy: 0.6474
 Epoch 542/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6661 -
 accuracy: 0.6488
 Epoch 543/1000
 16557/16557 [=====] - 1s 59us/sample - loss: 0.6656 -
 accuracy: 0.6496
 Epoch 544/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6649 -
 accuracy: 0.6514
 Epoch 545/1000

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16557/16557 [=====] - 1s 59us/sample - loss: 0.6648 -
accuracy: 0.6515
Epoch 546/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6644 -
accuracy: 0.6525
Epoch 547/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6644 -
accuracy: 0.6524
Epoch 548/1000
16557/16557 [=====] - 1s 59us/sample - loss: 0.6646 -
accuracy: 0.6519
Epoch 549/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6651 -
accuracy: 0.6502
Epoch 550/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6651 -
accuracy: 0.6504
Epoch 551/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6653 -
accuracy: 0.6496
Epoch 552/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6655 -
accuracy: 0.6499
Epoch 553/1000
16557/16557 [=====] - 1s 59us/sample - loss: 0.6652 -
accuracy: 0.6507
Epoch 554/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6648 -
accuracy: 0.6516
Epoch 555/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6644 -
accuracy: 0.6518
Epoch 556/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6643 -
accuracy: 0.6526
Epoch 557/1000
16557/16557 [=====] - 1s 58us/sample - loss: 0.6637 -
accuracy: 0.6534
Epoch 558/1000
16557/16557 [=====] - 1s 60us/sample - loss: 0.6642 -
accuracy: 0.6522
Epoch 559/1000
16557/16557 [=====] - 1s 61us/sample - loss: 0.6639 -
accuracy: 0.6528
Epoch 560/1000
16557/16557 [=====] - 1s 60us/sample - loss: 0.6647 -
accuracy: 0.6514
Epoch 561/1000

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16557/16557 [=====] - 1s 60us/sample - loss: 0.6651 -
 accuracy: 0.6507
 Epoch 562/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6677 -
 accuracy: 0.6465
 Epoch 563/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6654 -
 accuracy: 0.6507
 Epoch 564/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6649 -
 accuracy: 0.6517
 Epoch 565/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6660 -
 accuracy: 0.6488
 Epoch 566/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6649 -
 accuracy: 0.6512
 Epoch 567/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6652 -
 accuracy: 0.6507
 Epoch 568/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6647 -
 accuracy: 0.6519
 Epoch 569/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6644 -
 accuracy: 0.6527
 Epoch 570/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6652 -
 accuracy: 0.6508
 Epoch 571/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6651 -
 accuracy: 0.6507
 Epoch 572/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6658 -
 accuracy: 0.6494
 Epoch 573/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6649 -
 accuracy: 0.6517
 Epoch 574/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6643 -
 accuracy: 0.6527
 Epoch 575/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6636 -
 accuracy: 0.6537
 Epoch 576/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6642 -
 accuracy: 0.6527
 Epoch 577/1000

16557/16557 [=====] - 1s 63us/sample - loss: 0.6648 -
 accuracy: 0.6514
 Epoch 578/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6664 -
 accuracy: 0.6470
 Epoch 579/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6656 -
 accuracy: 0.6493
 Epoch 580/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6652 -
 accuracy: 0.6501
 Epoch 581/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6647 -
 accuracy: 0.6515
 Epoch 582/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6652 -
 accuracy: 0.6502
 Epoch 583/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6652 -
 accuracy: 0.6505
 Epoch 584/1000
 16557/16557 [=====] - 1s 75us/sample - loss: 0.6647 -
 accuracy: 0.6514
 Epoch 585/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6643 -
 accuracy: 0.6524
 Epoch 586/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6642 -
 accuracy: 0.6528
 Epoch 587/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6644 -
 accuracy: 0.6528
 Epoch 588/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6644 -
 accuracy: 0.6521
 Epoch 589/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6637 -
 accuracy: 0.6538
 Epoch 590/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6634 -
 accuracy: 0.6545
 Epoch 591/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6645 -
 accuracy: 0.6522
 Epoch 592/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6650 -
 accuracy: 0.6512
 Epoch 593/1000

16557/16557 [=====] - 1s 68us/sample - loss: 0.6648 -
 accuracy: 0.6511
 Epoch 594/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6649 -
 accuracy: 0.6510
 Epoch 595/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6649 -
 accuracy: 0.6514
 Epoch 596/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6647 -
 accuracy: 0.6519
 Epoch 597/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6642 -
 accuracy: 0.6532
 Epoch 598/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6640 -
 accuracy: 0.6535
 Epoch 599/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6635 -
 accuracy: 0.6547
 Epoch 600/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6640 -
 accuracy: 0.6536
 Epoch 601/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6640 -
 accuracy: 0.6536
 Epoch 602/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6641 -
 accuracy: 0.6533
 Epoch 603/1000
 16557/16557 [=====] - 1s 71us/sample - loss: 0.6636 -
 accuracy: 0.6545
 Epoch 604/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6640 -
 accuracy: 0.6533
 Epoch 605/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6641 -
 accuracy: 0.6528
 Epoch 606/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6649 -
 accuracy: 0.6521
 Epoch 607/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6643 -
 accuracy: 0.6528
 Epoch 608/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6642 -
 accuracy: 0.6529
 Epoch 609/1000

16557/16557 [=====] - 1s 66us/sample - loss: 0.6639 -
 accuracy: 0.6537
 Epoch 610/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6642 -
 accuracy: 0.6531
 Epoch 611/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6645 -
 accuracy: 0.6529
 Epoch 612/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6642 -
 accuracy: 0.6533
 Epoch 613/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6642 -
 accuracy: 0.6528
 Epoch 614/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6642 -
 accuracy: 0.6531
 Epoch 615/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6640 -
 accuracy: 0.6533
 Epoch 616/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6639 -
 accuracy: 0.6539
 Epoch 617/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6636 -
 accuracy: 0.6546
 Epoch 618/1000
 16557/16557 [=====] - 1s 73us/sample - loss: 0.6644 -
 accuracy: 0.6531
 Epoch 619/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6641 -
 accuracy: 0.6530
 Epoch 620/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6638 -
 accuracy: 0.6539
 Epoch 621/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6637 -
 accuracy: 0.6542
 Epoch 622/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6635 -
 accuracy: 0.6542
 Epoch 623/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6637 -
 accuracy: 0.6535
 Epoch 624/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6632 -
 accuracy: 0.6554
 Epoch 625/1000

16557/16557 [=====] - 1s 63us/sample - loss: 0.6634 -
 accuracy: 0.6545
 Epoch 626/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6638 -
 accuracy: 0.6533
 Epoch 627/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6644 -
 accuracy: 0.6526
 Epoch 628/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6645 -
 accuracy: 0.6529
 Epoch 629/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6645 -
 accuracy: 0.6533
 Epoch 630/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6641 -
 accuracy: 0.6537
 Epoch 631/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6641 -
 accuracy: 0.6536
 Epoch 632/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6638 -
 accuracy: 0.6541
 Epoch 633/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6634 -
 accuracy: 0.6548
 Epoch 634/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6640 -
 accuracy: 0.6537
 Epoch 635/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6640 -
 accuracy: 0.6538
 Epoch 636/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6636 -
 accuracy: 0.6548
 Epoch 637/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6635 -
 accuracy: 0.6543
 Epoch 638/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6635 -
 accuracy: 0.6543
 Epoch 639/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6643 -
 accuracy: 0.6520
 Epoch 640/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6643 -
 accuracy: 0.6525
 Epoch 641/1000

16557/16557 [=====] - 1s 60us/sample - loss: 0.6639 -
 accuracy: 0.6531
 Epoch 642/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6636 -
 accuracy: 0.6543
 Epoch 643/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6630 -
 accuracy: 0.6556
 Epoch 644/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6626 -
 accuracy: 0.6565
 Epoch 645/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6629 -
 accuracy: 0.6559
 Epoch 646/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6633 -
 accuracy: 0.6551
 Epoch 647/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6632 -
 accuracy: 0.6556
 Epoch 648/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6630 -
 accuracy: 0.6556
 Epoch 649/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6642 -
 accuracy: 0.6525
 Epoch 650/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6631 -
 accuracy: 0.6549
 Epoch 651/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6645 -
 accuracy: 0.6516
 Epoch 652/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6639 -
 accuracy: 0.6539
 Epoch 653/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6638 -
 accuracy: 0.6540
 Epoch 654/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6645 -
 accuracy: 0.6530
 Epoch 655/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6646 -
 accuracy: 0.6524
 Epoch 656/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6637 -
 accuracy: 0.6543
 Epoch 657/1000

16557/16557 [=====] - 1s 61us/sample - loss: 0.6632 -
 accuracy: 0.6554
 Epoch 658/1000
 16557/16557 [=====] - 1s 73us/sample - loss: 0.6634 -
 accuracy: 0.6546
 Epoch 659/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6628 -
 accuracy: 0.6560
 Epoch 660/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6637 -
 accuracy: 0.6535
 Epoch 661/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6649 -
 accuracy: 0.6519
 Epoch 662/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6646 -
 accuracy: 0.6532
 Epoch 663/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6641 -
 accuracy: 0.6543
 Epoch 664/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6639 -
 accuracy: 0.6536
 Epoch 665/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6640 -
 accuracy: 0.6533
 Epoch 666/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6641 -
 accuracy: 0.6537
 Epoch 667/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6636 -
 accuracy: 0.6549
 Epoch 668/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6636 -
 accuracy: 0.6553
 Epoch 669/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6630 -
 accuracy: 0.6559
 Epoch 670/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6638 -
 accuracy: 0.6539
 Epoch 671/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6636 -
 accuracy: 0.6545
 Epoch 672/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6638 -
 accuracy: 0.6533
 Epoch 673/1000

16557/16557 [=====] - 1s 60us/sample - loss: 0.6634 -
 accuracy: 0.6549
 Epoch 674/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6629 -
 accuracy: 0.6560
 Epoch 675/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6634 -
 accuracy: 0.6548
 Epoch 676/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6632 -
 accuracy: 0.6556
 Epoch 677/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6647 -
 accuracy: 0.6513
 Epoch 678/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6633 -
 accuracy: 0.6546
 Epoch 679/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6633 -
 accuracy: 0.6552
 Epoch 680/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6640 -
 accuracy: 0.6539
 Epoch 681/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6647 -
 accuracy: 0.6516
 Epoch 682/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6636 -
 accuracy: 0.6538
 Epoch 683/1000
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6634 -
 accuracy: 0.6545
 Epoch 684/1000
 16557/16557 [=====] - 1s 61us/sample - loss: 0.6636 -
 accuracy: 0.6543
 Epoch 685/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6633 -
 accuracy: 0.6551
 Epoch 686/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6629 -
 accuracy: 0.6556
 Epoch 687/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6638 -
 accuracy: 0.6533
 Epoch 688/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6642 -
 accuracy: 0.6521
 Epoch 689/1000

16557/16557 [=====] - 1s 64us/sample - loss: 0.6640 -
 accuracy: 0.6531
 Epoch 690/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6650 -
 accuracy: 0.6502
 Epoch 691/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6638 -
 accuracy: 0.6534
 Epoch 692/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6632 -
 accuracy: 0.6553
 Epoch 693/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6636 -
 accuracy: 0.6546
 Epoch 694/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6638 -
 accuracy: 0.6543
 Epoch 695/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6639 -
 accuracy: 0.6534
 Epoch 696/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6635 -
 accuracy: 0.6546
 Epoch 697/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6637 -
 accuracy: 0.6546
 Epoch 698/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6643 -
 accuracy: 0.6532
 Epoch 699/1000
 16557/16557 [=====] - 1s 75us/sample - loss: 0.6661 -
 accuracy: 0.6498
 Epoch 700/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6647 -
 accuracy: 0.6527
 Epoch 701/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6637 -
 accuracy: 0.6550
 Epoch 702/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6638 -
 accuracy: 0.6546
 Epoch 703/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6635 -
 accuracy: 0.6553
 Epoch 704/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6636 -
 accuracy: 0.6553
 Epoch 705/1000

16557/16557 [=====] - 1s 64us/sample - loss: 0.6632 -
 accuracy: 0.6559
 Epoch 706/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6630 -
 accuracy: 0.6560
 Epoch 707/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6629 -
 accuracy: 0.6564
 Epoch 708/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6631 -
 accuracy: 0.6559
 Epoch 709/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6631 -
 accuracy: 0.6558
 Epoch 710/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6633 -
 accuracy: 0.6549
 Epoch 711/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6641 -
 accuracy: 0.6538
 Epoch 712/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6637 -
 accuracy: 0.6548
 Epoch 713/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6638 -
 accuracy: 0.6544
 Epoch 714/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6633 -
 accuracy: 0.6547
 Epoch 715/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6635 -
 accuracy: 0.6550
 Epoch 716/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6641 -
 accuracy: 0.6540
 Epoch 717/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6634 -
 accuracy: 0.6550
 Epoch 718/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6634 -
 accuracy: 0.6555
 Epoch 719/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6632 -
 accuracy: 0.6560
 Epoch 720/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6627 -
 accuracy: 0.6565
 Epoch 721/1000

16557/16557 [=====] - 1s 64us/sample - loss: 0.6640 -
 accuracy: 0.6539
 Epoch 722/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6634 -
 accuracy: 0.6548
 Epoch 723/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6632 -
 accuracy: 0.6554
 Epoch 724/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6631 -
 accuracy: 0.6558
 Epoch 725/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6637 -
 accuracy: 0.6542
 Epoch 726/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6633 -
 accuracy: 0.6559
 Epoch 727/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6636 -
 accuracy: 0.6554
 Epoch 728/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6632 -
 accuracy: 0.6557
 Epoch 729/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6629 -
 accuracy: 0.6562
 Epoch 730/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6630 -
 accuracy: 0.6560
 Epoch 731/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6625 -
 accuracy: 0.6570
 Epoch 732/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6638 -
 accuracy: 0.6551
 Epoch 733/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6628 -
 accuracy: 0.6566
 Epoch 734/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6628 -
 accuracy: 0.6563
 Epoch 735/1000
 16557/16557 [=====] - 1s 72us/sample - loss: 0.6623 -
 accuracy: 0.6573
 Epoch 736/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6623 -
 accuracy: 0.6572
 Epoch 737/1000

16557/16557 [=====] - 1s 64us/sample - loss: 0.6630 -
 accuracy: 0.6556
 Epoch 738/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6624 -
 accuracy: 0.6571
 Epoch 739/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6631 -
 accuracy: 0.6553
 Epoch 740/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6639 -
 accuracy: 0.6541
 Epoch 741/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6636 -
 accuracy: 0.6550
 Epoch 742/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6634 -
 accuracy: 0.6545
 Epoch 743/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6628 -
 accuracy: 0.6564
 Epoch 744/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6627 -
 accuracy: 0.6563
 Epoch 745/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6632 -
 accuracy: 0.6553
 Epoch 746/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6630 -
 accuracy: 0.6559
 Epoch 747/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6640 -
 accuracy: 0.6530
 Epoch 748/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6634 -
 accuracy: 0.6550
 Epoch 749/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6634 -
 accuracy: 0.6542
 Epoch 750/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6626 -
 accuracy: 0.6564
 Epoch 751/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6636 -
 accuracy: 0.6540
 Epoch 752/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6634 -
 accuracy: 0.6552
 Epoch 753/1000

16557/16557 [=====] - 1s 65us/sample - loss: 0.6629 -
 accuracy: 0.6563
 Epoch 754/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6635 -
 accuracy: 0.6546
 Epoch 755/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6635 -
 accuracy: 0.6549
 Epoch 756/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6635 -
 accuracy: 0.6554
 Epoch 757/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6636 -
 accuracy: 0.6550
 Epoch 758/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6638 -
 accuracy: 0.6551
 Epoch 759/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6637 -
 accuracy: 0.6551
 Epoch 760/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6632 -
 accuracy: 0.6562
 Epoch 761/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6632 -
 accuracy: 0.6568
 Epoch 762/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6631 -
 accuracy: 0.6565
 Epoch 763/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6638 -
 accuracy: 0.6553
 Epoch 764/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6626 -
 accuracy: 0.6576
 Epoch 765/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6630 -
 accuracy: 0.6565
 Epoch 766/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6636 -
 accuracy: 0.6552
 Epoch 767/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6635 -
 accuracy: 0.6557
 Epoch 768/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6636 -
 accuracy: 0.6549
 Epoch 769/1000

16557/16557 [=====] - 1s 62us/sample - loss: 0.6629 -
 accuracy: 0.6569
 Epoch 770/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6629 -
 accuracy: 0.6569
 Epoch 771/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6624 -
 accuracy: 0.6576
 Epoch 772/1000
 16557/16557 [=====] - 1s 73us/sample - loss: 0.6633 -
 accuracy: 0.6554
 Epoch 773/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6622 -
 accuracy: 0.6581
 Epoch 774/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6623 -
 accuracy: 0.6580
 Epoch 775/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6626 -
 accuracy: 0.6574
 Epoch 776/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6628 -
 accuracy: 0.6568
 Epoch 777/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6630 -
 accuracy: 0.6559
 Epoch 778/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6636 -
 accuracy: 0.6553
 Epoch 779/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6624 -
 accuracy: 0.6575
 Epoch 780/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6627 -
 accuracy: 0.6568
 Epoch 781/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6625 -
 accuracy: 0.6572
 Epoch 782/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6629 -
 accuracy: 0.6565
 Epoch 783/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6626 -
 accuracy: 0.6575
 Epoch 784/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6627 -
 accuracy: 0.6571
 Epoch 785/1000

16557/16557 [=====] - 1s 67us/sample - loss: 0.6636 -
 accuracy: 0.6551
 Epoch 786/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6635 -
 accuracy: 0.6548
 Epoch 787/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6627 -
 accuracy: 0.6565
 Epoch 788/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6625 -
 accuracy: 0.6572
 Epoch 789/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6628 -
 accuracy: 0.6563
 Epoch 790/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6644 -
 accuracy: 0.6522
 Epoch 791/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6633 -
 accuracy: 0.6550
 Epoch 792/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6626 -
 accuracy: 0.6569
 Epoch 793/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6625 -
 accuracy: 0.6573
 Epoch 794/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6619 -
 accuracy: 0.6587
 Epoch 795/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6628 -
 accuracy: 0.6570
 Epoch 796/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6632 -
 accuracy: 0.6562
 Epoch 797/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6623 -
 accuracy: 0.6580
 Epoch 798/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6619 -
 accuracy: 0.6591
 Epoch 799/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6618 -
 accuracy: 0.6589
 Epoch 800/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6623 -
 accuracy: 0.6578
 Epoch 801/1000

16557/16557 [=====] - 1s 67us/sample - loss: 0.6621 -
 accuracy: 0.6586
 Epoch 802/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6635 -
 accuracy: 0.6551
 Epoch 803/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6621 -
 accuracy: 0.6579
 Epoch 804/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6620 -
 accuracy: 0.6584
 Epoch 805/1000
 16557/16557 [=====] - 1s 74us/sample - loss: 0.6623 -
 accuracy: 0.6572
 Epoch 806/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6625 -
 accuracy: 0.6572
 Epoch 807/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6624 -
 accuracy: 0.6575
 Epoch 808/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6623 -
 accuracy: 0.6575
 Epoch 809/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6622 -
 accuracy: 0.6577
 Epoch 810/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6619 -
 accuracy: 0.6585
 Epoch 811/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6621 -
 accuracy: 0.6581
 Epoch 812/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6625 -
 accuracy: 0.6573
 Epoch 813/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6626 -
 accuracy: 0.6572
 Epoch 814/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6629 -
 accuracy: 0.6566
 Epoch 815/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6635 -
 accuracy: 0.6553
 Epoch 816/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6629 -
 accuracy: 0.6568
 Epoch 817/1000

16557/16557 [=====] - 1s 66us/sample - loss: 0.6629 -
 accuracy: 0.6562
 Epoch 818/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6623 -
 accuracy: 0.6577
 Epoch 819/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6624 -
 accuracy: 0.6576
 Epoch 820/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6635 -
 accuracy: 0.6567
 Epoch 821/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6630 -
 accuracy: 0.6574
 Epoch 822/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6624 -
 accuracy: 0.6581
 Epoch 823/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6621 -
 accuracy: 0.6588
 Epoch 824/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6620 -
 accuracy: 0.6588
 Epoch 825/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6626 -
 accuracy: 0.6576
 Epoch 826/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6631 -
 accuracy: 0.6566
 Epoch 827/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6623 -
 accuracy: 0.6583
 Epoch 828/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6620 -
 accuracy: 0.6588
 Epoch 829/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6625 -
 accuracy: 0.6572
 Epoch 830/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6628 -
 accuracy: 0.6568
 Epoch 831/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6624 -
 accuracy: 0.6577
 Epoch 832/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6622 -
 accuracy: 0.6582
 Epoch 833/1000

16557/16557 [=====] - 1s 65us/sample - loss: 0.6622 -
 accuracy: 0.6583
 Epoch 834/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6619 -
 accuracy: 0.6588
 Epoch 835/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6620 -
 accuracy: 0.6587
 Epoch 836/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6618 -
 accuracy: 0.6588
 Epoch 837/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6598
 Epoch 838/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6623 -
 accuracy: 0.6581
 Epoch 839/1000
 16557/16557 [=====] - 1s 76us/sample - loss: 0.6632 -
 accuracy: 0.6565
 Epoch 840/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6628 -
 accuracy: 0.6568
 Epoch 841/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6621 -
 accuracy: 0.6587
 Epoch 842/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6626 -
 accuracy: 0.6581
 Epoch 843/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6625 -
 accuracy: 0.6582
 Epoch 844/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6640 -
 accuracy: 0.6553
 Epoch 845/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6619 -
 accuracy: 0.6592
 Epoch 846/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6622 -
 accuracy: 0.6586
 Epoch 847/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6623 -
 accuracy: 0.6578
 Epoch 848/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6619 -
 accuracy: 0.6586
 Epoch 849/1000

16557/16557 [=====] - 1s 65us/sample - loss: 0.6619 -
 accuracy: 0.6586
 Epoch 850/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6617 -
 accuracy: 0.6589
 Epoch 851/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6619 -
 accuracy: 0.6586
 Epoch 852/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6619 -
 accuracy: 0.6588
 Epoch 853/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6617 -
 accuracy: 0.6594
 Epoch 854/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6629 -
 accuracy: 0.6577
 Epoch 855/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6624 -
 accuracy: 0.6581
 Epoch 856/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6617 -
 accuracy: 0.6596
 Epoch 857/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6621 -
 accuracy: 0.6588
 Epoch 858/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6624 -
 accuracy: 0.6579
 Epoch 859/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6599
 Epoch 860/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6622 -
 accuracy: 0.6585
 Epoch 861/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6619 -
 accuracy: 0.6598
 Epoch 862/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6619 -
 accuracy: 0.6593
 Epoch 863/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6618 -
 accuracy: 0.6585
 Epoch 864/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6620 -
 accuracy: 0.6589
 Epoch 865/1000

16557/16557 [=====] - 1s 67us/sample - loss: 0.6614 -
 accuracy: 0.6601
 Epoch 866/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6628 -
 accuracy: 0.6569
 Epoch 867/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6618 -
 accuracy: 0.6587
 Epoch 868/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6596
 Epoch 869/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6624 -
 accuracy: 0.6583
 Epoch 870/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6620 -
 accuracy: 0.6592
 Epoch 871/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6617 -
 accuracy: 0.6597
 Epoch 872/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6621 -
 accuracy: 0.6581
 Epoch 873/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6595
 Epoch 874/1000
 16557/16557 [=====] - 1s 75us/sample - loss: 0.6615 -
 accuracy: 0.6597
 Epoch 875/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6620 -
 accuracy: 0.6589
 Epoch 876/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6622 -
 accuracy: 0.6582
 Epoch 877/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6621 -
 accuracy: 0.6585
 Epoch 878/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6595
 Epoch 879/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6615 -
 accuracy: 0.6600
 Epoch 880/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6615 -
 accuracy: 0.6600
 Epoch 881/1000

16557/16557 [=====] - 1s 68us/sample - loss: 0.6623 -
 accuracy: 0.6578
 Epoch 882/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6615 -
 accuracy: 0.6592
 Epoch 883/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6628 -
 accuracy: 0.6574
 Epoch 884/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6619 -
 accuracy: 0.6585
 Epoch 885/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6620 -
 accuracy: 0.6586
 Epoch 886/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6619 -
 accuracy: 0.6585
 Epoch 887/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6617 -
 accuracy: 0.6592
 Epoch 888/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6620 -
 accuracy: 0.6582
 Epoch 889/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6617 -
 accuracy: 0.6589
 Epoch 890/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6615 -
 accuracy: 0.6598
 Epoch 891/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6613 -
 accuracy: 0.6599
 Epoch 892/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6624 -
 accuracy: 0.6578
 Epoch 893/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6614 -
 accuracy: 0.6599
 Epoch 894/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6621 -
 accuracy: 0.6588
 Epoch 895/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6621 -
 accuracy: 0.6589
 Epoch 896/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6621 -
 accuracy: 0.6586
 Epoch 897/1000

16557/16557 [=====] - 1s 64us/sample - loss: 0.6619 -
 accuracy: 0.6588
 Epoch 898/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6632 -
 accuracy: 0.6557
 Epoch 899/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6622 -
 accuracy: 0.6579
 Epoch 900/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6616 -
 accuracy: 0.6594
 Epoch 901/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6616 -
 accuracy: 0.6592
 Epoch 902/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6613 -
 accuracy: 0.6602
 Epoch 903/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6617 -
 accuracy: 0.6592
 Epoch 904/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6615 -
 accuracy: 0.6597
 Epoch 905/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6614 -
 accuracy: 0.6598
 Epoch 906/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6623 -
 accuracy: 0.6583
 Epoch 907/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6619 -
 accuracy: 0.6592
 Epoch 908/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6627 -
 accuracy: 0.6572
 Epoch 909/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6623 -
 accuracy: 0.6579
 Epoch 910/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6622 -
 accuracy: 0.6580
 Epoch 911/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6622 -
 accuracy: 0.6585
 Epoch 912/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6619 -
 accuracy: 0.6595
 Epoch 913/1000

16557/16557 [=====] - 1s 62us/sample - loss: 0.6619 -
 accuracy: 0.6591
 Epoch 914/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6615 -
 accuracy: 0.6598
 Epoch 915/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6618 -
 accuracy: 0.6597
 Epoch 916/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6618 -
 accuracy: 0.6593
 Epoch 917/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6617 -
 accuracy: 0.6600
 Epoch 918/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6613 -
 accuracy: 0.6603
 Epoch 919/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6610 -
 accuracy: 0.6612
 Epoch 920/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6610 -
 accuracy: 0.6609
 Epoch 921/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6608 -
 accuracy: 0.6614
 Epoch 922/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6608 -
 accuracy: 0.6612
 Epoch 923/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6612 -
 accuracy: 0.6603
 Epoch 924/1000
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6621 -
 accuracy: 0.6586
 Epoch 925/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6620 -
 accuracy: 0.6588
 Epoch 926/1000
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6618 -
 accuracy: 0.6588
 Epoch 927/1000
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6616 -
 accuracy: 0.6594
 Epoch 928/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6619 -
 accuracy: 0.6588
 Epoch 929/1000

16557/16557 [=====] - 1s 66us/sample - loss: 0.6635 -
 accuracy: 0.6542
 Epoch 930/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6623 -
 accuracy: 0.6575
 Epoch 931/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6629 -
 accuracy: 0.6571
 Epoch 932/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6620 -
 accuracy: 0.6587
 Epoch 933/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6620 -
 accuracy: 0.6587
 Epoch 934/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6599
 Epoch 935/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6614 -
 accuracy: 0.6597
 Epoch 936/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6621 -
 accuracy: 0.6577
 Epoch 937/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6625 -
 accuracy: 0.6571
 Epoch 938/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6631 -
 accuracy: 0.6567
 Epoch 939/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6639 -
 accuracy: 0.6549
 Epoch 940/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6626 -
 accuracy: 0.6579
 Epoch 941/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6616 -
 accuracy: 0.6597
 Epoch 942/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6614 -
 accuracy: 0.6603
 Epoch 943/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6627 -
 accuracy: 0.6570
 Epoch 944/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6622 -
 accuracy: 0.6574
 Epoch 945/1000

16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6598
 Epoch 946/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6600
 Epoch 947/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6616 -
 accuracy: 0.6595
 Epoch 948/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6619 -
 accuracy: 0.6589
 Epoch 949/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6613 -
 accuracy: 0.6602
 Epoch 950/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6615 -
 accuracy: 0.6598
 Epoch 951/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6622 -
 accuracy: 0.6584
 Epoch 952/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6614 -
 accuracy: 0.6600
 Epoch 953/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6610 -
 accuracy: 0.6614
 Epoch 954/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6612 -
 accuracy: 0.6607
 Epoch 955/1000
 16557/16557 [=====] - 1s 66us/sample - loss: 0.6616 -
 accuracy: 0.6598
 Epoch 956/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6620 -
 accuracy: 0.6594
 Epoch 957/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6631 -
 accuracy: 0.6572
 Epoch 958/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6617 -
 accuracy: 0.6596
 Epoch 959/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6612 -
 accuracy: 0.6607
 Epoch 960/1000
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6618 -
 accuracy: 0.6590
 Epoch 961/1000

```

16557/16557 [=====] - 1s 65us/sample - loss: 0.6613 -
accuracy: 0.6604
Epoch 962/1000
16557/16557 [=====] - 1s 65us/sample - loss: 0.6611 -
accuracy: 0.6609
Epoch 963/1000
16557/16557 [=====] - 1s 70us/sample - loss: 0.6613 -
accuracy: 0.6606
Epoch 964/1000
16557/16557 [=====] - 1s 73us/sample - loss: 0.6618 -
accuracy: 0.6603
Epoch 965/1000
16557/16557 [=====] - 1s 71us/sample - loss: 0.6618 -
accuracy: 0.6600
Epoch 966/1000
16557/16557 [=====] - 1s 71us/sample - loss: 0.6612 -
accuracy: 0.6613
Epoch 967/1000
16557/16557 [=====] - 1s 69us/sample - loss: 0.6612 -
accuracy: 0.6611
Epoch 968/1000
16557/16557 [=====] - 1s 69us/sample - loss: 0.6620 -
accuracy: 0.6591
Epoch 969/1000
16557/16557 [=====] - 1s 68us/sample - loss: 0.6622 -
accuracy: 0.6591
Epoch 970/1000
16557/16557 [=====] - 1s 68us/sample - loss: 0.6621 -
accuracy: 0.6592
Epoch 971/1000
16557/16557 [=====] - 1s 68us/sample - loss: 0.6615 -
accuracy: 0.6604
Epoch 972/1000
16557/16557 [=====] - 1s 68us/sample - loss: 0.6615 -
accuracy: 0.6606
Epoch 973/1000
16557/16557 [=====] - 1s 67us/sample - loss: 0.6615 -
accuracy: 0.6606
Epoch 974/1000
16557/16557 [=====] - 1s 69us/sample - loss: 0.6621 -
accuracy: 0.6587
Epoch 975/1000
16557/16557 [=====] - 1s 68us/sample - loss: 0.6620 -
accuracy: 0.6585
Epoch 976/1000
16557/16557 [=====] - 1s 67us/sample - loss: 0.6633 -
accuracy: 0.6572
Epoch 977/1000

```

16557/16557 [=====] - 1s 68us/sample - loss: 0.6621 -
 accuracy: 0.6588
 Epoch 978/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6620 -
 accuracy: 0.6592
 Epoch 979/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6616 -
 accuracy: 0.6602
 Epoch 980/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6614 -
 accuracy: 0.6610
 Epoch 981/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6613 -
 accuracy: 0.6613
 Epoch 982/1000
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6610 -
 accuracy: 0.6615
 Epoch 983/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6609 -
 accuracy: 0.6618
 Epoch 984/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6614 -
 accuracy: 0.6606
 Epoch 985/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6628 -
 accuracy: 0.6581
 Epoch 986/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6623 -
 accuracy: 0.6587
 Epoch 987/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6620 -
 accuracy: 0.6590
 Epoch 988/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6635 -
 accuracy: 0.6539
 Epoch 989/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6624 -
 accuracy: 0.6573
 Epoch 990/1000
 16557/16557 [=====] - 1s 68us/sample - loss: 0.6618 -
 accuracy: 0.6588
 Epoch 991/1000
 16557/16557 [=====] - 1s 70us/sample - loss: 0.6617 -
 accuracy: 0.6599
 Epoch 992/1000
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6615 -
 accuracy: 0.6604
 Epoch 993/1000

```
16557/16557 [=====] - 1s 67us/sample - loss: 0.6614 -  
accuracy: 0.6602  
Epoch 994/1000  
16557/16557 [=====] - 1s 67us/sample - loss: 0.6615 -  
accuracy: 0.6598  
Epoch 995/1000  
16557/16557 [=====] - 1s 67us/sample - loss: 0.6609 -  
accuracy: 0.6615  
Epoch 996/1000  
16557/16557 [=====] - 1s 65us/sample - loss: 0.6607 -  
accuracy: 0.6619  
Epoch 997/1000  
16557/16557 [=====] - 1s 65us/sample - loss: 0.6610 -  
accuracy: 0.6611  
Epoch 998/1000  
16557/16557 [=====] - 1s 65us/sample - loss: 0.6614 -  
accuracy: 0.6603  
Epoch 999/1000  
16557/16557 [=====] - 1s 65us/sample - loss: 0.6619 -  
accuracy: 0.6588  
Epoch 1000/1000  
16557/16557 [=====] - 1s 65us/sample - loss: 0.6623 -  
accuracy: 0.6586  
5520/1 [=====]  
=====
```


Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

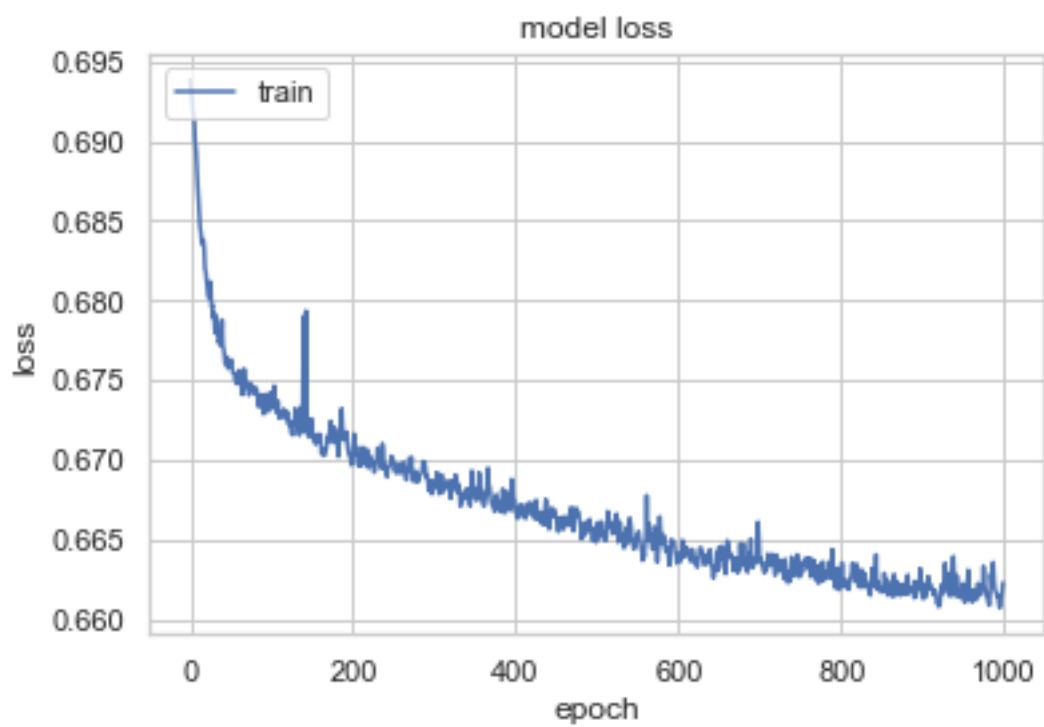
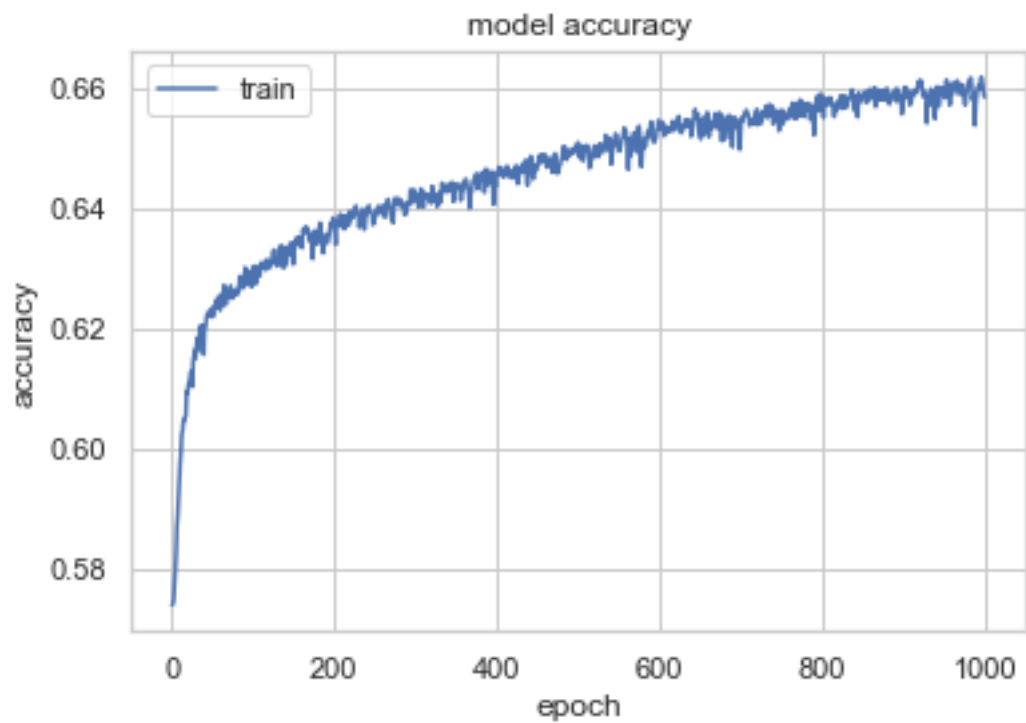
Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

```
=====] - 0s 71us/sample - loss: 0.7369 - accuracy: 0.5830
```

```
print("Accuracy: {}".format(test_acc))
plot_history(history)
```

Accuracy: 0.5829710364341736



4.1 Try with dropout layers if some values just don't matter

- Essentially, dropout layers just make some values randomly disappear with a given probability, so if some values don't matter when trying to make predictions, then those values in our nodes might be sent to 0, simplifying our overall model.

```
[65]: model = keras.Sequential([
        keras.layers.Dense(16, activation=tf.nn.relu),
        keras.layers.Dense(16, activation=tf.nn.relu),
        keras.layers.Dropout(.1),
        keras.layers.Dense(1, activation=tf.nn.sigmoid),
    ])
```

```
[66]: model.compile(optimizer='adam',
                    loss=tf.keras.losses.BinaryCrossentropy(from_logits=True),
                    metrics=['accuracy'])

history = model.fit(X_train, y_train, epochs=100)
test_loss, test_acc = model.evaluate(X_test, y_test)
```

Train on 16557 samples

Epoch 1/100

16557/16557 [=====] - 2s 94us/sample - loss: 0.7160 - accuracy: 0.5576

Epoch 2/100

16557/16557 [=====] - 1s 52us/sample - loss: 0.6947 - accuracy: 0.5731

Epoch 3/100

16557/16557 [=====] - 1s 72us/sample - loss: 0.6945 - accuracy: 0.5744

Epoch 4/100

16557/16557 [=====] - 1s 64us/sample - loss: 0.6944 - accuracy: 0.5744

Epoch 5/100

16557/16557 [=====] - 1s 57us/sample - loss: 0.6928 - accuracy: 0.5775

Epoch 6/100

16557/16557 [=====] - 1s 58us/sample - loss: 0.6922 - accuracy: 0.5791

Epoch 7/100

16557/16557 [=====] - 1s 60us/sample - loss: 0.6913 - accuracy: 0.5810

Epoch 8/100

16557/16557 [=====] - 1s 57us/sample - loss: 0.6908 - accuracy: 0.5812

Epoch 9/100

16557/16557 [=====] - 1s 58us/sample - loss: 0.6904 - accuracy: 0.5831

Epoch 10/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6902 -
accuracy: 0.5862

Epoch 11/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6896 -
accuracy: 0.5874

Epoch 12/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6895 -
accuracy: 0.5906

Epoch 13/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6891 -
accuracy: 0.5914

Epoch 14/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6880 -
accuracy: 0.5948

Epoch 15/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6877 -
accuracy: 0.5978

Epoch 16/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6933 -
accuracy: 0.5933

Epoch 17/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6966 -
accuracy: 0.5912

Epoch 18/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6967 -
accuracy: 0.5946

Epoch 19/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6942 -
accuracy: 0.5964

Epoch 20/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6846 -
accuracy: 0.6050

Epoch 21/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6839 -
accuracy: 0.6072

Epoch 22/100
16557/16557 [=====] - 1s 60us/sample - loss: 0.6839 -
accuracy: 0.6058

Epoch 23/100
16557/16557 [=====] - 1s 61us/sample - loss: 0.6817 -
accuracy: 0.6115

Epoch 24/100
16557/16557 [=====] - 1s 63us/sample - loss: 0.6819 -
accuracy: 0.6105

Epoch 25/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6817 -
accuracy: 0.6122

Epoch 26/100
16557/16557 [=====] - 1s 66us/sample - loss: 0.6811 -
accuracy: 0.6142

Epoch 27/100
16557/16557 [=====] - 1s 60us/sample - loss: 0.6817 -
accuracy: 0.6119

Epoch 28/100
16557/16557 [=====] - 1s 64us/sample - loss: 0.6806 -
accuracy: 0.6153

Epoch 29/100
16557/16557 [=====] - 1s 61us/sample - loss: 0.6814 -
accuracy: 0.6148

Epoch 30/100
16557/16557 [=====] - 1s 67us/sample - loss: 0.6807 -
accuracy: 0.6165

Epoch 31/100
16557/16557 [=====] - 1s 65us/sample - loss: 0.6796 -
accuracy: 0.6174

Epoch 32/100
16557/16557 [=====] - 1s 61us/sample - loss: 0.6781 -
accuracy: 0.6206

Epoch 33/100
16557/16557 [=====] - 1s 62us/sample - loss: 0.6792 -
accuracy: 0.6179

Epoch 34/100
16557/16557 [=====] - 1s 60us/sample - loss: 0.6786 -
accuracy: 0.6199

Epoch 35/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6775 -
accuracy: 0.6227

Epoch 36/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6763 -
accuracy: 0.6257

Epoch 37/100
16557/16557 [=====] - 1s 61us/sample - loss: 0.6773 -
accuracy: 0.6232

Epoch 38/100
16557/16557 [=====] - 1s 60us/sample - loss: 0.6766 -
accuracy: 0.6252

Epoch 39/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6766 -
accuracy: 0.6261

Epoch 40/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6767 -
accuracy: 0.6253

Epoch 41/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6761 -
accuracy: 0.6263

Epoch 42/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6750 -
accuracy: 0.6287

Epoch 43/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6756 -
accuracy: 0.6278

Epoch 44/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6754 -
accuracy: 0.6276

Epoch 45/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6749 -
accuracy: 0.6297

Epoch 46/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6759 -
accuracy: 0.6297

Epoch 47/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6746 -
accuracy: 0.6305

Epoch 48/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6742 -
accuracy: 0.6315

Epoch 49/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6745 -
accuracy: 0.6295

Epoch 50/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6749 -
accuracy: 0.6280

Epoch 51/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6747 -
accuracy: 0.6299

Epoch 52/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6742 -
accuracy: 0.6312

Epoch 53/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6745 -
accuracy: 0.6316

Epoch 54/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6736 -
accuracy: 0.6319

Epoch 55/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6743 -
accuracy: 0.6310

Epoch 56/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6736 -
accuracy: 0.6328

Epoch 57/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6730 -
accuracy: 0.6337

Epoch 58/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6740 -
accuracy: 0.6322
Epoch 59/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6728 -
accuracy: 0.6352
Epoch 60/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6737 -
accuracy: 0.6339
Epoch 61/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6743 -
accuracy: 0.6331
Epoch 62/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6738 -
accuracy: 0.6332
Epoch 63/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6731 -
accuracy: 0.6342
Epoch 64/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6718 -
accuracy: 0.6368
Epoch 65/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6731 -
accuracy: 0.6339
Epoch 66/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6726 -
accuracy: 0.6364s - loss: 0.6727 - ac
Epoch 67/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6740 -
accuracy: 0.6342
Epoch 68/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6734 -
accuracy: 0.6338
Epoch 69/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6739 -
accuracy: 0.6313
Epoch 70/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6725 -
accuracy: 0.6339
Epoch 71/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6716 -
accuracy: 0.6367
Epoch 72/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6714 -
accuracy: 0.6376
Epoch 73/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6716 -
accuracy: 0.6386

Epoch 74/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6710 -
accuracy: 0.6383

Epoch 75/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6706 -
accuracy: 0.6391

Epoch 76/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6707 -
accuracy: 0.6395

Epoch 77/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6711 -
accuracy: 0.6379

Epoch 78/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6716 -
accuracy: 0.6362

Epoch 79/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6714 -
accuracy: 0.6373

Epoch 80/100
16557/16557 [=====] - 1s 59us/sample - loss: 0.6710 -
accuracy: 0.6394

Epoch 81/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6716 -
accuracy: 0.6386

Epoch 82/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6713 -
accuracy: 0.6378

Epoch 83/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6706 -
accuracy: 0.6390

Epoch 84/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6720 -
accuracy: 0.6367

Epoch 85/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6710 -
accuracy: 0.6386

Epoch 86/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6715 -
accuracy: 0.6382

Epoch 87/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6714 -
accuracy: 0.6389

Epoch 88/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6705 -
accuracy: 0.6400

Epoch 89/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6712 -
accuracy: 0.6377

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

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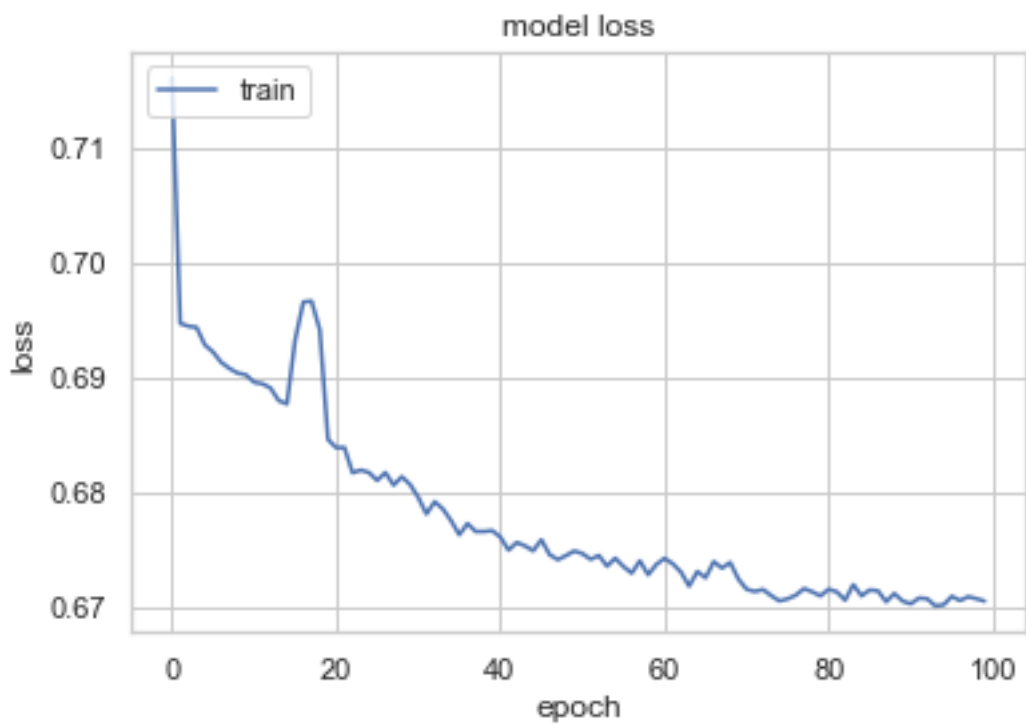
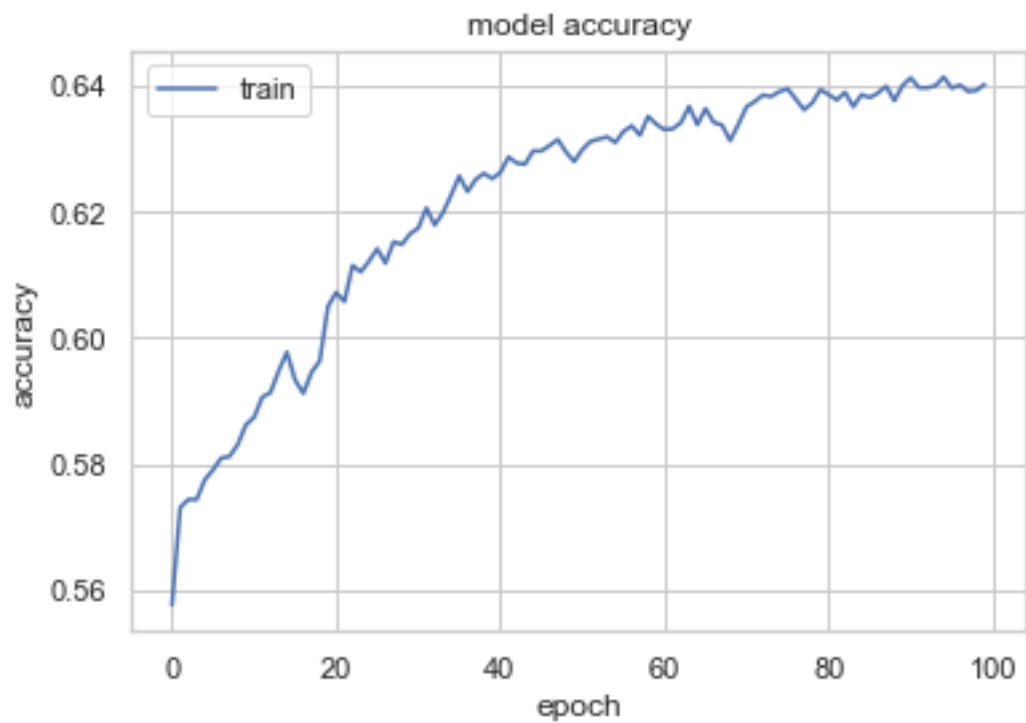
Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

```
=====] - 0s 55us/sample - loss: 0.7168 - accuracy: 0.5821
```

```
print("Accuracy: {}".format(test_acc))
plot_history(history)
```

Accuracy: 0.582065224647522



4.2 Try using tanh as activation to send to extreme values

- In the output layer, try using tanh to send values to -1 or 1, so we can maybe try to make our output more extreme (honestly this probably would have worked better with making it a strict binary classification problem)

```
[68]: model = keras.Sequential([
    keras.layers.Dense(16, activation=tf.nn.relu),
    keras.layers.Dense(16, activation=tf.nn.relu),
    keras.layers.Dropout(.1),
    keras.layers.Dense(1, activation=tf.nn.tanh),
])
```

```
[69]: model.compile(optimizer='adam',
                    loss=tf.keras.losses.BinaryCrossentropy(from_logits=True),
                    metrics=['accuracy'])

history = model.fit(X_train, y_train, epochs=100)
test_loss, test_acc = model.evaluate(X_test, y_test)
```

Train on 16557 samples

Epoch 1/100

16557/16557 [=====] - 2s 97us/sample - loss: 0.7421 - accuracy: 0.5636

Epoch 2/100

16557/16557 [=====] - 1s 53us/sample - loss: 0.7255 - accuracy: 0.5759

Epoch 3/100

16557/16557 [=====] - 1s 52us/sample - loss: 0.7184 - accuracy: 0.5800

Epoch 4/100

16557/16557 [=====] - 1s 53us/sample - loss: 0.7133 - accuracy: 0.5850

Epoch 5/100

16557/16557 [=====] - 1s 54us/sample - loss: 0.7075 - accuracy: 0.5885

Epoch 6/100

16557/16557 [=====] - 1s 53us/sample - loss: 0.7065 - accuracy: 0.5901

Epoch 7/100

16557/16557 [=====] - 1s 54us/sample - loss: 0.7022 - accuracy: 0.5914

Epoch 8/100

16557/16557 [=====] - 1s 54us/sample - loss: 0.7021 - accuracy: 0.5927

Epoch 9/100

16557/16557 [=====] - 1s 54us/sample - loss: 0.6985 - accuracy: 0.5944

Epoch 10/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6953 -
accuracy: 0.5961

Epoch 11/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6949 -
accuracy: 0.5995

Epoch 12/100
16557/16557 [=====] - 1s 53us/sample - loss: 0.6925 -
accuracy: 0.6013

Epoch 13/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6890 -
accuracy: 0.6038

Epoch 14/100
16557/16557 [=====] - 1s 53us/sample - loss: 0.6903 -
accuracy: 0.6025

Epoch 15/100
16557/16557 [=====] - 1s 53us/sample - loss: 0.6867 -
accuracy: 0.6058

Epoch 16/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6850 -
accuracy: 0.6052

Epoch 17/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6870 -
accuracy: 0.6089

Epoch 18/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6833 -
accuracy: 0.6097

Epoch 19/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6824 -
accuracy: 0.6053

Epoch 20/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6805 -
accuracy: 0.6101

Epoch 21/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6779 -
accuracy: 0.6129

Epoch 22/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6762 -
accuracy: 0.6108

Epoch 23/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6754 -
accuracy: 0.6115

Epoch 24/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6734 -
accuracy: 0.6136

Epoch 25/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6721 -
accuracy: 0.6165

Epoch 26/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6692 -
accuracy: 0.6187

Epoch 27/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6681 -
accuracy: 0.6226

Epoch 28/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6687 -
accuracy: 0.6212

Epoch 29/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6713 -
accuracy: 0.6221

Epoch 30/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6692 -
accuracy: 0.6209

Epoch 31/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6690 -
accuracy: 0.6223

Epoch 32/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6638 -
accuracy: 0.6252

Epoch 33/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6644 -
accuracy: 0.6254

Epoch 34/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6622 -
accuracy: 0.6285

Epoch 35/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6582 -
accuracy: 0.6316

Epoch 36/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6596 -
accuracy: 0.6316

Epoch 37/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6597 -
accuracy: 0.6319

Epoch 38/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6593 -
accuracy: 0.6326

Epoch 39/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6599 -
accuracy: 0.6320

Epoch 40/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6636 -
accuracy: 0.6312

Epoch 41/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6628 -
accuracy: 0.6347

Epoch 42/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6627 -
accuracy: 0.6318
Epoch 43/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6586 -
accuracy: 0.6359
Epoch 44/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6605 -
accuracy: 0.6325
Epoch 45/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6582 -
accuracy: 0.6376
Epoch 46/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6565 -
accuracy: 0.6382
Epoch 47/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6569 -
accuracy: 0.6417
Epoch 48/100
16557/16557 [=====] - 1s 54us/sample - loss: 0.6566 -
accuracy: 0.6418
Epoch 49/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6594 -
accuracy: 0.6386
Epoch 50/100
16557/16557 [=====] - 1s 55us/sample - loss: 0.6562 -
accuracy: 0.6418
Epoch 51/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6563 -
accuracy: 0.6415
Epoch 52/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6578 -
accuracy: 0.6399
Epoch 53/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6557 -
accuracy: 0.6417
Epoch 54/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6560 -
accuracy: 0.6433
Epoch 55/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6562 -
accuracy: 0.6408
Epoch 56/100
16557/16557 [=====] - 1s 61us/sample - loss: 0.6510 -
accuracy: 0.6469
Epoch 57/100
16557/16557 [=====] - 1s 69us/sample - loss: 0.6580 -
accuracy: 0.6403

Epoch 58/100
16557/16557 [=====] - 1s 61us/sample - loss: 0.6550 -
accuracy: 0.6416

Epoch 59/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6549 -
accuracy: 0.6424

Epoch 60/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6551 -
accuracy: 0.6429

Epoch 61/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6572 -
accuracy: 0.6421

Epoch 62/100
16557/16557 [=====] - 1s 61us/sample - loss: 0.6532 -
accuracy: 0.6446

Epoch 63/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6521 -
accuracy: 0.6460

Epoch 64/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6514 -
accuracy: 0.6440

Epoch 65/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6501 -
accuracy: 0.6451

Epoch 66/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6496 -
accuracy: 0.6462

Epoch 67/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6463 -
accuracy: 0.6501

Epoch 68/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6461 -
accuracy: 0.6491

Epoch 69/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6487 -
accuracy: 0.6476

Epoch 70/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6470 -
accuracy: 0.6485

Epoch 71/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6455 -
accuracy: 0.6496

Epoch 72/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6493 -
accuracy: 0.6469

Epoch 73/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6448 -
accuracy: 0.6510

Epoch 74/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6468 -
accuracy: 0.6486

Epoch 75/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6446 -
accuracy: 0.6511

Epoch 76/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6456 -
accuracy: 0.6491

Epoch 77/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6439 -
accuracy: 0.6521

Epoch 78/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6456 -
accuracy: 0.6506

Epoch 79/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6488 -
accuracy: 0.6475

Epoch 80/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6455 -
accuracy: 0.6499

Epoch 81/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6446 -
accuracy: 0.6520

Epoch 82/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6440 -
accuracy: 0.6530

Epoch 83/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6449 -
accuracy: 0.6527

Epoch 84/100
16557/16557 [=====] - 1s 57us/sample - loss: 0.6436 -
accuracy: 0.6534

Epoch 85/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6439 -
accuracy: 0.6537

Epoch 86/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6446 -
accuracy: 0.6528

Epoch 87/100
16557/16557 [=====] - 1s 58us/sample - loss: 0.6424 -
accuracy: 0.6552

Epoch 88/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6443 -
accuracy: 0.6537

Epoch 89/100
16557/16557 [=====] - 1s 56us/sample - loss: 0.6480 -
accuracy: 0.6509

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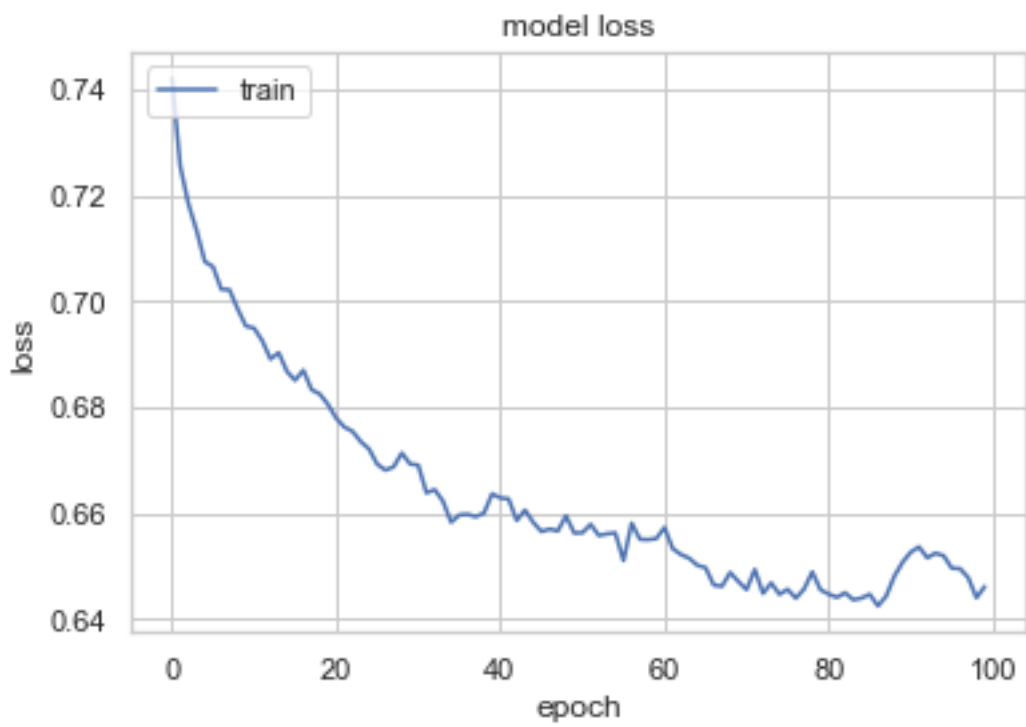
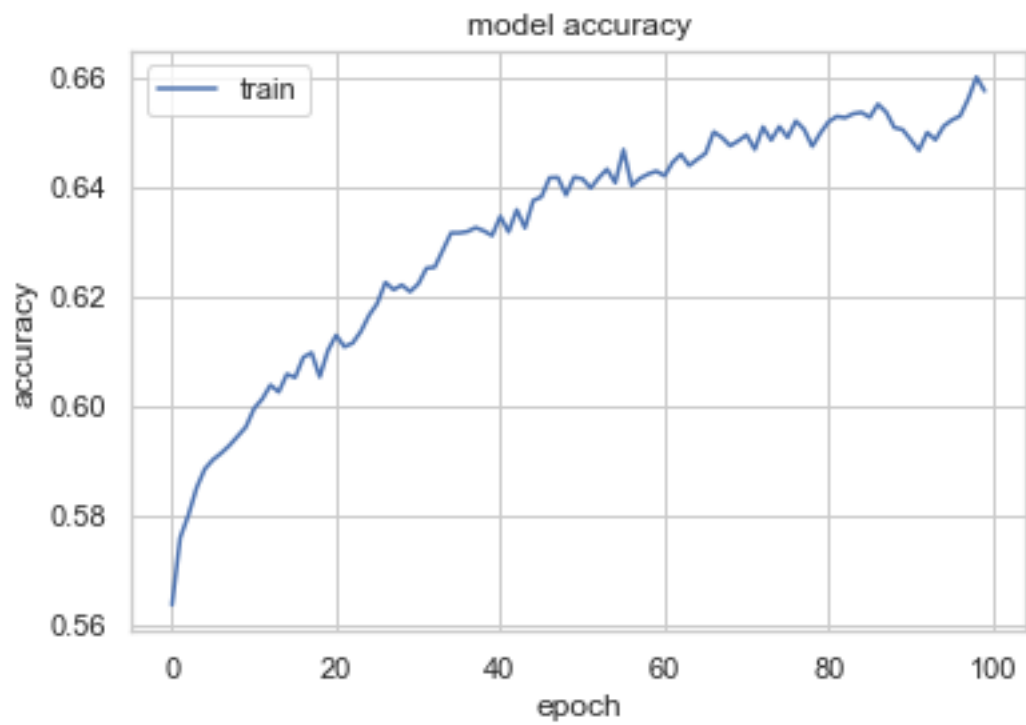
Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.


```
=====] - 0s 58us/sample - loss: 0.7524 - accuracy: 0.5734
```

```
print("Accuracy: {}".format(test_acc))
plot_history(history)
```

Accuracy: 0.573369562625885



4.3 Try adding more layers

- We can maybe add more layers to try to capture additional complexity

```
[75]: model = keras.Sequential([
        keras.layers.Dense(16, activation=tf.nn.relu),
        keras.layers.Dense(16, activation=tf.nn.relu),
        keras.layers.Dense(16, activation=tf.nn.relu),
        keras.layers.Dropout(.1),
        keras.layers.Dense(1, activation=tf.nn.sigmoid),
    ])
```

```
[76]: model.compile(optimizer='adam',
                    loss=tf.keras.losses.BinaryCrossentropy(from_logits=True),
                    metrics=['accuracy'])

history = model.fit(X_train, y_train, epochs=100)
test_loss, test_acc = model.evaluate(X_test, y_test)
```

Train on 16557 samples

Epoch 1/100

16557/16557 [=====] - 2s 122us/sample - loss: 0.6940 - accuracy: 0.5738

Epoch 2/100

16557/16557 [=====] - 1s 63us/sample - loss: 0.6931 - accuracy: 0.5739

Epoch 3/100

16557/16557 [=====] - 1s 62us/sample - loss: 0.6927 - accuracy: 0.5739

Epoch 4/100

16557/16557 [=====] - 1s 62us/sample - loss: 0.6920 - accuracy: 0.5764

Epoch 5/100

16557/16557 [=====] - 1s 63us/sample - loss: 0.6912 - accuracy: 0.5806

Epoch 6/100

16557/16557 [=====] - 1s 62us/sample - loss: 0.6905 - accuracy: 0.5837

Epoch 7/100

16557/16557 [=====] - 1s 62us/sample - loss: 0.6896 - accuracy: 0.5875

Epoch 8/100

16557/16557 [=====] - 1s 61us/sample - loss: 0.6897 - accuracy: 0.5890

Epoch 9/100

16557/16557 [=====] - 1s 61us/sample - loss: 0.6887 - accuracy: 0.5909

Epoch 10/100

16557/16557 [=====] - 1s 62us/sample - loss: 0.6878 -
 accuracy: 0.5924
 Epoch 11/100
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6879 -
 accuracy: 0.5933
 Epoch 12/100
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6880 -
 accuracy: 0.5927
 Epoch 13/100
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6879 -
 accuracy: 0.5920
 Epoch 14/100
 16557/16557 [=====] - 1s 60us/sample - loss: 0.6863 -
 accuracy: 0.5956
 Epoch 15/100
 16557/16557 [=====] - 1s 71us/sample - loss: 0.6859 -
 accuracy: 0.5980
 Epoch 16/100
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6854 -
 accuracy: 0.5994
 Epoch 17/100
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6851 -
 accuracy: 0.6003
 Epoch 18/100
 16557/16557 [=====] - 1s 84us/sample - loss: 0.6859 -
 accuracy: 0.5995
 Epoch 19/100
 16557/16557 [=====] - 1s 74us/sample - loss: 0.6843 -
 accuracy: 0.6017
 Epoch 20/100
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6839 -
 accuracy: 0.6045
 Epoch 21/100
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6834 -
 accuracy: 0.6055
 Epoch 22/100
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6829 -
 accuracy: 0.6060
 Epoch 23/100
 16557/16557 [=====] - 1s 84us/sample - loss: 0.6824 -
 accuracy: 0.6071
 Epoch 24/100
 16557/16557 [=====] - 1s 79us/sample - loss: 0.6824 -
 accuracy: 0.6078
 Epoch 25/100
 16557/16557 [=====] - 1s 82us/sample - loss: 0.6816 -
 accuracy: 0.6091
 Epoch 26/100

16557/16557 [=====] - 1s 68us/sample - loss: 0.6829 -
 accuracy: 0.6078
 Epoch 27/100
 16557/16557 [=====] - 1s 75us/sample - loss: 0.6836 -
 accuracy: 0.6064
 Epoch 28/100
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6829 -
 accuracy: 0.6067
 Epoch 29/100
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6830 -
 accuracy: 0.6071
 Epoch 30/100
 16557/16557 [=====] - 1s 67us/sample - loss: 0.6820 -
 accuracy: 0.6083
 Epoch 31/100
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6819 -
 accuracy: 0.6088
 Epoch 32/100
 16557/16557 [=====] - 1s 63us/sample - loss: 0.6820 -
 accuracy: 0.6085
 Epoch 33/100
 16557/16557 [=====] - 1s 65us/sample - loss: 0.6815 -
 accuracy: 0.6100
 Epoch 34/100
 16557/16557 [=====] - 1s 74us/sample - loss: 0.6810 -
 accuracy: 0.6119
 Epoch 35/100
 16557/16557 [=====] - 1s 70us/sample - loss: 0.6821 -
 accuracy: 0.6069
 Epoch 36/100
 16557/16557 [=====] - 1s 64us/sample - loss: 0.6814 -
 accuracy: 0.6104
 Epoch 37/100
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6799 -
 accuracy: 0.6140
 Epoch 38/100
 16557/16557 [=====] - 1s 62us/sample - loss: 0.6809 -
 accuracy: 0.6138
 Epoch 39/100
 16557/16557 [=====] - 1s 70us/sample - loss: 0.6810 -
 accuracy: 0.6138
 Epoch 40/100
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6808 -
 accuracy: 0.6119
 Epoch 41/100
 16557/16557 [=====] - 1s 76us/sample - loss: 0.6801 -
 accuracy: 0.6130
 Epoch 42/100

16557/16557 [=====] - 2s 100us/sample - loss: 0.6806 -
 accuracy: 0.6139
 Epoch 43/100
 16557/16557 [=====] - 1s 79us/sample - loss: 0.6802 -
 accuracy: 0.6128
 Epoch 44/100
 16557/16557 [=====] - 1s 76us/sample - loss: 0.6792 -
 accuracy: 0.6162
 Epoch 45/100
 16557/16557 [=====] - 1s 76us/sample - loss: 0.6804 -
 accuracy: 0.6153
 Epoch 46/100
 16557/16557 [=====] - 1s 77us/sample - loss: 0.6807 -
 accuracy: 0.6146
 Epoch 47/100
 16557/16557 [=====] - 1s 79us/sample - loss: 0.6801 -
 accuracy: 0.6180
 Epoch 48/100
 16557/16557 [=====] - 1s 77us/sample - loss: 0.6799 -
 accuracy: 0.6183
 Epoch 49/100
 16557/16557 [=====] - 1s 80us/sample - loss: 0.6804 -
 accuracy: 0.6174
 Epoch 50/100
 16557/16557 [=====] - 1s 84us/sample - loss: 0.6803 -
 accuracy: 0.6163
 Epoch 51/100
 16557/16557 [=====] - 1s 84us/sample - loss: 0.6795 -
 accuracy: 0.6180
 Epoch 52/100
 16557/16557 [=====] - 1s 81us/sample - loss: 0.6790 -
 accuracy: 0.6184
 Epoch 53/100
 16557/16557 [=====] - 1s 87us/sample - loss: 0.6797 -
 accuracy: 0.6142
 Epoch 54/100
 16557/16557 [=====] - 2s 98us/sample - loss: 0.6785 -
 accuracy: 0.6169
 Epoch 55/100
 16557/16557 [=====] - 2s 103us/sample - loss: 0.6782 -
 accuracy: 0.6177
 Epoch 56/100
 16557/16557 [=====] - 2s 145us/sample - loss: 0.6778 -
 accuracy: 0.6190
 Epoch 57/100
 16557/16557 [=====] - 2s 101us/sample - loss: 0.6773 -
 accuracy: 0.6210
 Epoch 58/100

16557/16557 [=====] - 1s 90us/sample - loss: 0.6771 -
 accuracy: 0.6222
 Epoch 59/100
 16557/16557 [=====] - 2s 150us/sample - loss: 0.6777 -
 accuracy: 0.6211
 Epoch 60/100
 16557/16557 [=====] - 1s 85us/sample - loss: 0.6786 -
 accuracy: 0.6196
 Epoch 61/100
 16557/16557 [=====] - 2s 100us/sample - loss: 0.6781 -
 accuracy: 0.6197
 Epoch 62/100
 16557/16557 [=====] - 1s 83us/sample - loss: 0.6773 -
 accuracy: 0.6226
 Epoch 63/100
 16557/16557 [=====] - 1s 86us/sample - loss: 0.6790 -
 accuracy: 0.6188
 Epoch 64/100
 16557/16557 [=====] - 2s 94us/sample - loss: 0.6794 -
 accuracy: 0.6163
 Epoch 65/100
 16557/16557 [=====] - 1s 82us/sample - loss: 0.6778 -
 accuracy: 0.6204
 Epoch 66/100
 16557/16557 [=====] - 2s 91us/sample - loss: 0.6773 -
 accuracy: 0.6212
 Epoch 67/100
 16557/16557 [=====] - 2s 100us/sample - loss: 0.6778 -
 accuracy: 0.6208
 Epoch 68/100
 16557/16557 [=====] - 1s 89us/sample - loss: 0.6774 -
 accuracy: 0.6219
 Epoch 69/100
 16557/16557 [=====] - 1s 82us/sample - loss: 0.6770 -
 accuracy: 0.6228
 Epoch 70/100
 16557/16557 [=====] - 1s 78us/sample - loss: 0.6784 -
 accuracy: 0.6193
 Epoch 71/100
 16557/16557 [=====] - 1s 76us/sample - loss: 0.6782 -
 accuracy: 0.6195
 Epoch 72/100
 16557/16557 [=====] - 1s 79us/sample - loss: 0.6777 -
 accuracy: 0.6227
 Epoch 73/100
 16557/16557 [=====] - 1s 73us/sample - loss: 0.6802 -
 accuracy: 0.6180
 Epoch 74/100

16557/16557 [=====] - 1s 73us/sample - loss: 0.6793 -
 accuracy: 0.6199
 Epoch 75/100
 16557/16557 [=====] - 1s 74us/sample - loss: 0.6792 -
 accuracy: 0.6187
 Epoch 76/100
 16557/16557 [=====] - 1s 72us/sample - loss: 0.6787 -
 accuracy: 0.6182
 Epoch 77/100
 16557/16557 [=====] - 1s 73us/sample - loss: 0.6791 -
 accuracy: 0.6171
 Epoch 78/100
 16557/16557 [=====] - 1s 74us/sample - loss: 0.6779 -
 accuracy: 0.6202
 Epoch 79/100
 16557/16557 [=====] - 1s 72us/sample - loss: 0.6783 -
 accuracy: 0.6216
 Epoch 80/100
 16557/16557 [=====] - 1s 73us/sample - loss: 0.6777 -
 accuracy: 0.6207
 Epoch 81/100
 16557/16557 [=====] - 1s 71us/sample - loss: 0.6777 -
 accuracy: 0.6216
 Epoch 82/100
 16557/16557 [=====] - 1s 73us/sample - loss: 0.6782 -
 accuracy: 0.6212
 Epoch 83/100
 16557/16557 [=====] - 1s 72us/sample - loss: 0.6777 -
 accuracy: 0.6229
 Epoch 84/100
 16557/16557 [=====] - 1s 70us/sample - loss: 0.6772 -
 accuracy: 0.6243
 Epoch 85/100
 16557/16557 [=====] - 1s 71us/sample - loss: 0.6764 -
 accuracy: 0.6253
 Epoch 86/100
 16557/16557 [=====] - 1s 69us/sample - loss: 0.6763 -
 accuracy: 0.6249
 Epoch 87/100
 16557/16557 [=====] - 1s 70us/sample - loss: 0.6754 -
 accuracy: 0.6268
 Epoch 88/100
 16557/16557 [=====] - 1s 73us/sample - loss: 0.6755 -
 accuracy: 0.6266s - loss: 0.6754 - accuracy: 0.
 Epoch 89/100
 16557/16557 [=====] - 1s 70us/sample - loss: 0.6767 -
 accuracy: 0.6252
 Epoch 90/100

Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

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Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

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Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

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Handwriting practice lines consisting of multiple sets of three horizontal lines (top solid, middle dashed, bottom solid) for letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple sets of three horizontal lines (top, middle, and bottom lines) for letter formation.

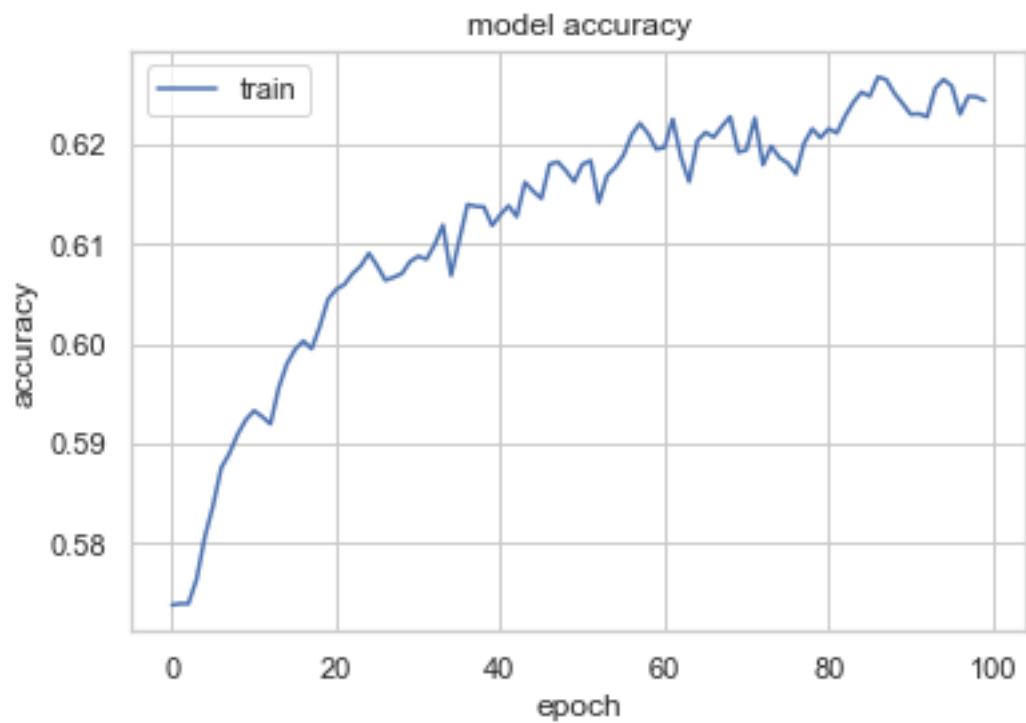
Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

Handwriting practice lines consisting of multiple rows of dashed lines on a solid background, designed for tracing and letter formation.

```
=====] - 1s 97us/sample - loss: 0.7343 - accuracy: 0.5844
```

```
print("Accuracy: {}".format(test_acc))
plot_history(history)
```

Accuracy: 0.5844202637672424



5 Conclusions and Future Directions

- The performance of SVC was around 50%, and the recall of if a stock didn't beat SPY was quite good.
- For deep learning, the first simple model achieved ~62% accuracy on the training set, and ~58% on the testing set.
 - Pumping up the amount of epochs for this model didn't really help – we see that the loss curve starts to flatten out at around 0.6750, which is very high loss. Thus, we probably need to make a much more complex model than the naive model.
- Making the layers much larger also didn't really help, it didn't significantly improve accuracy.
- Adding a dropout layer also didn't help the testing accuracy
- Adding further layers slightly helped the accuracy, but not by much

In sum, it's hard to predict the stock market – any potential alpha here is likely already exhausted, as this is old data. Also, there's some shortcuts with data cleaning we did and reducing outliers that reduce the generalization capacity.

5.1 Future Directions

- Turning the problem into just a binary classification might work slightly better, but not by much – it's already pretty much like this
- Exploring more complex networks in general could also be beneficial. Perhaps at very high epochs with a more clever model, we can derive more insights.

[]: