Leander_ACS61011_Project

March 19, 2024

0.1 Multiclass Automated Speech Recognition using a Baseline and an Advanced Model

0.1.1 Prerequisites

[43]: # get the data from github and unzip

• Download the speech image transformed data from GitHub and unzip it in the current directory:

!wget https://raw.githubusercontent.com/andrsn/data/main/speechImageData.zip

• Install all the necessary libraries for our notebook

```
[2]: !pip install scikeras pydub

Collecting scikeras
```

2024-03-19 08:58:45 (177 MB/s) - 'speechImageData.zip.1' saved [9872924/9872924]

```
Downloading scikeras-0.12.0-py3-none-any.whl (27 kB)

Collecting pydub

Downloading pydub-0.25.1-py2.py3-none-any.whl (32 kB)

Requirement already satisfied: packaging>=0.21 in

/usr/local/lib/python3.10/dist-packages (from scikeras) (24.0)

Requirement already satisfied: scikit-learn>=1.0.0 in

/usr/local/lib/python3.10/dist-packages (from scikeras) (1.2.2)
```

```
Requirement already satisfied: numpy>=1.17.3 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (1.25.2)
Requirement already satisfied: scipy>=1.3.2 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (1.11.4)
Requirement already satisfied: joblib>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (1.3.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (3.3.0)
Installing collected packages: pydub, scikeras
Successfully installed pydub-0.25.1 scikeras-0.12.0
```

0.1.2 1. Import Libraries and define constants

We will start by importing the necessary libraries and defining the constants that will be used throughout the notebook.

```
[44]: import random
      import shutil
      import librosa
      import soundfile as sf
      import numpy as np
      import seaborn as sns
      import tensorflow as tf
      import matplotlib.pyplot as plt
      from keras import optimizers, regularizers
      from keras.models import Sequential
      from keras.layers import Dense, Dropout, BatchNormalization, Input, Conv2D,
       →MaxPooling2D, Flatten
      from keras.applications import MobileNetV2
      from keras.applications.mobilenet_v2 import preprocess_input
      from pydub import AudioSegment
      from sklearn.metrics import confusion_matrix
      from sklearn.model_selection import RandomizedSearchCV
      from scikeras.wrappers import KerasClassifier
      from hyperopt import fmin, tpe, hp, STATUS_OK, Trials, space_eval
      NUM CLASSES = 12
      BATCH SIZE = 128
      IMG\_SIZE = (98, 50)
      TIMEPOOL_SIZE = 12
```

0.1.3 2. Data Preprocessing

• Create usable keras dataset components from the extracted files.

• In total, there are 12 classes of different spoken words and the spectrograms, which form the input image data are of size 98x50 pixels.

```
[45]: # Load the data
      train_ds = tf.keras.utils.image_dataset_from_directory(
          directory='/content/speechImageData/TrainData',
          labels='inferred',
          color_mode="grayscale",
          label_mode='categorical',
          batch_size=BATCH_SIZE,
          image_size=IMG_SIZE
      )
      val_ds = tf.keras.utils.image_dataset_from_directory(
          directory='/content/speechImageData/ValData',
          labels='inferred',
          color_mode="grayscale",
          label_mode='categorical',
          batch_size=BATCH_SIZE,
          image_size=IMG_SIZE
      # Extract the training input images and output class labels
      x_train = []
      y_train = []
      for images, labels in train_ds.take(-1):
          x_train.append(images.numpy())
          y_train.append(labels.numpy())
      x_train = np.concatenate(x_train, axis=0)
      y_train = np.concatenate(y_train, axis=0)
      print(y_train)
      # Extract the validation input images and output class labels
      x_val = []
      y_val = []
      for images, labels in val_ds.take(-1):
          x_val.append(images.numpy())
          y_val.append(labels.numpy())
      x_val = np.concatenate(x_val, axis=0)
      y_val = np.concatenate(y_val, axis=0)
      print(y_val)
```

Found 2023 files belonging to 12 classes.

```
Found 1181 files belonging to 12 classes.

[[0. 0. 1. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 1.]

[0. 0. 0. ... 0. 0. 0.]

...

[0. 0. 0. ... 0. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]

[0. 1. 0. ... 0. 0. 0.]

[0. 1. 0. ... 0. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]
```

0.1.4 3. Model Design

We now attempt to approach the problem with five tasks:

Task 1: Baseline Model

Model features: The following are its features:

- The baseline model is a simple Convolutional Neural Network (CNN) with one input layer, four hidden layers, one fully connected layer and an output layer.
- The input layer consists of the following:
 - A Conv2D layer with 32 filters, a kernel size of 3x3, and a ReLU activation function.
 - A BatchNormalization layer.
 - A MaxPooling2D layer with a pool size of 2x2.
- The four hidden layers are replicated from the input layer.
- A time pooling layer is added to the model to combat the start time of the audio.
- The fully connected layer consists of 1024 units and a ReLU activation function.
- The output layer consists of 12 units and a softmax activation function.

Additional model hyperparameters:

- The model uses the Adam optimizer with a learning rate of 0.001.
- The L2 regularization parameter is set to 0.001.
- While training, an early stopping callback is used to stop the training process if the validation accuracy does not decrease for 5 epochs.

```
fully_connected_num_filters = 1024
    # define model
    model = Sequential()
    # input layer
    model.add(Input(shape=(IMG_SIZE[0], IMG_SIZE[1], 1)))
    model.add(Conv2D(input_num_filters, kernel_size =(3, 3), padding='same',_
 →activation='relu'))
    model.add(BatchNormalization())
    # hidden layers
    for i in range(num_layers):
        if passthrough:
            model.add(Conv2D(num_filters[i], kernel_size =(3, 3),__
 →padding='same', activation='relu'))
        else:
            model.add(Conv2D(num_filters, kernel_size =(3, 3), padding='same',__
 ⇔activation='relu'))
        model.add(BatchNormalization())
        model.add(MaxPooling2D(pool_size = (2, 2), strides=(2, 2),
 →padding='same'))
    # Time based pooling
    model.add(MaxPooling2D(pool_size=(TIMEPOOL_SIZE, 1),__
 ⇔strides=(TIMEPOOL_SIZE, 1), padding='same'))
    # fully connected layer
    model.add(Flatten())
    model.add(Dense(fully_connected_num_filters,_

-kernel_regularizer=regularizers.12(0.01), activation='relu'))

    model.add(Dropout(0.2))
    # output layer
    model.add(Dense(NUM_CLASSES, activation='softmax'))
    # set adam optimizer
    opt = optimizers.Adam(learning_rate=0.001)
    model.compile(loss="categorical_crossentropy", optimizer=opt,
 →metrics=["accuracy"])
    return model
hidden_num_layers = [2, 3, 4, 5]
hidden_num_filters = [128, 128, 128, 128]
```

```
task1_model = t1_model(len(hidden_num_layers), hidden_num_filters,_u 
passthrough=True)
task1_model.summary()

early_stopping = tf.keras.callbacks.EarlyStopping(monitor='val_accuracy',_u 
patience=5)
```

Model: "sequential"

Layer (type)		Param #
	(None, 98, 50, 32)	
<pre>batch_normalization (Batch Normalization)</pre>	(None, 98, 50, 32)	128
conv2d_1 (Conv2D)	(None, 98, 50, 128)	36992
<pre>batch_normalization_1 (Bat chNormalization)</pre>	(None, 98, 50, 128)	512
<pre>max_pooling2d (MaxPooling2 D)</pre>	(None, 49, 25, 128)	0
conv2d_2 (Conv2D)	(None, 49, 25, 128)	147584
<pre>batch_normalization_2 (Bat chNormalization)</pre>	(None, 49, 25, 128)	512
<pre>max_pooling2d_1 (MaxPoolin g2D)</pre>	(None, 25, 13, 128)	0
conv2d_3 (Conv2D)	(None, 25, 13, 128)	147584
<pre>batch_normalization_3 (Bat chNormalization)</pre>	(None, 25, 13, 128)	512
<pre>max_pooling2d_2 (MaxPoolin g2D)</pre>	(None, 13, 7, 128)	0
conv2d_4 (Conv2D)	(None, 13, 7, 128)	147584
<pre>batch_normalization_4 (Bat chNormalization)</pre>	(None, 13, 7, 128)	512
<pre>max_pooling2d_3 (MaxPoolin g2D)</pre>	(None, 7, 4, 128)	0

Non-trainable params: 1088 (4.25 KB)

Trainable params: 1018764 (3.89 MB)

T1.A - Model Training This section trains the deep convolutional network using the Adam algorithm.

```
[6]: history = task1_model.fit(x_train, y_train, batch_size=BATCH_SIZE, epochs=15,_ovalidation_data=(x_val, y_val), callbacks=[early_stopping])
```

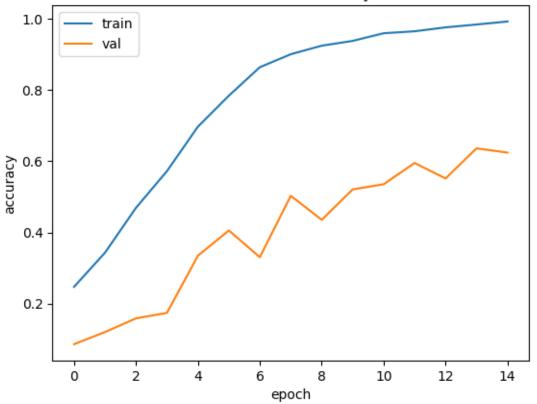
```
Epoch 1/15
accuracy: 0.2474 - val_loss: 14.2474 - val_accuracy: 0.0863
Epoch 2/15
accuracy: 0.3433 - val_loss: 11.3006 - val_accuracy: 0.1204
Epoch 3/15
accuracy: 0.4693 - val_loss: 8.7848 - val_accuracy: 0.1588
Epoch 4/15
accuracy: 0.5722 - val_loss: 7.1032 - val_accuracy: 0.1742
accuracy: 0.6972 - val_loss: 5.4696 - val_accuracy: 0.3348
accuracy: 0.7841 - val_loss: 4.8391 - val_accuracy: 0.4056
accuracy: 0.8641 - val_loss: 4.8459 - val_accuracy: 0.3305
Epoch 8/15
accuracy: 0.9005 - val_loss: 4.1311 - val_accuracy: 0.5030
```

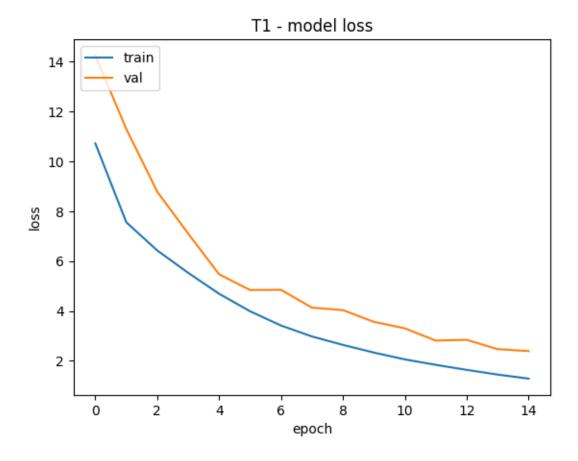
```
Epoch 9/15
accuracy: 0.9245 - val_loss: 4.0338 - val_accuracy: 0.4355
Epoch 10/15
accuracy: 0.9380 - val_loss: 3.5626 - val_accuracy: 0.5209
accuracy: 0.9595 - val_loss: 3.3018 - val_accuracy: 0.5354
Epoch 12/15
accuracy: 0.9650 - val_loss: 2.8098 - val_accuracy: 0.5952
Epoch 13/15
accuracy: 0.9760 - val_loss: 2.8411 - val_accuracy: 0.5517
Epoch 14/15
accuracy: 0.9840 - val_loss: 2.4654 - val_accuracy: 0.6362
Epoch 15/15
accuracy: 0.9925 - val_loss: 2.3897 - val_accuracy: 0.6243
```

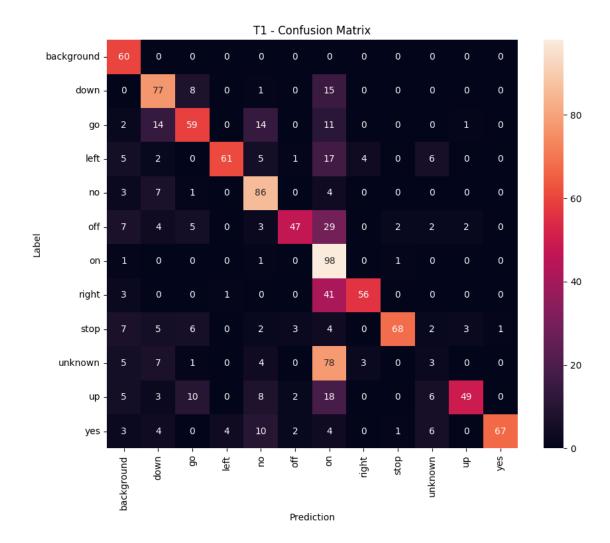
T1.B - Plot Training History and Confusion Matrix Here, we plot the training history and confusion matrix of the baseline model.

```
[7]: # summarize history for accuracy
     plt.plot(history.history['accuracy'])
     plt.plot(history.history['val_accuracy'])
     plt.title('T1 - model accuracy')
     plt.ylabel('accuracy')
     plt.xlabel('epoch')
     plt.legend(['train', 'val'], loc='upper left')
     plt.show()
     # summarize history for loss
     plt.plot(history.history['loss'])
     plt.plot(history.history['val_loss'])
     plt.title('T1 - model loss')
     plt.ylabel('loss')
     plt.xlabel('epoch')
     plt.legend(['train', 'val'], loc='upper left')
     plt.show()
     # Print accuracy
     score = task1_model.evaluate(x_val, y_val, verbose=0)
     print('T1 - validation accuracy:', score[1])
     # Print confusion matrix
```









As we can see, the model is not performing well on the validation set, especially for the class 'unknown'. But that is expected as the class is a blanket of words not detected by any other classes. Overall, this baseline model achieves a validation accuracy of around 0.6.

We now proceed to the next task.

Task 2: Random Search for Hyperparameter Tuning We now use random search to find the best hyperparameters for the baseline model. Random search is better than grid search because it is faster and more efficient. Here, we use the RandomSearchCV class from the scikit-learn library to find the best hyperparameters for the baseline model.

```
[8]: # define the random search parameters
param_grid = {
    'num_layers': hidden_num_layers,
    'num_filters': hidden_num_filters,
}
```

```
# Create a KerasClassifier
task2_model = KerasClassifier(model=t1_model, epochs=40, batch_size=BATCH_SIZE,__
 overbose=1, num_layers=hidden_num_layers, num_filters=hidden_num_filters)
# Create a RandomizedSearchCV
random search = RandomizedSearchCV(estimator=task2 model,
 →param_distributions=param_grid, n_iter=2, cv=3, verbose=2)
# Fit the RandomizedSearchCV
random_result = random_search.fit(x_train, y_train, validation_data=(x_val,_
 →y val), callbacks=[early stopping], verbose=1)
# Summarize results
print("Best: %f using %s" % (random_result.best_score_, random_result.
 ⇔best_params_))
Fitting 3 folds for each of 2 candidates, totalling 6 fits
accuracy: 0.2264 - val_loss: 13.8767 - val_accuracy: 0.0931
Epoch 2/40
accuracy: 0.2879 - val_loss: 11.5925 - val_accuracy: 0.1178
Epoch 3/40
accuracy: 0.3598 - val_loss: 9.1822 - val_accuracy: 0.1776
Epoch 4/40
accuracy: 0.4558 - val_loss: 8.1694 - val_accuracy: 0.1067
Epoch 5/40
accuracy: 0.5840 - val_loss: 7.3425 - val_accuracy: 0.1119
Epoch 6/40
accuracy: 0.6769 - val_loss: 6.1923 - val_accuracy: 0.1827
Epoch 7/40
accuracy: 0.7774 - val_loss: 6.1963 - val_accuracy: 0.1324
Epoch 8/40
accuracy: 0.8658 - val_loss: 5.4464 - val_accuracy: 0.1887
Epoch 9/40
accuracy: 0.8808 - val_loss: 4.8642 - val_accuracy: 0.3501
Epoch 10/40
```

```
accuracy: 0.9130 - val_loss: 4.4623 - val_accuracy: 0.4330
Epoch 11/40
accuracy: 0.9295 - val_loss: 4.4548 - val_accuracy: 0.3322
Epoch 12/40
accuracy: 0.9498 - val_loss: 3.8134 - val_accuracy: 0.5431
Epoch 13/40
accuracy: 0.9580 - val_loss: 3.9290 - val_accuracy: 0.4236
Epoch 14/40
accuracy: 0.9670 - val_loss: 3.4745 - val_accuracy: 0.5107
Epoch 15/40
accuracy: 0.9745 - val_loss: 3.4132 - val_accuracy: 0.4833
Epoch 16/40
accuracy: 0.9880 - val_loss: 3.3338 - val_accuracy: 0.4646
Epoch 17/40
accuracy: 0.9910 - val_loss: 2.8684 - val_accuracy: 0.5935
Epoch 18/40
accuracy: 0.9963 - val_loss: 2.7534 - val_accuracy: 0.5969
Epoch 19/40
accuracy: 0.9985 - val_loss: 2.6736 - val_accuracy: 0.5858
accuracy: 0.9993 - val_loss: 2.5792 - val_accuracy: 0.5816
Epoch 21/40
accuracy: 0.9985 - val_loss: 2.3766 - val_accuracy: 0.6243
Epoch 22/40
accuracy: 0.9985 - val loss: 2.5306 - val accuracy: 0.5722
Epoch 23/40
accuracy: 0.9993 - val_loss: 2.1540 - val_accuracy: 0.6345
Epoch 24/40
accuracy: 1.0000 - val_loss: 2.0190 - val_accuracy: 0.6550
Epoch 25/40
11/11 [============ ] - 2s 141ms/step - loss: 0.8371 -
accuracy: 1.0000 - val_loss: 1.9569 - val_accuracy: 0.6524
Epoch 26/40
```

```
accuracy: 1.0000 - val_loss: 1.8787 - val_accuracy: 0.6670
Epoch 27/40
accuracy: 1.0000 - val_loss: 1.8113 - val_accuracy: 0.6635
Epoch 28/40
accuracy: 1.0000 - val_loss: 1.7217 - val_accuracy: 0.6832
Epoch 29/40
accuracy: 1.0000 - val_loss: 1.6801 - val_accuracy: 0.6892
Epoch 30/40
accuracy: 1.0000 - val_loss: 1.6554 - val_accuracy: 0.6874
Epoch 31/40
accuracy: 1.0000 - val_loss: 1.5761 - val_accuracy: 0.6892
Epoch 32/40
accuracy: 1.0000 - val_loss: 1.5520 - val_accuracy: 0.6943
Epoch 33/40
accuracy: 1.0000 - val_loss: 1.4838 - val_accuracy: 0.6994
Epoch 34/40
accuracy: 0.9993 - val_loss: 1.5233 - val_accuracy: 0.6772
Epoch 35/40
accuracy: 0.9993 - val_loss: 1.5073 - val_accuracy: 0.6951
accuracy: 1.0000 - val_loss: 1.4462 - val_accuracy: 0.6781
Epoch 37/40
accuracy: 1.0000 - val_loss: 1.5108 - val_accuracy: 0.6832
Epoch 38/40
accuracy: 1.0000 - val loss: 1.3635 - val accuracy: 0.6823
Epoch 39/40
accuracy: 1.0000 - val_loss: 1.3346 - val_accuracy: 0.6849
Epoch 40/40
accuracy: 1.0000 - val_loss: 1.3071 - val_accuracy: 0.6943
6/6 [======= ] - 1s 129ms/step
[CV] END ...num_filters=128, num_layers=4; total time= 1.4min
Epoch 1/40
accuracy: 0.2226 - val_loss: 25.4376 - val_accuracy: 0.1221
```

```
Epoch 2/40
accuracy: 0.3471 - val_loss: 32.4891 - val_accuracy: 0.0999
Epoch 3/40
accuracy: 0.4865 - val_loss: 24.1249 - val_accuracy: 0.0956
accuracy: 0.5982 - val_loss: 21.2051 - val_accuracy: 0.1008
Epoch 5/40
accuracy: 0.6844 - val_loss: 14.9276 - val_accuracy: 0.1067
Epoch 6/40
accuracy: 0.7669 - val_loss: 11.2893 - val_accuracy: 0.1076
Epoch 7/40
accuracy: 0.8141 - val_loss: 9.7536 - val_accuracy: 0.1076
Epoch 8/40
accuracy: 0.8538 - val_loss: 7.1853 - val_accuracy: 0.1324
Epoch 9/40
accuracy: 0.8823 - val_loss: 5.6822 - val_accuracy: 0.1862
Epoch 10/40
accuracy: 0.9273 - val_loss: 4.2388 - val_accuracy: 0.4406
Epoch 11/40
accuracy: 0.9430 - val_loss: 3.9191 - val_accuracy: 0.5158
Epoch 12/40
accuracy: 0.9633 - val_loss: 3.8045 - val_accuracy: 0.4483
Epoch 13/40
accuracy: 0.9723 - val_loss: 3.3968 - val_accuracy: 0.5722
Epoch 14/40
accuracy: 0.9738 - val_loss: 3.2807 - val_accuracy: 0.5525
Epoch 15/40
accuracy: 0.9858 - val_loss: 3.1062 - val_accuracy: 0.5739
Epoch 16/40
accuracy: 0.9910 - val_loss: 2.8455 - val_accuracy: 0.6225
Epoch 17/40
accuracy: 0.9888 - val_loss: 2.8030 - val_accuracy: 0.6183
```

```
Epoch 18/40
accuracy: 0.9903 - val_loss: 2.6260 - val_accuracy: 0.6063
Epoch 19/40
accuracy: 0.9948 - val_loss: 2.4277 - val_accuracy: 0.6465
Epoch 20/40
accuracy: 0.9955 - val_loss: 2.5712 - val_accuracy: 0.5414
Epoch 21/40
accuracy: 0.9948 - val_loss: 2.5747 - val_accuracy: 0.5064
Epoch 22/40
accuracy: 0.9993 - val_loss: 2.3225 - val_accuracy: 0.5909
Epoch 23/40
11/11 [============ ] - 2s 141ms/step - loss: 0.9595 -
accuracy: 1.0000 - val_loss: 2.2104 - val_accuracy: 0.6490
Epoch 24/40
accuracy: 0.9993 - val_loss: 1.9860 - val_accuracy: 0.6670
Epoch 25/40
accuracy: 1.0000 - val_loss: 1.9189 - val_accuracy: 0.6661
Epoch 26/40
accuracy: 1.0000 - val_loss: 1.9123 - val_accuracy: 0.6541
Epoch 27/40
accuracy: 1.0000 - val_loss: 1.9176 - val_accuracy: 0.6388
Epoch 28/40
accuracy: 0.9993 - val_loss: 1.6978 - val_accuracy: 0.7062
Epoch 29/40
accuracy: 1.0000 - val_loss: 1.7061 - val_accuracy: 0.7062
Epoch 30/40
accuracy: 1.0000 - val_loss: 1.5055 - val_accuracy: 0.7225
Epoch 31/40
accuracy: 1.0000 - val_loss: 1.4667 - val_accuracy: 0.7216
accuracy: 1.0000 - val_loss: 1.4737 - val_accuracy: 0.7054
Epoch 33/40
accuracy: 1.0000 - val_loss: 1.4606 - val_accuracy: 0.7114
```

```
Epoch 34/40
accuracy: 1.0000 - val_loss: 1.3770 - val_accuracy: 0.7276
Epoch 35/40
accuracy: 1.0000 - val_loss: 1.3526 - val_accuracy: 0.7284
accuracy: 1.0000 - val_loss: 1.3202 - val_accuracy: 0.7293
Epoch 37/40
accuracy: 1.0000 - val_loss: 1.3188 - val_accuracy: 0.7259
Epoch 38/40
accuracy: 1.0000 - val_loss: 1.2920 - val_accuracy: 0.7225
Epoch 39/40
11/11 [============ ] - 2s 145ms/step - loss: 0.2463 -
accuracy: 1.0000 - val_loss: 1.2630 - val_accuracy: 0.7361
Epoch 40/40
accuracy: 1.0000 - val_loss: 1.2392 - val_accuracy: 0.7455
6/6 [======== ] - Os 37ms/step
[CV] END ...num_filters=128, num_layers=4; total time= 1.4min
Epoch 1/40
accuracy: 0.2376 - val_loss: 32.8397 - val_accuracy: 0.0854
Epoch 2/40
accuracy: 0.3126 - val_loss: 20.6583 - val_accuracy: 0.1289
Epoch 3/40
accuracy: 0.4093 - val_loss: 19.3227 - val_accuracy: 0.1281
Epoch 4/40
accuracy: 0.5030 - val_loss: 13.2528 - val_accuracy: 0.1452
Epoch 5/40
accuracy: 0.5930 - val_loss: 12.8476 - val_accuracy: 0.1076
Epoch 6/40
accuracy: 0.6777 - val_loss: 11.3853 - val_accuracy: 0.1059
Epoch 7/40
accuracy: 0.7586 - val_loss: 9.2108 - val_accuracy: 0.1093
Epoch 8/40
accuracy: 0.8073 - val_loss: 7.0090 - val_accuracy: 0.1127
Epoch 9/40
```

```
accuracy: 0.8568 - val_loss: 5.5443 - val_accuracy: 0.1785
Epoch 10/40
accuracy: 0.9063 - val_loss: 4.5565 - val_accuracy: 0.3604
Epoch 11/40
accuracy: 0.9235 - val_loss: 4.2615 - val_accuracy: 0.4031
Epoch 12/40
accuracy: 0.9483 - val_loss: 4.0213 - val_accuracy: 0.4355
Epoch 13/40
accuracy: 0.9588 - val_loss: 3.5896 - val_accuracy: 0.5406
Epoch 14/40
accuracy: 0.9700 - val_loss: 3.5604 - val_accuracy: 0.4996
Epoch 15/40
accuracy: 0.9768 - val_loss: 3.2232 - val_accuracy: 0.5491
Epoch 16/40
accuracy: 0.9888 - val_loss: 2.9803 - val_accuracy: 0.5816
Epoch 17/40
accuracy: 0.9895 - val_loss: 3.1183 - val_accuracy: 0.4611
Epoch 18/40
accuracy: 0.9903 - val_loss: 2.8769 - val_accuracy: 0.5115
Epoch 19/40
accuracy: 0.9910 - val_loss: 2.8432 - val_accuracy: 0.5354
Epoch 20/40
accuracy: 0.9963 - val_loss: 2.8132 - val_accuracy: 0.4851
Epoch 21/40
accuracy: 0.9970 - val_loss: 2.6385 - val_accuracy: 0.5107
Epoch 22/40
accuracy: 0.9985 - val_loss: 2.3351 - val_accuracy: 0.5961
Epoch 23/40
accuracy: 0.9993 - val_loss: 2.2147 - val_accuracy: 0.6097
Epoch 24/40
accuracy: 0.9993 - val_loss: 1.9705 - val_accuracy: 0.6695
Epoch 25/40
```

```
accuracy: 0.9993 - val_loss: 1.8940 - val_accuracy: 0.6772
Epoch 26/40
accuracy: 0.9993 - val_loss: 1.8901 - val_accuracy: 0.6422
Epoch 27/40
accuracy: 0.9993 - val_loss: 1.8182 - val_accuracy: 0.6576
Epoch 28/40
accuracy: 1.0000 - val_loss: 1.6874 - val_accuracy: 0.6857
Epoch 29/40
accuracy: 1.0000 - val_loss: 1.6397 - val_accuracy: 0.6840
Epoch 30/40
accuracy: 1.0000 - val_loss: 1.5512 - val_accuracy: 0.6977
Epoch 31/40
accuracy: 1.0000 - val_loss: 1.4495 - val_accuracy: 0.7199
Epoch 32/40
accuracy: 1.0000 - val_loss: 1.4352 - val_accuracy: 0.7105
Epoch 33/40
accuracy: 1.0000 - val_loss: 1.3737 - val_accuracy: 0.7199
Epoch 34/40
accuracy: 1.0000 - val_loss: 1.3875 - val_accuracy: 0.7062
Epoch 35/40
accuracy: 1.0000 - val_loss: 1.3131 - val_accuracy: 0.7182
Epoch 36/40
accuracy: 1.0000 - val_loss: 1.3124 - val_accuracy: 0.7131
Epoch 37/40
accuracy: 1.0000 - val_loss: 1.2455 - val_accuracy: 0.7250
Epoch 38/40
accuracy: 1.0000 - val_loss: 1.2425 - val_accuracy: 0.7156
Epoch 39/40
accuracy: 1.0000 - val_loss: 1.1800 - val_accuracy: 0.7242
Epoch 40/40
accuracy: 1.0000 - val_loss: 1.2018 - val_accuracy: 0.7250
6/6 [======= ] - Os 37ms/step
```

```
[CV] END ...num_filters=128, num_layers=4; total time= 1.4min
Epoch 1/40
accuracy: 0.2159 - val_loss: 144.2063 - val_accuracy: 0.1153
Epoch 2/40
accuracy: 0.2639 - val_loss: 45.9289 - val_accuracy: 0.1512
Epoch 3/40
accuracy: 0.3006 - val_loss: 24.7214 - val_accuracy: 0.1529
Epoch 4/40
accuracy: 0.3568 - val_loss: 17.7981 - val_accuracy: 0.1887
Epoch 5/40
accuracy: 0.4063 - val_loss: 16.0583 - val_accuracy: 0.1930
Epoch 6/40
accuracy: 0.4183 - val_loss: 14.1732 - val_accuracy: 0.1879
Epoch 7/40
accuracy: 0.4610 - val_loss: 12.8334 - val_accuracy: 0.2169
Epoch 8/40
accuracy: 0.4700 - val_loss: 11.5747 - val_accuracy: 0.2118
Epoch 9/40
accuracy: 0.4715 - val_loss: 10.7535 - val_accuracy: 0.2400
accuracy: 0.5150 - val_loss: 9.6990 - val_accuracy: 0.2477
Epoch 11/40
accuracy: 0.5217 - val_loss: 9.0288 - val_accuracy: 0.2579
Epoch 12/40
accuracy: 0.5607 - val loss: 8.4303 - val accuracy: 0.2775
Epoch 13/40
accuracy: 0.5660 - val_loss: 7.7082 - val_accuracy: 0.2946
Epoch 14/40
accuracy: 0.5937 - val_loss: 7.2469 - val_accuracy: 0.2921
Epoch 15/40
11/11 [=========== ] - 1s 137ms/step - loss: 6.0389 -
accuracy: 0.5982 - val_loss: 6.7174 - val_accuracy: 0.3023
Epoch 16/40
```

```
accuracy: 0.6454 - val_loss: 6.4784 - val_accuracy: 0.2886
Epoch 17/40
accuracy: 0.6522 - val_loss: 6.2586 - val_accuracy: 0.3194
Epoch 18/40
accuracy: 0.6949 - val_loss: 5.7913 - val_accuracy: 0.2946
Epoch 19/40
accuracy: 0.7421 - val_loss: 5.4281 - val_accuracy: 0.3202
Epoch 20/40
accuracy: 0.7429 - val_loss: 5.1881 - val_accuracy: 0.3100
Epoch 21/40
accuracy: 0.7646 - val_loss: 4.8687 - val_accuracy: 0.3553
Epoch 22/40
accuracy: 0.7954 - val_loss: 4.4096 - val_accuracy: 0.3945
Epoch 23/40
accuracy: 0.8193 - val_loss: 4.3398 - val_accuracy: 0.3766
Epoch 24/40
accuracy: 0.8433 - val_loss: 4.0212 - val_accuracy: 0.4048
Epoch 25/40
accuracy: 0.8553 - val_loss: 3.9478 - val_accuracy: 0.4022
accuracy: 0.8583 - val_loss: 3.7463 - val_accuracy: 0.4099
Epoch 27/40
accuracy: 0.8966 - val_loss: 3.5105 - val_accuracy: 0.4193
Epoch 28/40
accuracy: 0.9048 - val loss: 3.4258 - val accuracy: 0.4039
Epoch 29/40
accuracy: 0.9160 - val_loss: 3.2944 - val_accuracy: 0.4236
Epoch 30/40
accuracy: 0.9280 - val_loss: 3.3306 - val_accuracy: 0.3655
Epoch 31/40
11/11 [============ ] - 1s 138ms/step - loss: 1.5650 -
accuracy: 0.9453 - val_loss: 3.1694 - val_accuracy: 0.3971
Epoch 32/40
```

```
accuracy: 0.9550 - val_loss: 3.0431 - val_accuracy: 0.4184
Epoch 33/40
accuracy: 0.9528 - val_loss: 3.0111 - val_accuracy: 0.3757
Epoch 34/40
accuracy: 0.9678 - val_loss: 2.9146 - val_accuracy: 0.4133
Epoch 35/40
accuracy: 0.9520 - val_loss: 2.6262 - val_accuracy: 0.4509
Epoch 36/40
accuracy: 0.9685 - val_loss: 2.6297 - val_accuracy: 0.4458
Epoch 37/40
accuracy: 0.9745 - val_loss: 2.7664 - val_accuracy: 0.4150
Epoch 38/40
accuracy: 0.9783 - val_loss: 2.5347 - val_accuracy: 0.4330
Epoch 39/40
accuracy: 0.9805 - val_loss: 2.5638 - val_accuracy: 0.4210
Epoch 40/40
accuracy: 0.9813 - val_loss: 2.4638 - val_accuracy: 0.4227
6/6 [=======] - 0s 26ms/step
[CV] END ...num_filters=128, num_layers=2; total time= 1.0min
Epoch 1/40
accuracy: 0.1837 - val_loss: 62.3380 - val_accuracy: 0.1230
Epoch 2/40
accuracy: 0.2721 - val_loss: 34.0311 - val_accuracy: 0.1238
Epoch 3/40
accuracy: 0.2826 - val_loss: 22.0282 - val_accuracy: 0.1418
Epoch 4/40
accuracy: 0.3433 - val_loss: 17.6936 - val_accuracy: 0.1605
Epoch 5/40
accuracy: 0.3966 - val_loss: 15.5702 - val_accuracy: 0.1751
accuracy: 0.4108 - val_loss: 14.0672 - val_accuracy: 0.1939
Epoch 7/40
accuracy: 0.4295 - val_loss: 12.3845 - val_accuracy: 0.1964
```

```
Epoch 8/40
accuracy: 0.4513 - val_loss: 11.2333 - val_accuracy: 0.2109
accuracy: 0.4828 - val_loss: 10.7125 - val_accuracy: 0.2058
Epoch 10/40
accuracy: 0.5052 - val_loss: 9.7285 - val_accuracy: 0.2280
Epoch 11/40
accuracy: 0.4993 - val_loss: 9.1299 - val_accuracy: 0.2272
Epoch 12/40
accuracy: 0.5150 - val_loss: 8.2954 - val_accuracy: 0.2425
Epoch 13/40
11/11 [=========== ] - 1s 138ms/step - loss: 7.0522 -
accuracy: 0.5435 - val_loss: 7.9627 - val_accuracy: 0.2374
Epoch 14/40
accuracy: 0.5645 - val_loss: 7.2452 - val_accuracy: 0.2878
Epoch 15/40
accuracy: 0.6019 - val_loss: 6.7965 - val_accuracy: 0.3040
Epoch 16/40
accuracy: 0.6364 - val_loss: 6.3979 - val_accuracy: 0.3057
Epoch 17/40
accuracy: 0.6297 - val_loss: 6.0695 - val_accuracy: 0.3049
Epoch 18/40
accuracy: 0.6582 - val_loss: 5.7635 - val_accuracy: 0.3040
Epoch 19/40
accuracy: 0.6829 - val_loss: 5.4927 - val_accuracy: 0.3117
Epoch 20/40
accuracy: 0.7069 - val_loss: 5.4290 - val_accuracy: 0.2921
Epoch 21/40
accuracy: 0.7609 - val_loss: 5.1855 - val_accuracy: 0.3049
accuracy: 0.7534 - val_loss: 4.9377 - val_accuracy: 0.3228
Epoch 23/40
accuracy: 0.7871 - val_loss: 4.6678 - val_accuracy: 0.3348
```

```
Epoch 24/40
accuracy: 0.7999 - val_loss: 4.3308 - val_accuracy: 0.3493
Epoch 25/40
accuracy: 0.8133 - val_loss: 4.2266 - val_accuracy: 0.3612
accuracy: 0.8096 - val_loss: 4.1021 - val_accuracy: 0.3442
Epoch 27/40
accuracy: 0.8651 - val_loss: 4.1683 - val_accuracy: 0.3219
Epoch 28/40
accuracy: 0.8583 - val_loss: 3.7881 - val_accuracy: 0.3672
Epoch 29/40
11/11 [=========== ] - 1s 133ms/step - loss: 2.0500 -
accuracy: 0.8741 - val_loss: 3.9174 - val_accuracy: 0.3177
Epoch 30/40
accuracy: 0.8861 - val_loss: 3.7989 - val_accuracy: 0.3271
Epoch 31/40
accuracy: 0.9010 - val_loss: 3.7108 - val_accuracy: 0.3296
Epoch 32/40
accuracy: 0.9078 - val_loss: 3.5375 - val_accuracy: 0.3424
Epoch 33/40
accuracy: 0.9183 - val_loss: 3.5819 - val_accuracy: 0.3040
Epoch 34/40
accuracy: 0.9138 - val_loss: 3.3956 - val_accuracy: 0.3544
Epoch 35/40
accuracy: 0.9288 - val_loss: 3.3046 - val_accuracy: 0.3450
Epoch 36/40
accuracy: 0.9363 - val_loss: 3.0731 - val_accuracy: 0.3655
Epoch 37/40
accuracy: 0.9280 - val_loss: 3.1597 - val_accuracy: 0.3450
Epoch 38/40
accuracy: 0.9460 - val_loss: 3.0413 - val_accuracy: 0.3476
Epoch 39/40
accuracy: 0.9453 - val_loss: 2.7230 - val_accuracy: 0.4330
```

```
Epoch 40/40
accuracy: 0.9565 - val_loss: 2.8738 - val_accuracy: 0.3903
6/6 [======== ] - 0s 25ms/step
[CV] END ...num_filters=128, num_layers=2; total time= 1.0min
Epoch 1/40
accuracy: 0.2016 - val_loss: 241.7034 - val_accuracy: 0.1050
Epoch 2/40
accuracy: 0.2399 - val_loss: 81.2366 - val_accuracy: 0.1161
Epoch 3/40
accuracy: 0.3163 - val_loss: 32.8363 - val_accuracy: 0.1623
accuracy: 0.3501 - val_loss: 21.4714 - val_accuracy: 0.1443
Epoch 5/40
accuracy: 0.3951 - val_loss: 16.9548 - val_accuracy: 0.1879
accuracy: 0.4033 - val_loss: 14.7623 - val_accuracy: 0.1947
Epoch 7/40
accuracy: 0.4370 - val_loss: 13.0904 - val_accuracy: 0.1956
Epoch 8/40
accuracy: 0.4408 - val_loss: 11.5762 - val_accuracy: 0.2408
Epoch 9/40
accuracy: 0.4723 - val_loss: 10.7658 - val_accuracy: 0.2400
Epoch 10/40
accuracy: 0.4708 - val_loss: 9.8572 - val_accuracy: 0.2383
Epoch 11/40
accuracy: 0.4933 - val_loss: 9.3581 - val_accuracy: 0.2143
Epoch 12/40
accuracy: 0.5120 - val_loss: 8.4432 - val_accuracy: 0.2511
Epoch 13/40
accuracy: 0.5097 - val_loss: 7.9257 - val_accuracy: 0.2716
Epoch 14/40
accuracy: 0.5345 - val_loss: 7.3791 - val_accuracy: 0.2827
Epoch 15/40
```

```
accuracy: 0.5390 - val_loss: 6.9679 - val_accuracy: 0.2895
Epoch 16/40
accuracy: 0.5795 - val_loss: 6.6862 - val_accuracy: 0.2656
Epoch 17/40
accuracy: 0.5930 - val_loss: 6.1836 - val_accuracy: 0.2844
Epoch 18/40
accuracy: 0.6319 - val_loss: 5.9577 - val_accuracy: 0.2699
Epoch 19/40
accuracy: 0.6514 - val_loss: 5.6340 - val_accuracy: 0.2792
Epoch 20/40
accuracy: 0.6882 - val_loss: 5.7783 - val_accuracy: 0.2383
Epoch 21/40
accuracy: 0.7001 - val_loss: 5.1679 - val_accuracy: 0.2724
Epoch 22/40
accuracy: 0.7151 - val_loss: 5.1495 - val_accuracy: 0.2630
Epoch 23/40
accuracy: 0.7699 - val_loss: 4.9527 - val_accuracy: 0.2442
Epoch 24/40
accuracy: 0.7909 - val_loss: 4.7208 - val_accuracy: 0.2673
Epoch 25/40
accuracy: 0.8448 - val_loss: 4.8564 - val_accuracy: 0.2417
Epoch 26/40
accuracy: 0.8568 - val loss: 4.7041 - val accuracy: 0.2485
Epoch 27/40
accuracy: 0.8943 - val_loss: 4.4139 - val_accuracy: 0.2716
Epoch 28/40
accuracy: 0.9033 - val_loss: 4.2696 - val_accuracy: 0.2844
Epoch 29/40
accuracy: 0.9018 - val_loss: 4.1188 - val_accuracy: 0.2699
Epoch 30/40
accuracy: 0.9183 - val_loss: 3.9333 - val_accuracy: 0.2562
Epoch 31/40
```

```
accuracy: 0.9235 - val_loss: 3.9512 - val_accuracy: 0.2605
Epoch 32/40
accuracy: 0.9295 - val_loss: 3.7759 - val_accuracy: 0.2647
Epoch 33/40
accuracy: 0.9378 - val_loss: 3.7879 - val_accuracy: 0.2699
Epoch 34/40
accuracy: 0.9438 - val_loss: 3.6823 - val_accuracy: 0.2707
Epoch 35/40
accuracy: 0.9340 - val_loss: 3.2742 - val_accuracy: 0.3202
Epoch 36/40
accuracy: 0.9460 - val_loss: 3.2655 - val_accuracy: 0.3433
Epoch 37/40
accuracy: 0.9423 - val_loss: 3.2585 - val_accuracy: 0.3143
Epoch 38/40
accuracy: 0.9565 - val_loss: 3.5053 - val_accuracy: 0.2921
Epoch 39/40
accuracy: 0.9603 - val_loss: 3.5056 - val_accuracy: 0.2792
Epoch 40/40
accuracy: 0.9693 - val_loss: 3.0227 - val_accuracy: 0.3595
6/6 [=======] - Os 35ms/step
[CV] END ...num_filters=128, num_layers=2; total time= 1.4min
Epoch 1/40
accuracy: 0.2384 - val_loss: 27.2533 - val_accuracy: 0.0871
accuracy: 0.3303 - val loss: 11.6465 - val accuracy: 0.1033
Epoch 3/40
accuracy: 0.4863 - val_loss: 8.6276 - val_accuracy: 0.1443
Epoch 4/40
accuracy: 0.6297 - val_loss: 7.6298 - val_accuracy: 0.1076
Epoch 5/40
16/16 [============= ] - 2s 137ms/step - loss: 4.5353 -
accuracy: 0.7356 - val_loss: 5.8419 - val_accuracy: 0.1648
Epoch 6/40
```

```
accuracy: 0.8191 - val_loss: 4.6807 - val_accuracy: 0.4449
Epoch 7/40
accuracy: 0.8756 - val_loss: 4.4571 - val_accuracy: 0.4031
Epoch 8/40
accuracy: 0.9135 - val_loss: 4.0164 - val_accuracy: 0.4791
Epoch 9/40
accuracy: 0.9425 - val_loss: 3.6529 - val_accuracy: 0.4774
Epoch 10/40
accuracy: 0.9620 - val_loss: 3.2199 - val_accuracy: 0.5884
Epoch 11/40
accuracy: 0.9740 - val_loss: 3.0913 - val_accuracy: 0.5585
Epoch 12/40
accuracy: 0.9725 - val_loss: 2.7496 - val_accuracy: 0.6157
Epoch 13/40
accuracy: 0.9845 - val_loss: 2.4136 - val_accuracy: 0.6610
Epoch 14/40
accuracy: 0.9840 - val_loss: 2.2252 - val_accuracy: 0.6823
Epoch 15/40
accuracy: 0.9885 - val_loss: 2.1646 - val_accuracy: 0.6627
accuracy: 0.9890 - val_loss: 1.9213 - val_accuracy: 0.6789
Epoch 17/40
accuracy: 0.9925 - val_loss: 2.1592 - val_accuracy: 0.6020
Epoch 18/40
accuracy: 0.9970 - val loss: 1.6989 - val accuracy: 0.7259
Epoch 19/40
accuracy: 0.9990 - val_loss: 1.6790 - val_accuracy: 0.6994
Epoch 20/40
accuracy: 0.9985 - val_loss: 1.4915 - val_accuracy: 0.7259
Epoch 21/40
16/16 [============ ] - 2s 138ms/step - loss: 0.5799 -
accuracy: 0.9995 - val_loss: 1.4173 - val_accuracy: 0.7267
Epoch 22/40
```

```
accuracy: 0.9985 - val_loss: 1.5078 - val_accuracy: 0.6934
Epoch 23/40
accuracy: 1.0000 - val_loss: 1.2565 - val_accuracy: 0.7506
Epoch 24/40
accuracy: 0.9995 - val_loss: 1.2706 - val_accuracy: 0.7387
Epoch 25/40
accuracy: 0.9950 - val_loss: 1.2062 - val_accuracy: 0.7361
Epoch 26/40
accuracy: 0.9970 - val_loss: 1.2706 - val_accuracy: 0.7199
Epoch 27/40
accuracy: 0.9940 - val_loss: 1.3983 - val_accuracy: 0.6874
Epoch 28/40
accuracy: 0.9925 - val_loss: 1.3316 - val_accuracy: 0.6934
Epoch 29/40
accuracy: 0.9810 - val_loss: 1.3649 - val_accuracy: 0.7079
Epoch 30/40
accuracy: 0.9705 - val_loss: 1.5190 - val_accuracy: 0.6627
Epoch 31/40
accuracy: 0.9605 - val_loss: 1.5947 - val_accuracy: 0.6567
accuracy: 0.9590 - val_loss: 1.9958 - val_accuracy: 0.6191
Epoch 33/40
accuracy: 0.9545 - val_loss: 2.5672 - val_accuracy: 0.4979
Epoch 34/40
accuracy: 0.9650 - val loss: 1.7657 - val accuracy: 0.6260
Epoch 35/40
accuracy: 0.9820 - val_loss: 1.8249 - val_accuracy: 0.6755
Epoch 36/40
accuracy: 0.9860 - val_loss: 1.5537 - val_accuracy: 0.6781
Epoch 37/40
16/16 [============ ] - 2s 139ms/step - loss: 0.2114 -
accuracy: 0.9890 - val_loss: 1.6634 - val_accuracy: 0.6550
Epoch 38/40
```

As seen during the cross validation split of the Randomized Search for the best fit, the accuracy of the model has improved to 0.84. The best values for the hyperparameters are: * num_layers: 4 * num_filters: 128

We now redefine the model with these hyperparameters.

```
[9]: hidden_num_filters = [128, 128, 128, 128]
hidden_num_layers = [2, 3, 4, 5]

task2_model = t1_model(len(hidden_num_layers), hidden_num_filters,__
passthrough=True)
task2_model.summary()
```

Model: "sequential_8"

Layer (type)		Param #
conv2d_34 (Conv2D)		320
<pre>batch_normalization_34 (Ba tchNormalization)</pre>	(None, 98, 50, 32)	128
conv2d_35 (Conv2D)	(None, 98, 50, 128)	36992
<pre>batch_normalization_35 (Ba tchNormalization)</pre>	(None, 98, 50, 128)	512
<pre>max_pooling2d_34 (MaxPooli ng2D)</pre>	(None, 49, 25, 128)	0
conv2d_36 (Conv2D)	(None, 49, 25, 128)	147584
<pre>batch_normalization_36 (Ba tchNormalization)</pre>	(None, 49, 25, 128)	512
<pre>max_pooling2d_35 (MaxPooli ng2D)</pre>	(None, 25, 13, 128)	0
conv2d_37 (Conv2D)	(None, 25, 13, 128)	147584

```
batch_normalization_37 (Ba (None, 25, 13, 128)
                                                       512
tchNormalization)
max_pooling2d_36 (MaxPooli (None, 13, 7, 128)
                                                       0
ng2D)
conv2d_38 (Conv2D)
                             (None, 13, 7, 128)
                                                       147584
batch_normalization_38 (Ba (None, 13, 7, 128)
                                                       512
tchNormalization)
max_pooling2d_37 (MaxPooli
                            (None, 7, 4, 128)
                                                       0
ng2D)
max_pooling2d_38 (MaxPooli (None, 1, 4, 128)
                                                       0
ng2D)
flatten_8 (Flatten)
                            (None, 512)
                                                       0
                            (None, 1024)
dense 16 (Dense)
                                                       525312
dropout 8 (Dropout)
                             (None, 1024)
dense 17 (Dense)
                             (None, 12)
                                                       12300
```

Total params: 1019852 (3.89 MB)
Trainable params: 1018764 (3.89 MB)
Non-trainable params: 1088 (4.25 KB)

Task 3: Model Averaging Scheme Here, we randomly sample three subsets of the training data and train the T2 model on each of them. We then proceed with a series of voting schemes to determine the final prediction.

```
[10]: # define number of samples
    train_samples = 2001

# create data index
    data_index = list(range(1, train_samples))

subsets = 3

# keep track of history
    shuffle_history = []

for i in range(subsets):
```

```
# create random index using sampling with replacement
    idx = random.choices(data_index, k=train_samples)
    # define first shuffle
    x_train_shuffle = np.zeros([train_samples, IMG_SIZE[0], IMG_SIZE[1], 1])
    y_train_shuffle = np.zeros([train_samples, NUM_CLASSES])
    # resample the data
    for j in range(train samples):
       x_train_shuffle[j] = x_train[idx[j], :, :, :]
       y_train_shuffle[j] = y_train[idx[j], :]
    # train the model
    shuffle_history.append(task2_model.fit(x_train_shuffle, y_train_shuffle,_u
 →batch_size=BATCH_SIZE, epochs=40, validation_data=(x_val, y_val),
 ⇔callbacks=[early_stopping]))
# Take the majority class prediction and use the mode for all the three models,
 ⇔to determine final prediction
# Create a matrix of predictions
y_pred = np.zeros([len(y_val), subsets])
# iterate through the models
for i in range(subsets):
    # save the predictions
    y_pred[:, i] = np.argmax(shuffle history[i].model.predict(x_val), axis=1)
# convert to integer
y_pred = np.array(y_pred, dtype=int)
# take the mode
y_pred = np.squeeze(np.apply_along_axis(lambda x: np.bincount(x).argmax(),_
 ⇒axis=1, arr=y_pred))
# Print accuracy
score = task2_model.evaluate(x_val, y_val, verbose=0)
print('T3 - validation accuracy:', score[1])
Epoch 1/40
accuracy: 0.2214 - val_loss: 18.5219 - val_accuracy: 0.1016
accuracy: 0.3593 - val_loss: 20.7082 - val_accuracy: 0.0897
accuracy: 0.4873 - val_loss: 16.3425 - val_accuracy: 0.1332
Epoch 4/40
```

```
accuracy: 0.6142 - val_loss: 12.7257 - val_accuracy: 0.1059
Epoch 5/40
accuracy: 0.7331 - val_loss: 8.8176 - val_accuracy: 0.1127
Epoch 6/40
accuracy: 0.8216 - val_loss: 6.2845 - val_accuracy: 0.1196
Epoch 7/40
accuracy: 0.8776 - val_loss: 5.4244 - val_accuracy: 0.1768
Epoch 8/40
accuracy: 0.9220 - val_loss: 4.4875 - val_accuracy: 0.3279
accuracy: 0.9575 - val_loss: 4.4722 - val_accuracy: 0.2570
Epoch 10/40
accuracy: 0.9645 - val_loss: 3.8638 - val_accuracy: 0.3595
Epoch 11/40
accuracy: 0.9735 - val_loss: 3.7465 - val_accuracy: 0.3595
Epoch 12/40
accuracy: 0.9840 - val_loss: 3.1284 - val_accuracy: 0.4927
Epoch 13/40
accuracy: 0.9900 - val_loss: 4.2781 - val_accuracy: 0.2605
Epoch 14/40
accuracy: 0.9900 - val_loss: 3.3047 - val_accuracy: 0.4304
Epoch 15/40
accuracy: 0.9950 - val loss: 3.0466 - val accuracy: 0.3903
Epoch 16/40
accuracy: 0.9960 - val_loss: 3.0104 - val_accuracy: 0.4150
Epoch 17/40
accuracy: 0.9985 - val_loss: 2.4636 - val_accuracy: 0.4774
Epoch 1/40
accuracy: 0.9160 - val_loss: 2.3834 - val_accuracy: 0.5278
Epoch 2/40
accuracy: 0.9465 - val_loss: 2.0721 - val_accuracy: 0.5918
Epoch 3/40
```

```
accuracy: 0.9590 - val_loss: 2.3790 - val_accuracy: 0.5705
Epoch 4/40
accuracy: 0.9805 - val_loss: 1.9279 - val_accuracy: 0.6567
Epoch 5/40
accuracy: 0.9880 - val_loss: 1.8948 - val_accuracy: 0.6832
Epoch 6/40
accuracy: 0.9960 - val_loss: 1.8606 - val_accuracy: 0.7037
Epoch 7/40
accuracy: 0.9950 - val_loss: 1.7802 - val_accuracy: 0.6892
accuracy: 0.9990 - val_loss: 1.7808 - val_accuracy: 0.6994
Epoch 9/40
accuracy: 0.9990 - val_loss: 1.6781 - val_accuracy: 0.6874
Epoch 10/40
accuracy: 1.0000 - val_loss: 1.5526 - val_accuracy: 0.6951
Epoch 11/40
accuracy: 1.0000 - val_loss: 1.4867 - val_accuracy: 0.7037
Epoch 1/40
accuracy: 0.9570 - val_loss: 1.5943 - val_accuracy: 0.6055
Epoch 2/40
accuracy: 0.9670 - val_loss: 1.5060 - val_accuracy: 0.6951
Epoch 3/40
accuracy: 0.9850 - val loss: 1.5928 - val accuracy: 0.6020
Epoch 4/40
accuracy: 0.9915 - val_loss: 1.4400 - val_accuracy: 0.7045
Epoch 5/40
accuracy: 0.9890 - val_loss: 1.7171 - val_accuracy: 0.6507
Epoch 6/40
accuracy: 0.9960 - val_loss: 1.4202 - val_accuracy: 0.7139
Epoch 7/40
accuracy: 0.9995 - val_loss: 1.4839 - val_accuracy: 0.7045
Epoch 8/40
```

```
accuracy: 0.9995 - val_loss: 1.2630 - val_accuracy: 0.7071
Epoch 9/40
accuracy: 0.9990 - val_loss: 1.2872 - val_accuracy: 0.7182
Epoch 10/40
accuracy: 0.9995 - val_loss: 1.2885 - val_accuracy: 0.7242
Epoch 11/40
accuracy: 0.9995 - val_loss: 1.1652 - val_accuracy: 0.7370
Epoch 12/40
accuracy: 0.9995 - val_loss: 1.1433 - val_accuracy: 0.7387
accuracy: 0.9995 - val_loss: 1.1431 - val_accuracy: 0.7395
Epoch 14/40
accuracy: 0.9995 - val_loss: 1.0792 - val_accuracy: 0.7515
Epoch 15/40
accuracy: 0.9995 - val_loss: 1.1098 - val_accuracy: 0.7404
Epoch 16/40
accuracy: 0.9995 - val_loss: 1.0631 - val_accuracy: 0.7455
Epoch 17/40
accuracy: 1.0000 - val_loss: 1.1018 - val_accuracy: 0.7421
Epoch 18/40
accuracy: 1.0000 - val_loss: 1.0780 - val_accuracy: 0.7387
Epoch 19/40
accuracy: 1.0000 - val loss: 1.0424 - val accuracy: 0.7421
37/37 [========] - Os 10ms/step
37/37 [========= ] - Os 8ms/step
37/37 [======== ] - Os 8ms/step
T3 - validation accuracy: 0.7421007752418518
```

After resampling for three times, the accuracy drops to about 0.75.

This is maybe an indication that resampling with replacement reduces the overall accuracy of the model due to a drop in the amount of data available for training.

0.1.5 Task 4: Hyperparameter Tuning using Bayesian Optimization

We now use Bayesian optimization to find the best hyperparameters for the advanced model.

Bayesian optimization is a probabilistic model-based optimization algorithm that is used to find the best hyperparameters for a model.

```
[12]: # Define the search space for hyperparameters
      space = {'num_layers': hp.choice('num_layers', [1, 2, 3, 4, 5]),
          'num_filters': hp.choice('num_filters', [8, 16, 32, 64, 128]),}
      # Define the objective function
      def t4 model(params):
          # number of convolutional filters
          input_num_filters = 32
          fully_connected_num_filters = 1024
          # define model
          model = Sequential()
          # input layer
          model.add(Input(shape=(IMG_SIZE[0], IMG_SIZE[1], 1)))
          model.add(Conv2D(input_num_filters, kernel_size =(3, 3), padding='same',u
       ⇔activation='relu'))
          model.add(BatchNormalization())
          model.add(MaxPooling2D(pool_size =(2, 2), strides=(2, 2), padding='same'))
          # hidden layers
          for _ in range(params['num_layers']):
              model.add(Conv2D(params['num filters'], kernel size =(3, 3),
       →padding='same', activation='relu'))
              model.add(BatchNormalization())
              model.add(MaxPooling2D(pool_size =(2, 2), strides=(2, 2),
       →padding='same'))
          # Time based pooling
          model.add(MaxPooling2D(pool_size=(TIMEPOOL_SIZE, 1),__
       ⇔strides=(TIMEPOOL_SIZE, 1), padding='same'))
          # fully connected layer
          model.add(Flatten())
          model.add(Dense(fully_connected_num_filters,_
       →kernel_regularizer=regularizers.12(0.01), activation='relu'))
          model.add(Dropout(0.2))
          # output layer
          model.add(Dense(NUM_CLASSES, activation='softmax'))
          # set adam optimizer
          opt = optimizers.Adam(learning rate=0.001)
```

```
0.9201
0.9214
0.9182
0.9199
0.9208
0.9206
0.9184
0.9195 - val_loss: 1.5420 - val_accuracy: 0.6396
Epoch 11/30
1/96 [...] - ETA: 2s - loss: 0.8127 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 0.6106 - accuracy:
0.9048
9/96 [=>...] - ETA: 1s - loss: 0.4971 - accuracy:
```

```
0.9259
12/96 [==>...] - ETA: 1s - loss: 0.4539 - accuracy:
0.9365
15/96 [===>...] - ETA: 1s - loss: 0.4452 - accuracy:
0.9333
18/96 [====>...] - ETA: 1s - loss: 0.4234 - accuracy:
22/96 [====>...] - ETA: 1s - loss: 0.4206 - accuracy:
0.9416
26/96 [======>...] - ETA: 1s - loss: 0.4238 - accuracy:
0.9322
30/96 [======>...] - ETA: 1s - loss: 0.4250 - accuracy:
0.9317
34/96 [=======>...] - ETA: Os - loss: 0.4076 - accuracy:
0.9370
37/96 [=======>...] - ETA: Os - loss: 0.3982 - accuracy:
0.9421
41/96 [========>...] - ETA: Os - loss: 0.4130 - accuracy:
0.9350
0.9340
0.9324
0.9326
0.9302
0.9326
0.9333
0.9316
0.9332
96/96 [============= ] - ETA: Os - loss: 0.4272 - accuracy:
0.9290
```

```
0.9290 - val_loss: 1.6089 - val_accuracy: 0.6302
Epoch 12/30
1/96 [...] - ETA: 2s - loss: 0.4408 - accuracy:
0.8571
5/96 [>...] - ETA: 1s - loss: 0.5202 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 0.5544 - accuracy:
0.8905
15/96 [===>...] - ETA: Os - loss: 0.4798 - accuracy:
0.9079
20/96 [====>...] - ETA: Os - loss: 0.4714 - accuracy:
25/96 [=====>...] - ETA: Os - loss: 0.4783 - accuracy:
0.9067
30/96 [======>...] - ETA: Os - loss: 0.4506 - accuracy:
0.9190
34/96 [=======>...] - ETA: Os - loss: 0.4520 - accuracy:
0.9160
39/96 [=======>...] - ETA: Os - loss: 0.4353 - accuracy:
44/96 [========>...] - ETA: Os - loss: 0.4501 - accuracy:
0.9123
0.9155
0.9186
0.9130
0.9112
0.9115
0.9094
0.9091
0.9089
96/96 [============ ] - 1s 15ms/step - loss: 0.4518 - accuracy:
0.9085 - val_loss: 1.8827 - val_accuracy: 0.5901
```

Epoch 13/30

```
1/96 [...] - ETA: 2s - loss: 0.3056 - accuracy:
0.9524
6/96 [>...] - ETA: 1s - loss: 0.3472 - accuracy:
0.9444
11/96 [==>...] - ETA: Os - loss: 0.3773 - accuracy:
0.9307
15/96 [===>...] - ETA: Os - loss: 0.4244 - accuracy:
0.9143
20/96 [====>...] - ETA: Os - loss: 0.4352 - accuracy:
0.9119
25/96 [=====>...] - ETA: Os - loss: 0.4341 - accuracy:
0.9124
30/96 [======>...] - ETA: Os - loss: 0.4283 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 0.4101 - accuracy:
0.9184
41/96 [========>...] - ETA: Os - loss: 0.4163 - accuracy:
0.9164
0.9193
0.9226
0.9227
0.9272
0.9292
0.9307
0.9314
96/96 [============ ] - 1s 14ms/step - loss: 0.3828 - accuracy:
0.9310 - val_loss: 1.6588 - val_accuracy: 0.6678
Epoch 14/30
1/96 [...] - ETA: 1s - loss: 0.2894 - accuracy:
0.9524
```

```
5/96 [>...] - ETA: 1s - loss: 0.3169 - accuracy:
0.9524
9/96 [=>...] - ETA: 1s - loss: 0.3287 - accuracy:
0.9524
13/96 [===>...] - ETA: 1s - loss: 0.3246 - accuracy:
0.9524
18/96 [====>...] - ETA: Os - loss: 0.3456 - accuracy:
0.9418
23/96 [=====>...] - ETA: Os - loss: 0.3625 - accuracy:
0.9379
27/96 [======>...] - ETA: Os - loss: 0.3608 - accuracy:
0.9347
31/96 [======>...] - ETA: Os - loss: 0.3708 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.3691 - accuracy:
41/96 [========>...] - ETA: Os - loss: 0.3683 - accuracy:
0.9292
46/96 [========>...] - ETA: Os - loss: 0.3751 - accuracy:
0.9296
0.9309
0.9303
0.9282
0.9286
0.9295
0.9283
0.9288
96/96 [============== ] - ETA: Os - loss: 0.3815 - accuracy:
0.9290
0.9290 - val_loss: 1.6870 - val_accuracy: 0.6729
Epoch 15/30
1/96 [...] - ETA: 1s - loss: 0.2642 - accuracy:
0.9524
6/96 [>...] - ETA: 1s - loss: 0.4299 - accuracy:
```

```
0.9048
11/96 [==>...] - ETA: Os - loss: 0.3742 - accuracy:
0.9351
16/96 [====>...] - ETA: Os - loss: 0.3433 - accuracy:
0.9435
21/96 [====>...] - ETA: Os - loss: 0.3569 - accuracy:
26/96 [======>...] - ETA: Os - loss: 0.3424 - accuracy:
0.9432
31/96 [======>...] - ETA: Os - loss: 0.3393 - accuracy:
0.9462
36/96 [=======>...] - ETA: Os - loss: 0.3586 - accuracy:
0.9392
41/96 [========>...] - ETA: Os - loss: 0.3542 - accuracy:
0.9397
0.9410
0.9379
0.9385
0.9363
0.9347
0.9361
0.9347
0.9357
0.9365 - val_loss: 1.7403 - val_accuracy: 0.6721
Epoch 16/30
1/96 [...] - ETA: 2s - loss: 0.2323 - accuracy:
1.0000
6/96 [>...] - ETA: Os - loss: 0.2924 - accuracy:
10/96 [==>...] - ETA: Os - loss: 0.2651 - accuracy:
0.9571
```

```
15/96 [===>...] - ETA: Os - loss: 0.2817 - accuracy:
0.9556
20/96 [====>...] - ETA: Os - loss: 0.3100 - accuracy:
0.9476
25/96 [=====>...] - ETA: Os - loss: 0.3052 - accuracy:
0.9524
30/96 [=======>...] - ETA: Os - loss: 0.3091 - accuracy:
0.9492
35/96 [=======>...] - ETA: Os - loss: 0.3029 - accuracy:
0.9510
40/96 [========>...] - ETA: Os - loss: 0.2987 - accuracy:
0.9536
45/96 [========>...] - ETA: Os - loss: 0.2960 - accuracy:
0.9524
0.9524
0.9502
0.9491
0.9493
0.9489
0.9470 - val_loss: 2.2128 - val_accuracy: 0.5773
Epoch 17/30
1/96 [...] - ETA: 1s - loss: 0.2128 - accuracy:
0.9524
6/96 [>...] - ETA: 1s - loss: 0.2713 - accuracy:
0.9683
11/96 [==>...] - ETA: Os - loss: 0.3032 - accuracy:
0.9437
17/96 [====>...] - ETA: Os - loss: 0.3113 - accuracy:
0.9384
21/96 [====>...] - ETA: Os - loss: 0.3096 - accuracy:
0.9365
25/96 [=====>...] - ETA: Os - loss: 0.3109 - accuracy:
```

```
0.9352
30/96 [======>...] - ETA: Os - loss: 0.3253 - accuracy:
0.9286
36/96 [========>...] - ETA: Os - loss: 0.3112 - accuracy:
0.9365
41/96 [=======>...] - ETA: Os - loss: 0.3180 - accuracy:
46/96 [=========>...] - ETA: Os - loss: 0.3208 - accuracy:
0.9379
0.9396
0.9415
0.9439
0.9411
0.9430
0.9451
96/96 [============== ] - ETA: Os - loss: 0.3059 - accuracy:
0.9450
0.9450 - val_loss: 2.1110 - val_accuracy: 0.6336
Epoch 18/30
1/96 [...] - ETA: 2s - loss: 0.1986 - accuracy:
1.0000
6/96 [>...] - ETA: Os - loss: 0.2325 - accuracy:
0.9762
12/96 [==>...] - ETA: Os - loss: 0.2500 - accuracy:
0.9722
17/96 [====>...] - ETA: Os - loss: 0.2406 - accuracy:
0.9804
23/96 [=====>...] - ETA: Os - loss: 0.2435 - accuracy:
28/96 [======>...] - ETA: Os - loss: 0.2432 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 0.2363 - accuracy:
0.9755
```

```
39/96 [========>...] - ETA: Os - loss: 0.2420 - accuracy:
0.9744
44/96 [========>...] - ETA: Os - loss: 0.2518 - accuracy:
0.9699
0.9656
0.9653
0.9635
0.9604
0.9583
0.9591
0.9588
0.9596
96/96 [============= ] - ETA: Os - loss: 0.2680 - accuracy:
0.9590
0.9590 - val_loss: 1.8945 - val_accuracy: 0.6294
Epoch 19/30
1/96 [...] - ETA: 2s - loss: 0.2276 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.2789 - accuracy:
9/96 [=>...] - ETA: 1s - loss: 0.2307 - accuracy:
0.9683
13/96 [===>...] - ETA: 1s - loss: 0.2489 - accuracy:
0.9707
17/96 [====>...] - ETA: 1s - loss: 0.2474 - accuracy:
0.9720
21/96 [====>...] - ETA: 1s - loss: 0.2385 - accuracy:
0.9728
25/96 [=====>...] - ETA: Os - loss: 0.2670 - accuracy:
0.9638
29/96 [=======>...] - ETA: Os - loss: 0.2700 - accuracy:
```

```
0.9606
33/96 [======>...] - ETA: Os - loss: 0.2809 - accuracy:
0.9553
37/96 [========>...] - ETA: Os - loss: 0.2796 - accuracy:
0.9550
40/96 [=======>...] - ETA: Os - loss: 0.2770 - accuracy:
44/96 [========>...] - ETA: Os - loss: 0.2736 - accuracy:
0.9556
0.9554
0.9551
0.9524
0.9487
0.9482
0.9518
0.9534
0.9535 - val_loss: 1.6111 - val_accuracy: 0.6832
Epoch 20/30
1/96 [...] - ETA: 1s - loss: 0.1394 - accuracy:
1.0000
4/96 [>...] - ETA: 1s - loss: 0.1666 - accuracy:
1.0000
9/96 [=>...] - ETA: 1s - loss: 0.2128 - accuracy:
13/96 [===>...] - ETA: 1s - loss: 0.1956 - accuracy:
18/96 [====>...] - ETA: 1s - loss: 0.1971 - accuracy:
0.9841
```

```
22/96 [====>...] - ETA: Os - loss: 0.1947 - accuracy:
0.9827
26/96 [======>...] - ETA: Os - loss: 0.1971 - accuracy:
0.9835
30/96 [======>...] - ETA: Os - loss: 0.2008 - accuracy:
0.9810
34/96 [=======>...] - ETA: Os - loss: 0.1960 - accuracy:
0.9832
38/96 [=======>...] - ETA: Os - loss: 0.2032 - accuracy:
0.9774
42/96 [========>...] - ETA: Os - loss: 0.1986 - accuracy:
0.9796
0.9712
0.9708
0.9696
0.9685
0.9669
0.9654
0.9648
0.9648
0.9640 - val_loss: 1.9677 - val_accuracy: 0.6294
Epoch 21/30
1/96 [...] - ETA: 1s - loss: 0.1517 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.2234 - accuracy:
0.9762
11/96 [==>...] - ETA: 1s - loss: 0.2277 - accuracy:
0.9740
15/96 [===>...] - ETA: Os - loss: 0.2512 - accuracy:
```

```
0.9683
19/96 [====>...] - ETA: Os - loss: 0.2596 - accuracy:
0.9624
24/96 [=====>...] - ETA: Os - loss: 0.2556 - accuracy:
0.9583
28/96 [======>...] - ETA: Os - loss: 0.2436 - accuracy:
0.9609
32/96 [=======>...] - ETA: Os - loss: 0.2457 - accuracy:
0.9598
37/96 [========>...] - ETA: Os - loss: 0.2403 - accuracy:
0.9614
41/96 [========>...] - ETA: Os - loss: 0.2408 - accuracy:
0.9628
45/96 [========>...] - ETA: Os - loss: 0.2351 - accuracy:
0.9638
0.9630
0.9637
0.9622
0.9635
0.9646
0.9644
0.9627
0.9635 - val_loss: 1.9454 - val_accuracy: 0.6892
Epoch 22/30
1/96 [...] - ETA: 2s - loss: 0.1685 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.1914 - accuracy:
11/96 [==>...] - ETA: Os - loss: 0.1728 - accuracy:
16/96 [====>...] - ETA: Os - loss: 0.1709 - accuracy:
0.9821
```

```
21/96 [====>...] - ETA: Os - loss: 0.1729 - accuracy:
0.9819
26/96 [======>...] - ETA: Os - loss: 0.1765 - accuracy:
0.9780
31/96 [======>...] - ETA: Os - loss: 0.1763 - accuracy:
0.9785
36/96 [=======>...] - ETA: 0s - loss: 0.1718 - accuracy:
0.9815
41/96 [========>...] - ETA: Os - loss: 0.1700 - accuracy:
0.9826
0.9831
0.9839
0.9844
0.9839
0.9841
0.9852
0.9855
0.9853
0.9855 - val_loss: 1.7679 - val_accuracy: 0.6328
Epoch 23/30
1/96 [...] - ETA: 2s - loss: 0.1479 - accuracy:
1.0000
6/96 [>...] - ETA: Os - loss: 0.1288 - accuracy:
0.9921
11/96 [==>...] - ETA: Os - loss: 0.1621 - accuracy:
0.9913
16/96 [====>...] - ETA: Os - loss: 0.1520 - accuracy:
0.9911
20/96 [====>...] - ETA: Os - loss: 0.1524 - accuracy:
0.9881
25/96 [=====>...] - ETA: Os - loss: 0.1499 - accuracy:
```

```
0.9886
29/96 [======>...] - ETA: Os - loss: 0.1483 - accuracy:
0.9885
34/96 [=======>...] - ETA: Os - loss: 0.1517 - accuracy:
0.9860
39/96 [=======>...] - ETA: Os - loss: 0.1485 - accuracy:
44/96 [========>...] - ETA: Os - loss: 0.1540 - accuracy:
0.9848
0.9845
0.9844
0.9832
0.9837
0.9829
0.9827
96/96 [============== ] - ETA: Os - loss: 0.1575 - accuracy:
0.9835
0.9835 - val_loss: 1.8603 - val_accuracy: 0.6405
Epoch 24/30
1/96 [...] - ETA: 1s - loss: 0.1958 - accuracy:
0.9048
6/96 [>...] - ETA: Os - loss: 0.2168 - accuracy:
0.9762
12/96 [==>...] - ETA: Os - loss: 0.2285 - accuracy:
0.9524
17/96 [====>...] - ETA: Os - loss: 0.2197 - accuracy:
0.9524
22/96 [====>...] - ETA: Os - loss: 0.2403 - accuracy:
26/96 [======>...] - ETA: Os - loss: 0.2463 - accuracy:
31/96 [======>...] - ETA: Os - loss: 0.2534 - accuracy:
0.9524
```

```
36/96 [=======>...] - ETA: Os - loss: 0.2529 - accuracy:
0.9550
41/96 [========>...] - ETA: Os - loss: 0.2437 - accuracy:
0.9593
0.9556
0.9570
0.9591
0.9616
0.9604
0.9612
0.9612
0.9613
0.9593
0.9584
96/96 [============= ] - 1s 14ms/step - loss: 0.2403 - accuracy:
0.9580 - val_loss: 1.9113 - val_accuracy: 0.6089
Epoch 25/30
1/96 [...] - ETA: 2s - loss: 0.1604 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 0.2523 - accuracy:
0.9444
11/96 [==>...] - ETA: Os - loss: 0.2400 - accuracy:
16/96 [====>...] - ETA: Os - loss: 0.2370 - accuracy:
0.9583
21/96 [====>...] - ETA: Os - loss: 0.2547 - accuracy:
0.9524
26/96 [======>...] - ETA: Os - loss: 0.2421 - accuracy:
0.9579
30/96 [======>...] - ETA: Os - loss: 0.2497 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.2653 - accuracy:
0.9471
41/96 [========>...] - ETA: Os - loss: 0.2696 - accuracy:
```

```
0.9454
0.9463
0.9451
0.9418
0.9433
0.9446
0.9449
0.9426
0.9420
0.9405 - val_loss: 2.3600 - val_accuracy: 0.6405
Epoch 26/30
1/96 [...] - ETA: 2s - loss: 0.3020 - accuracy:
0.9524
5/96 [>...] - ETA: 1s - loss: 0.3410 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 0.2727 - accuracy:
0.9524
15/96 [===>...] - ETA: Os - loss: 0.2728 - accuracy:
0.9556
20/96 [====>...] - ETA: Os - loss: 0.2682 - accuracy:
0.9548
25/96 [=====>...] - ETA: Os - loss: 0.2781 - accuracy:
0.9562
30/96 [======>...] - ETA: Os - loss: 0.2923 - accuracy:
0.9524
35/96 [=======>...] - ETA: Os - loss: 0.2912 - accuracy:
0.9483
40/96 [========>...] - ETA: Os - loss: 0.3045 - accuracy:
46/96 [========>...] - ETA: Os - loss: 0.3106 - accuracy:
0.9390
```

```
0.9420
0.9449
0.9448
0.9453
0.9476
0.9477
0.9465 - val_loss: 2.0110 - val_accuracy: 0.5918
Epoch 1/30
1/96 [...] - ETA: 6:03 - loss: 10.5251 - accuracy:
0.0952
6/96 [>...] - ETA: Os - loss: 11.7066 - accuracy:
0.2381
11/96 [==>...] - ETA: Os - loss: 10.8317 - accuracy:
17/96 [====>...] - ETA: Os - loss: 10.0666 - accuracy:
0.2493
22/96 [====>...] - ETA: Os - loss: 9.7091 - accuracy:
0.2554
28/96 [======>...] - ETA: Os - loss: 9.3193 - accuracy:
0.2517
34/96 [=======>...] - ETA: Os - loss: 8.9919 - accuracy:
39/96 [========>...] - ETA: Os - loss: 8.7854 - accuracy:
45/96 [=========>...] - ETA: Os - loss: 8.5759 - accuracy:
0.2635
0.2724
0.2753
0.2865
```

```
0.2929
0.2978
0.2990
0.3095
0.3193
0.3193 - val_loss: 7.3942 - val_accuracy: 0.1469
Epoch 2/30
1/96 [...] - ETA: 1s - loss: 5.4862 - accuracy:
0.2857
7/96 [=>...] - ETA: Os - loss: 5.3892 - accuracy:
0.4354
12/96 [==>...] - ETA: Os - loss: 5.3717 - accuracy:
0.4444
17/96 [====>...] - ETA: Os - loss: 5.2819 - accuracy:
0.4594
22/96 [====>...] - ETA: Os - loss: 5.2014 - accuracy:
0.4567
27/96 [======>...] - ETA: Os - loss: 5.2383 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 5.2130 - accuracy:
38/96 [=======>...] - ETA: Os - loss: 5.1251 - accuracy:
0.4586
0.4618
0.4684
0.4691
0.4651
0.4664
```

```
0.4722
0.4746
0.4792
0.4793 - val_loss: 4.2107 - val_accuracy: 0.3997
Epoch 3/30
1/96 [...] - ETA: 1s - loss: 3.9606 - accuracy:
0.5238
6/96 [>...] - ETA: 1s - loss: 3.5386 - accuracy:
0.6587
11/96 [==>...] - ETA: Os - loss: 3.4719 - accuracy:
0.6926
16/96 [====>...] - ETA: Os - loss: 3.4340 - accuracy:
0.6756
22/96 [====>...] - ETA: Os - loss: 3.4297 - accuracy:
28/96 [======>...] - ETA: Os - loss: 3.4415 - accuracy:
34/96 [=======>...] - ETA: Os - loss: 3.3891 - accuracy:
0.6471
39/96 [=======>...] - ETA: Os - loss: 3.3797 - accuracy:
0.6410
0.6209
0.6233
0.6252
0.6283
0.6300
0.6337
```

```
0.6337 - val_loss: 3.1264 - val_accuracy: 0.5141
Epoch 4/30
1/96 [...] - ETA: 2s - loss: 2.7605 - accuracy:
0.6190
7/96 [=>...] - ETA: 0s - loss: 2.7136 - accuracy:
0.6395
13/96 [===>...] - ETA: Os - loss: 2.5918 - accuracy:
0.6667
19/96 [====>...] - ETA: Os - loss: 2.5114 - accuracy:
0.6717
24/96 [=====>...] - ETA: Os - loss: 2.5257 - accuracy:
30/96 [======>...] - ETA: Os - loss: 2.4974 - accuracy:
36/96 [======>...] - ETA: Os - loss: 2.4646 - accuracy:
0.6918
41/96 [=======>...] - ETA: Os - loss: 2.4319 - accuracy:
0.6945
0.6925
0.6969
0.6946
0.7089
0.7126
0.7153
96/96 [============== ] - ETA: Os - loss: 2.2479 - accuracy:
0.7171
0.7171 - val_loss: 2.7829 - val_accuracy: 0.4492
Epoch 5/30
1/96 [...] - ETA: 2s - loss: 1.9616 - accuracy:
0.6667
7/96 [=>...] - ETA: Os - loss: 1.9515 - accuracy:
```

```
0.7551
13/96 [===>...] - ETA: Os - loss: 1.8912 - accuracy:
0.7802
18/96 [====>...] - ETA: Os - loss: 1.8782 - accuracy:
0.7593
24/96 [=====>...] - ETA: Os - loss: 1.8830 - accuracy:
30/96 [======>...] - ETA: Os - loss: 1.8864 - accuracy:
0.7524
35/96 [=======>...] - ETA: Os - loss: 1.8611 - accuracy:
0.7578
40/96 [========>...] - ETA: Os - loss: 1.8409 - accuracy:
0.7560
0.7429
0.7463
0.7492
0.7500
0.7485
0.7487
0.7452
0.7441 - val_loss: 2.2219 - val_accuracy: 0.5756
Epoch 6/30
1/96 [...] - ETA: 2s - loss: 1.4207 - accuracy:
0.9048
6/96 [>...] - ETA: Os - loss: 1.5129 - accuracy:
0.7937
11/96 [==>...] - ETA: Os - loss: 1.5186 - accuracy:
15/96 [===>...] - ETA: Os - loss: 1.4500 - accuracy:
0.7841
```

```
21/96 [====>...] - ETA: Os - loss: 1.3841 - accuracy:
0.8027
27/96 [======>...] - ETA: Os - loss: 1.3914 - accuracy:
0.8007
33/96 [=======>...] - ETA: Os - loss: 1.3975 - accuracy:
0.8023
37/96 [=======>...] - ETA: 0s - loss: 1.3836 - accuracy:
0.8044
42/96 [========>...] - ETA: Os - loss: 1.3864 - accuracy:
0.8039
0.8033
0.8020
0.7926
0.7899
0.7870
0.7890
0.7874
0.7881
0.7840
0.7841 - val_loss: 1.8833 - val_accuracy: 0.6234
Epoch 7/30
1/96 [...] - ETA: 1s - loss: 0.9947 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 1.2842 - accuracy:
0.8016
11/96 [==>...] - ETA: Os - loss: 1.2372 - accuracy:
0.8095
16/96 [====>...] - ETA: Os - loss: 1.2277 - accuracy:
21/96 [====>...] - ETA: Os - loss: 1.2061 - accuracy:
0.8367
26/96 [======>...] - ETA: Os - loss: 1.2217 - accuracy:
0.8278
32/96 [=======>...] - ETA: Os - loss: 1.2007 - accuracy:
```

```
0.8333
38/96 [=======>...] - ETA: Os - loss: 1.1868 - accuracy:
0.8358
43/96 [========>...] - ETA: Os - loss: 1.1651 - accuracy:
0.8394
0.8372
0.8392
0.8387
0.8333
0.8282
0.8256
0.8221 - val_loss: 2.2460 - val_accuracy: 0.5816
Epoch 8/30
1/96 [...] - ETA: 1s - loss: 0.9762 - accuracy:
0.8571
7/96 [=>...] - ETA: Os - loss: 1.0258 - accuracy:
0.8503
13/96 [===>...] - ETA: Os - loss: 1.0699 - accuracy:
0.8388
17/96 [====>...] - ETA: Os - loss: 1.0858 - accuracy:
0.8319
23/96 [=====>...] - ETA: Os - loss: 1.0769 - accuracy:
0.8344
29/96 [======>...] - ETA: Os - loss: 1.0885 - accuracy:
0.8276
33/96 [=======>...] - ETA: Os - loss: 1.0822 - accuracy:
0.8268
39/96 [========>...] - ETA: Os - loss: 1.0834 - accuracy:
45/96 [========>...] - ETA: Os - loss: 1.0664 - accuracy:
0.8280
```

```
0.8293
58/96 [=========>...] - ETA: Os - loss: 1.0508 - accuracy:
0.8360
0.8361
0.8350
0.8309
0.8290
0.8276
0.8276 - val_loss: 2.1770 - val_accuracy: 0.5226
Epoch 9/30
1/96 [...] - ETA: 2s - loss: 0.9165 - accuracy:
0.8095
7/96 [=>...] - ETA: Os - loss: 0.8572 - accuracy:
0.8503
12/96 [==>...] - ETA: Os - loss: 0.9102 - accuracy:
0.8413
16/96 [====>...] - ETA: Os - loss: 0.9236 - accuracy:
0.8393
20/96 [====>...] - ETA: Os - loss: 0.9178 - accuracy:
0.8476
25/96 [=====>...] - ETA: Os - loss: 0.9401 - accuracy:
0.8400
30/96 [======>...] - ETA: Os - loss: 0.9194 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 0.9083 - accuracy:
0.8463
39/96 [========>...] - ETA: Os - loss: 0.8995 - accuracy:
0.8486
43/96 [========>...] - ETA: Os - loss: 0.8949 - accuracy:
0.8483
47/96 [========>...] - ETA: Os - loss: 0.9023 - accuracy:
0.8470
0.8497
```

```
0.8563
0.8603
0.8593
0.8592
0.8603
0.8639
0.8638
0.8631 - val_loss: 1.7263 - val_accuracy: 0.6319
Epoch 10/30
1/96 [...] - ETA: 2s - loss: 0.6489 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 0.7636 - accuracy:
0.8476
9/96 [=>...] - ETA: 1s - loss: 0.7455 - accuracy:
0.8571
13/96 [===>...] - ETA: 1s - loss: 0.7156 - accuracy:
0.8755
17/96 [====>...] - ETA: 1s - loss: 0.7197 - accuracy:
0.8739
21/96 [====>...] - ETA: 1s - loss: 0.6991 - accuracy:
0.8844
25/96 [=====>...] - ETA: 1s - loss: 0.7002 - accuracy:
0.8857
30/96 [======>...] - ETA: Os - loss: 0.7030 - accuracy:
0.8873
34/96 [=======>...] - ETA: Os - loss: 0.6980 - accuracy:
0.8908
38/96 [========>...] - ETA: Os - loss: 0.6956 - accuracy:
0.8910
42/96 [========>...] - ETA: Os - loss: 0.6857 - accuracy:
0.8957
46/96 [========>...] - ETA: Os - loss: 0.6908 - accuracy:
0.8933
```

```
0.8933
0.8948
0.8941
0.8900
0.8895
0.8878
0.8852
0.8837
0.8821 - val_loss: 1.8616 - val_accuracy: 0.5448
Epoch 11/30
1/96 [...] - ETA: 1s - loss: 0.6504 - accuracy:
0.9048
7/96 [=>...] - ETA: Os - loss: 0.8638 - accuracy:
0.8435
11/96 [==>...] - ETA: Os - loss: 0.8568 - accuracy:
0.8485
17/96 [====>...] - ETA: Os - loss: 0.8521 - accuracy:
0.8543
22/96 [====>...] - ETA: Os - loss: 0.8306 - accuracy:
0.8571
27/96 [======>...] - ETA: Os - loss: 0.8403 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 0.8606 - accuracy:
0.8543
37/96 [======>...] - ETA: Os - loss: 0.8687 - accuracy:
0.8520
43/96 [========>...] - ETA: Os - loss: 0.8456 - accuracy:
0.8562
0.8527
```

```
0.8563
0.8593
0.8551
0.8552
0.8583
0.8588
0.8598
0.8606 - val_loss: 1.7459 - val_accuracy: 0.6319
Epoch 12/30
1/96 [...] - ETA: 2s - loss: 1.0516 - accuracy:
0.8095
5/96 [>...] - ETA: 1s - loss: 0.6518 - accuracy:
0.9333
10/96 [==>...] - ETA: 1s - loss: 0.6569 - accuracy:
0.9190
16/96 [====>...] - ETA: Os - loss: 0.6485 - accuracy:
0.9107
22/96 [====>...] - ETA: Os - loss: 0.6705 - accuracy:
27/96 [======>...] - ETA: Os - loss: 0.6773 - accuracy:
33/96 [======>...] - ETA: Os - loss: 0.6653 - accuracy:
0.9062
0.9109
0.9102
0.9105
0.9107
0.9118
0.9048
```

```
0.9035
0.9032
0.9038
0.9030 - val_loss: 1.5877 - val_accuracy: 0.6670
Epoch 13/30
1/96 [...] - ETA: 1s - loss: 0.6241 - accuracy:
0.8571
5/96 [>...] - ETA: 1s - loss: 0.6381 - accuracy:
0.9048
11/96 [==>...] - ETA: Os - loss: 0.6136 - accuracy:
0.9091
17/96 [====>...] - ETA: Os - loss: 0.6139 - accuracy:
0.9132
22/96 [====>...] - ETA: Os - loss: 0.6076 - accuracy:
0.9134
27/96 [======>...] - ETA: Os - loss: 0.6131 - accuracy:
0.9118
33/96 [=======>...] - ETA: Os - loss: 0.5999 - accuracy:
0.9206
39/96 [========>...] - ETA: Os - loss: 0.6192 - accuracy:
44/96 [========>...] - ETA: Os - loss: 0.6324 - accuracy:
0.9091
0.9048
0.9082
0.9076
0.9087
0.9075
```

```
0.9063
96/96 [============= ] - 1s 14ms/step - loss: 0.6361 - accuracy:
0.9040 - val_loss: 1.5600 - val_accuracy: 0.6388
Epoch 14/30
1/96 [...] - ETA: 1s - loss: 0.5274 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 0.5573 - accuracy:
0.9365
12/96 [==>...] - ETA: Os - loss: 0.5999 - accuracy:
0.9246
18/96 [====>...] - ETA: Os - loss: 0.5978 - accuracy:
0.9206
24/96 [=====>...] - ETA: Os - loss: 0.6231 - accuracy:
0.9087
29/96 [======>...] - ETA: Os - loss: 0.6092 - accuracy:
0.9097
35/96 [=======>...] - ETA: Os - loss: 0.6096 - accuracy:
0.9102
41/96 [=======>...] - ETA: Os - loss: 0.6090 - accuracy:
0.9094
47/96 [=========>...] - ETA: Os - loss: 0.6124 - accuracy:
0.9088
0.9075
0.9105
0.9131
0.9149
0.9147
0.9168
0.9163
96/96 [============ ] - 1s 12ms/step - loss: 0.5930 - accuracy:
0.9165 - val_loss: 2.0494 - val_accuracy: 0.6516
Epoch 15/30
1/96 [...] - ETA: 2s - loss: 0.3966 - accuracy:
```

```
1.0000
6/96 [>...] - ETA: Os - loss: 0.5330 - accuracy:
0.9286
12/96 [==>...] - ETA: Os - loss: 0.5319 - accuracy:
0.9286
16/96 [====>...] - ETA: Os - loss: 0.5149 - accuracy:
22/96 [====>...] - ETA: Os - loss: 0.5308 - accuracy:
0.9221
27/96 [======>...] - ETA: Os - loss: 0.5296 - accuracy:
0.9242
33/96 [=======>...] - ETA: Os - loss: 0.5361 - accuracy:
0.9250
38/96 [=======>...] - ETA: Os - loss: 0.5390 - accuracy:
0.9236
42/96 [========>...] - ETA: Os - loss: 0.5318 - accuracy:
0.9286
0.9291
0.9254
0.9259
0.9262
0.9259
0.9214
0.9224
0.9225 - val_loss: 1.7809 - val_accuracy: 0.6550
Epoch 16/30
1/96 [...] - ETA: 1s - loss: 0.3687 - accuracy:
1.0000
7/96 [=>...] - ETA: Os - loss: 0.4440 - accuracy:
12/96 [==>...] - ETA: Os - loss: 0.4869 - accuracy:
16/96 [====>...] - ETA: Os - loss: 0.4844 - accuracy:
0.9315
```

```
22/96 [====>...] - ETA: Os - loss: 0.4763 - accuracy:
0.9351
28/96 [======>...] - ETA: Os - loss: 0.4878 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 0.4816 - accuracy:
0.9336
0.9365
44/96 [========>...] - ETA: Os - loss: 0.4833 - accuracy:
0.9372
0.9368
0.9300
0.9253
0.9255
0.9254
0.9230
0.9228
0.9225 - val_loss: 2.4731 - val_accuracy: 0.5004
Epoch 17/30
1/96 [...] - ETA: 1s - loss: 0.6503 - accuracy:
0.8571
6/96 [>...] - ETA: 1s - loss: 0.5605 - accuracy:
12/96 [==>...] - ETA: Os - loss: 0.5667 - accuracy:
0.9127
18/96 [====>...] - ETA: Os - loss: 0.5508 - accuracy:
0.9233
24/96 [=====>...] - ETA: Os - loss: 0.5218 - accuracy:
29/96 [=======>...] - ETA: Os - loss: 0.5148 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 0.5089 - accuracy:
0.9374
41/96 [========>...] - ETA: Os - loss: 0.5106 - accuracy:
```

```
0.9373
0.9358
0.9359
0.9335
0.9345
0.9351
0.9363
0.9361
0.9333
0.9298
0.9300 - val_loss: 2.3456 - val_accuracy: 0.5884
Epoch 1/30
1/96 [...] - ETA: 6:01 - loss: 7.9447 - accuracy:
0.0000e+00
5/96 [>...] - ETA: 1s - loss: 8.9223 - accuracy:
8/96 [=>...] - ETA: 1s - loss: 8.5882 - accuracy:
0.0833
13/96 [===>...] - ETA: 1s - loss: 7.8447 - accuracy:
0.1502
18/96 [====>...] - ETA: 1s - loss: 7.5151 - accuracy:
0.1640
22/96 [====>...] - ETA: Os - loss: 7.2458 - accuracy:
0.1753
27/96 [======>...] - ETA: Os - loss: 7.0384 - accuracy:
0.1869
32/96 [=======>...] - ETA: Os - loss: 6.8221 - accuracy:
0.1979
37/96 [========>...] - ETA: Os - loss: 6.6585 - accuracy:
0.2072
42/96 [========>...] - ETA: Os - loss: 6.4979 - accuracy:
0.2361
```

```
0.2564
0.2693
0.2749
0.2816
0.2889
0.3033
0.3087
0.3151
96/96 [============ ] - 6s 19ms/step - loss: 5.5113 - accuracy:
0.3163 - val_loss: 6.3272 - val_accuracy: 0.1631
Epoch 2/30
1/96 [...] - ETA: 2s - loss: 4.2119 - accuracy:
0.4762
5/96 [>...] - ETA: 1s - loss: 4.1062 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 3.9112 - accuracy:
0.5476
15/96 [===>...] - ETA: Os - loss: 3.8848 - accuracy:
0.5460
18/96 [====>...] - ETA: 1s - loss: 3.8850 - accuracy:
0.5344
23/96 [=====>...] - ETA: Os - loss: 3.8804 - accuracy:
28/96 [======>...] - ETA: Os - loss: 3.8321 - accuracy:
0.5306
32/96 [=======>...] - ETA: Os - loss: 3.7988 - accuracy:
0.5327
36/96 [========>...] - ETA: Os - loss: 3.7813 - accuracy:
41/96 [=======>...] - ETA: Os - loss: 3.7656 - accuracy:
0.5250
0.5373
```

```
0.5394
0.5485
0.5537
0.5518
0.5532
0.5551
0.5626
0.5641
96/96 [============= ] - ETA: Os - loss: 3.4173 - accuracy:
0.5687
0.5687 - val_loss: 3.5698 - val_accuracy: 0.4108
Epoch 3/30
1/96 [...] - ETA: 1s - loss: 2.5467 - accuracy:
0.7143
7/96 [=>...] - ETA: Os - loss: 2.5940 - accuracy:
12/96 [==>...] - ETA: Os - loss: 2.7467 - accuracy:
0.6270
17/96 [====>...] - ETA: Os - loss: 2.6790 - accuracy:
0.6695
21/96 [====>...] - ETA: Os - loss: 2.6527 - accuracy:
0.6780
26/96 [======>...] - ETA: Os - loss: 2.6482 - accuracy:
0.6722
32/96 [=======>...] - ETA: Os - loss: 2.6308 - accuracy:
0.6801
35/96 [=======>...] - ETA: Os - loss: 2.6184 - accuracy:
0.6762
40/96 [======>...] - ETA: Os - loss: 2.5924 - accuracy:
45/96 [========>...] - ETA: Os - loss: 2.5730 - accuracy:
0.6848
```

```
0.6814
0.6813
0.6837
0.6832
0.6863
0.6932
0.6927 - val_loss: 2.9573 - val_accuracy: 0.5013
Epoch 4/30
1/96 [...] - ETA: 2s - loss: 2.2706 - accuracy:
0.7143
5/96 [>...] - ETA: 1s - loss: 1.9976 - accuracy:
0.7429
9/96 [=>...] - ETA: 1s - loss: 2.0010 - accuracy:
12/96 [==>...] - ETA: 1s - loss: 2.0013 - accuracy:
0.7500
16/96 [====>...] - ETA: 1s - loss: 1.9375 - accuracy:
0.7738
20/96 [====>...] - ETA: 1s - loss: 1.9375 - accuracy:
0.7690
24/96 [=====>...] - ETA: 1s - loss: 1.9141 - accuracy:
28/96 [======>...] - ETA: Os - loss: 1.9106 - accuracy:
32/96 [=======>...] - ETA: Os - loss: 1.8820 - accuracy:
0.7783
36/96 [=======>...] - ETA: Os - loss: 1.8483 - accuracy:
0.7870
40/96 [========>...] - ETA: Os - loss: 1.8500 - accuracy:
0.7833
44/96 [========>...] - ETA: Os - loss: 1.8377 - accuracy:
0.7857
```

```
0.7833
0.7898
0.7874
0.7917
0.7878
0.7917
0.7923
0.7925
0.7911
96/96 [============== ] - ETA: Os - loss: 1.7498 - accuracy:
0.7876
0.7876 - val_loss: 2.6317 - val_accuracy: 0.4586
Epoch 5/30
1/96 [...] - ETA: 2s - loss: 1.3712 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 1.4687 - accuracy:
0.7905
8/96 [=>...] - ETA: 1s - loss: 1.4529 - accuracy:
0.8214
12/96 [==>...] - ETA: 1s - loss: 1.4092 - accuracy:
0.8452
16/96 [====>...] - ETA: 1s - loss: 1.4370 - accuracy:
0.8363
20/96 [====>...] - ETA: 1s - loss: 1.4444 - accuracy:
0.8310
23/96 [=====>...] - ETA: 1s - loss: 1.4752 - accuracy:
27/96 [======>...] - ETA: 1s - loss: 1.4666 - accuracy:
31/96 [======>...] - ETA: Os - loss: 1.4505 - accuracy:
0.8295
```

```
35/96 [=======>...] - ETA: Os - loss: 1.4794 - accuracy:
0.8231
39/96 [========>...] - ETA: Os - loss: 1.4701 - accuracy:
0.8184
0.8178
0.8133
0.8166
58/96 [=========>...] - ETA: Os - loss: 1.4659 - accuracy:
0.8038
0.8040
0.8042
0.8032
0.8065
0.8072
0.8075
0.8101 - val_loss: 2.3274 - val_accuracy: 0.5013
Epoch 6/30
1/96 [...] - ETA: 2s - loss: 1.1778 - accuracy:
0.8095
5/96 [>...] - ETA: 1s - loss: 1.1336 - accuracy:
0.8762
10/96 [==>...] - ETA: 1s - loss: 1.1150 - accuracy:
0.8762
15/96 [===>...] - ETA: Os - loss: 1.1287 - accuracy:
0.8667
19/96 [====>...] - ETA: Os - loss: 1.1240 - accuracy:
```

```
0.8697
23/96 [=====>...] - ETA: Os - loss: 1.1281 - accuracy:
0.8592
28/96 [======>...] - ETA: 0s - loss: 1.1029 - accuracy:
0.8673
33/96 [=======>...] - ETA: Os - loss: 1.0842 - accuracy:
38/96 [=======>...] - ETA: Os - loss: 1.0856 - accuracy:
0.8709
42/96 [========>...] - ETA: Os - loss: 1.0800 - accuracy:
0.8734
0.8805
0.8748
0.8749
0.8695
0.8677
0.8676 - val loss: 2.5110 - val accuracy: 0.5295
Epoch 7/30
1/96 [...] - ETA: 2s - loss: 1.0438 - accuracy:
0.7619
6/96 [>...] - ETA: 1s - loss: 0.8714 - accuracy:
0.8810
10/96 [==>...] - ETA: 1s - loss: 0.8848 - accuracy:
0.8905
14/96 [===>...] - ETA: 1s - loss: 0.8918 - accuracy:
19/96 [====>...] - ETA: Os - loss: 0.9016 - accuracy:
0.8897
```

```
24/96 [=====>...] - ETA: Os - loss: 0.8716 - accuracy:
0.8948
29/96 [======>...] - ETA: Os - loss: 0.8580 - accuracy:
34/96 [=======>...] - ETA: Os - loss: 0.8584 - accuracy:
0.9006
39/96 [========>...] - ETA: Os - loss: 0.8557 - accuracy:
0.9011
44/96 [========>...] - ETA: Os - loss: 0.8460 - accuracy:
0.9069
0.9038
58/96 [=========>...] - ETA: Os - loss: 0.8519 - accuracy:
0.9023
0.9010
0.8964
0.8924
0.8852
0.8847
0.8826
0.8807
0.8821 - val_loss: 1.6980 - val_accuracy: 0.6541
Epoch 8/30
1/96 [...] - ETA: 2s - loss: 0.8236 - accuracy:
7/96 [=>...] - ETA: Os - loss: 0.6950 - accuracy:
0.9048
12/96 [==>...] - ETA: Os - loss: 0.6881 - accuracy:
0.9206
18/96 [====>...] - ETA: Os - loss: 0.7008 - accuracy:
23/96 [=====>...] - ETA: Os - loss: 0.7462 - accuracy:
29/96 [======>...] - ETA: Os - loss: 0.7506 - accuracy:
0.8949
35/96 [=======>...] - ETA: Os - loss: 0.7453 - accuracy:
```

```
0.8980
39/96 [========>...] - ETA: Os - loss: 0.7631 - accuracy:
0.8877
44/96 [========>...] - ETA: Os - loss: 0.7570 - accuracy:
0.8896
0.8944
0.8916
0.8893
0.8909
0.8878
0.8837
96/96 [============ ] - 1s 14ms/step - loss: 0.7548 - accuracy:
0.8841 - val_loss: 2.0164 - val_accuracy: 0.5551
Epoch 9/30
1/96 [...] - ETA: 2s - loss: 0.5990 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 0.5966 - accuracy:
0.9206
11/96 [==>...] - ETA: Os - loss: 0.6385 - accuracy:
0.9091
16/96 [====>...] - ETA: Os - loss: 0.6828 - accuracy:
0.8899
21/96 [====>...] - ETA: Os - loss: 0.6869 - accuracy:
0.8934
26/96 [======>...] - ETA: Os - loss: 0.6735 - accuracy:
0.9011
31/96 [=======>...] - ETA: Os - loss: 0.6755 - accuracy:
0.9002
36/96 [======>...] - ETA: Os - loss: 0.6624 - accuracy:
40/96 [========>...] - ETA: Os - loss: 0.6551 - accuracy:
0.9036
```

```
0.8975
0.8945
0.8988
0.8985
0.8990
0.9000
0.9037
0.9053
96/96 [============== ] - ETA: Os - loss: 0.6352 - accuracy:
0.9035
0.9035 - val_loss: 1.7006 - val_accuracy: 0.6371
Epoch 10/30
1/96 [...] - ETA: 2s - loss: 0.7130 - accuracy:
0.8571
5/96 [>...] - ETA: 1s - loss: 0.5601 - accuracy:
0.9143
10/96 [==>...] - ETA: 1s - loss: 0.6229 - accuracy:
0.9000
15/96 [===>...] - ETA: Os - loss: 0.5980 - accuracy:
0.9079
19/96 [====>...] - ETA: Os - loss: 0.5697 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 0.5511 - accuracy:
0.9206
29/96 [======>...] - ETA: Os - loss: 0.5451 - accuracy:
0.9245
34/96 [=======>...] - ETA: Os - loss: 0.5407 - accuracy:
39/96 [========>...] - ETA: Os - loss: 0.5378 - accuracy:
0.9297
```

```
0.9286
0.9264
0.9262
0.9275
0.9252
0.9218
0.9187
0.9180 - val_loss: 1.4621 - val_accuracy: 0.6234
Epoch 11/30
1/96 [...] - ETA: 1s - loss: 0.5418 - accuracy:
0.9048
6/96 [>...] - ETA: Os - loss: 0.6365 - accuracy:
0.8730
12/96 [==>...] - ETA: Os - loss: 0.5368 - accuracy:
0.9087
17/96 [====>...] - ETA: Os - loss: 0.5140 - accuracy:
23/96 [=====>...] - ETA: Os - loss: 0.5337 - accuracy:
29/96 [======>...] - ETA: Os - loss: 0.5378 - accuracy:
0.9130
35/96 [=======>...] - ETA: Os - loss: 0.5385 - accuracy:
0.9170
0.9214
46/96 [========>...] - ETA: Os - loss: 0.5159 - accuracy:
0.9244
0.9194
0.9184
0.9199
```

```
0.9209
0.9198
0.9135
0.9123
0.9130
0.9125 - val_loss: 1.5031 - val_accuracy: 0.7122
Epoch 12/30
1/96 [...] - ETA: 2s - loss: 0.3801 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 0.4090 - accuracy:
0.9603
11/96 [==>...] - ETA: Os - loss: 0.4171 - accuracy:
0.9481
16/96 [====>...] - ETA: Os - loss: 0.4400 - accuracy:
0.9435
21/96 [====>...] - ETA: Os - loss: 0.4351 - accuracy:
0.9478
26/96 [======>...] - ETA: Os - loss: 0.4253 - accuracy:
0.9505
31/96 [======>...] - ETA: Os - loss: 0.4557 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.4511 - accuracy:
0.9405
41/96 [========>...] - ETA: Os - loss: 0.4436 - accuracy:
0.9431
46/96 [========>...] - ETA: Os - loss: 0.4333 - accuracy:
0.9451
0.9485
0.9475
0.9439
```

```
0.9433
0.9417
0.9401
0.9410 - val_loss: 1.1943 - val_accuracy: 0.7165
Epoch 13/30
1/96 [...] - ETA: 2s - loss: 0.3584 - accuracy:
0.9524
5/96 [>...] - ETA: 1s - loss: 0.3966 - accuracy:
0.9333
9/96 [=>...] - ETA: 1s - loss: 0.3798 - accuracy:
0.9365
13/96 [===>...] - ETA: 1s - loss: 0.3502 - accuracy:
0.9487
17/96 [====>...] - ETA: 1s - loss: 0.3777 - accuracy:
0.9412
21/96 [====>...] - ETA: 1s - loss: 0.3707 - accuracy:
0.9456
24/96 [=====>...] - ETA: 1s - loss: 0.3738 - accuracy:
0.9425
28/96 [======>...] - ETA: Os - loss: 0.3678 - accuracy:
0.9422
32/96 [=======>...] - ETA: Os - loss: 0.3634 - accuracy:
0.9420
36/96 [=======>...] - ETA: Os - loss: 0.3717 - accuracy:
0.9378
40/96 [========>...] - ETA: Os - loss: 0.3737 - accuracy:
0.9345
43/96 [========>...] - ETA: Os - loss: 0.3754 - accuracy:
0.9347
0.9331
0.9309
0.9333
0.9322
0.9358
0.9349
```

```
0.9337
0.9332
0.9336
0.9333
0.9341
0.9345 - val_loss: 1.3433 - val_accuracy: 0.6943
Epoch 14/30
1/96 [...] - ETA: 1s - loss: 0.2514 - accuracy:
1.0000
4/96 [>...] - ETA: 1s - loss: 0.2860 - accuracy:
0.9881
7/96 [=>...] - ETA: 1s - loss: 0.2984 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 0.3375 - accuracy:
0.9619
14/96 [===>...] - ETA: 1s - loss: 0.3655 - accuracy:
0.9456
18/96 [====>...] - ETA: 1s - loss: 0.3584 - accuracy:
22/96 [====>...] - ETA: 1s - loss: 0.3574 - accuracy:
0.9524
26/96 [======>...] - ETA: 1s - loss: 0.3544 - accuracy:
0.9487
30/96 [=======>...] - ETA: Os - loss: 0.3579 - accuracy:
0.9460
34/96 [=======>...] - ETA: Os - loss: 0.3575 - accuracy:
39/96 [========>...] - ETA: Os - loss: 0.3615 - accuracy:
43/96 [========>...] - ETA: Os - loss: 0.3695 - accuracy:
0.9435
0.9431
0.9432
```

```
0.9430
0.9443
0.9462
0.9463
0.9468
0.9443
0.9453
0.9445
0.9445 - val_loss: 1.4663 - val_accuracy: 0.7190
Epoch 15/30
1/96 [...] - ETA: 2s - loss: 0.2233 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.2363 - accuracy:
0.9841
11/96 [==>...] - ETA: Os - loss: 0.2737 - accuracy:
0.9654
16/96 [====>...] - ETA: Os - loss: 0.2855 - accuracy:
0.9673
21/96 [====>...] - ETA: Os - loss: 0.3039 - accuracy:
0.9569
25/96 [=====>...] - ETA: Os - loss: 0.3353 - accuracy:
0.9543
30/96 [=======>...] - ETA: Os - loss: 0.3325 - accuracy:
0.9540
35/96 [=======>...] - ETA: Os - loss: 0.3379 - accuracy:
40/96 [=======>...] - ETA: Os - loss: 0.3362 - accuracy:
0.9524
```

```
0.9552
0.9541
0.9556
0.9546
0.9545
0.9550
0.9534
0.9529
0.9530 - val_loss: 1.3207 - val_accuracy: 0.7045
Epoch 16/30
1/96 [...] - ETA: 2s - loss: 0.2685 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 0.2605 - accuracy:
0.9683
11/96 [==>...] - ETA: Os - loss: 0.2792 - accuracy:
0.9567
16/96 [====>...] - ETA: Os - loss: 0.2660 - accuracy:
0.9613
21/96 [====>...] - ETA: Os - loss: 0.2719 - accuracy:
0.9569
26/96 [======>...] - ETA: Os - loss: 0.2740 - accuracy:
30/96 [======>...] - ETA: Os - loss: 0.2676 - accuracy:
0.9635
35/96 [=======>...] - ETA: Os - loss: 0.2670 - accuracy:
0.9646
40/96 [========>...] - ETA: Os - loss: 0.2592 - accuracy:
0.9667
45/96 [========>...] - ETA: Os - loss: 0.2615 - accuracy:
0.9672
0.9686
```

```
0.9688
0.9698
0.9692
0.9687
0.9702
0.9675
0.9681
0.9670 - val_loss: 1.3230 - val_accuracy: 0.7114
Epoch 17/30
1/96 [...] - ETA: 2s - loss: 0.1653 - accuracy:
1.0000
4/96 [>...] - ETA: 1s - loss: 0.1973 - accuracy:
0.9881
8/96 [=>...] - ETA: 1s - loss: 0.1964 - accuracy:
0.9881
13/96 [===>...] - ETA: 1s - loss: 0.2259 - accuracy:
18/96 [====>...] - ETA: Os - loss: 0.2432 - accuracy:
0.9735
24/96 [=====>...] - ETA: Os - loss: 0.2345 - accuracy:
0.9782
28/96 [======>...] - ETA: Os - loss: 0.2256 - accuracy:
0.9813
33/96 [=======>...] - ETA: Os - loss: 0.2223 - accuracy:
0.9812
38/96 [========>...] - ETA: Os - loss: 0.2204 - accuracy:
0.9787
43/96 [========>...] - ETA: Os - loss: 0.2207 - accuracy:
0.9767
0.9724
```

```
0.9713
0.9693
0.9687
0.9686
0.9667
0.9679
0.9680 - val_loss: 1.2400 - val_accuracy: 0.7276
Epoch 18/30
1/96 [...] - ETA: 2s - loss: 0.1600 - accuracy:
1.0000
6/96 [>...] - ETA: Os - loss: 0.2291 - accuracy:
0.9603
11/96 [==>...] - ETA: Os - loss: 0.2306 - accuracy:
0.9654
16/96 [====>...] - ETA: Os - loss: 0.2244 - accuracy:
0.9732
21/96 [====>...] - ETA: Os - loss: 0.2276 - accuracy:
0.9705
26/96 [======>...] - ETA: Os - loss: 0.2489 - accuracy:
0.9689
32/96 [=======>...] - ETA: Os - loss: 0.2492 - accuracy:
0.9717
36/96 [======>...] - ETA: Os - loss: 0.2501 - accuracy:
0.9735
41/96 [=======>...] - ETA: Os - loss: 0.2480 - accuracy:
47/96 [=========>...] - ETA: Os - loss: 0.2467 - accuracy:
0.9726
0.9721
0.9708
0.9678
```

```
0.9663
0.9643
0.9647
0.9651
0.9665
96/96 [============= ] - ETA: Os - loss: 0.2464 - accuracy:
0.9660
0.9660 - val_loss: 1.3544 - val_accuracy: 0.7259
Epoch 19/30
1/96 [...] - ETA: 2s - loss: 0.1701 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.2727 - accuracy:
0.9810
10/96 [==>...] - ETA: Os - loss: 0.3067 - accuracy:
0.9667
15/96 [===>...] - ETA: Os - loss: 0.2707 - accuracy:
0.9746
20/96 [====>...] - ETA: Os - loss: 0.2646 - accuracy:
0.9714
25/96 [=====>...] - ETA: Os - loss: 0.2493 - accuracy:
0.9733
30/96 [======>...] - ETA: Os - loss: 0.2724 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 0.2662 - accuracy:
0.9673
40/96 [=======>...] - ETA: Os - loss: 0.2741 - accuracy:
0.9595
0.9619
0.9590
0.9584
0.9540
0.9538
0.9511
```

```
0.9488
0.9445
96/96 [============= ] - ETA: Os - loss: 0.3280 - accuracy:
0.9445
0.9445 - val_loss: 1.5257 - val_accuracy: 0.6362
Epoch 20/30
1/96 [...] - ETA: 2s - loss: 0.4493 - accuracy:
5/96 [>...] - ETA: 1s - loss: 0.3820 - accuracy:
0.9238
10/96 [==>...] - ETA: 1s - loss: 0.3512 - accuracy:
0.9429
15/96 [===>...] - ETA: Os - loss: 0.3541 - accuracy:
0.9429
20/96 [====>...] - ETA: Os - loss: 0.3424 - accuracy:
0.9476
25/96 [=====>...] - ETA: Os - loss: 0.3314 - accuracy:
0.9505
29/96 [======>...] - ETA: Os - loss: 0.3242 - accuracy:
0.9507
34/96 [=======>...] - ETA: Os - loss: 0.3169 - accuracy:
39/96 [=======>...] - ETA: Os - loss: 0.3052 - accuracy:
0.9585
0.9578
0.9602
0.9619
0.9584
0.9570
```

```
0.9557
0.9555
0.9555 - val_loss: 1.4269 - val_accuracy: 0.6422
Epoch 21/30
1/96 [...] - ETA: 2s - loss: 0.2096 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 0.3463 - accuracy:
0.9206
12/96 [==>...] - ETA: Os - loss: 0.3951 - accuracy:
17/96 [====>...] - ETA: Os - loss: 0.4042 - accuracy:
0.9244
21/96 [====>...] - ETA: Os - loss: 0.4019 - accuracy:
0.9297
26/96 [======>...] - ETA: Os - loss: 0.3763 - accuracy:
0.9359
31/96 [======>...] - ETA: Os - loss: 0.3765 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.3611 - accuracy:
0.9444
41/96 [========>...] - ETA: Os - loss: 0.3618 - accuracy:
0.9431
0.9422
0.9464
0.9490
0.9485
0.9464
0.9462
96/96 [============= ] - 1s 14ms/step - loss: 0.3322 - accuracy:
0.9465 - val_loss: 1.8403 - val_accuracy: 0.7028
```

Epoch 22/30

```
1/96 [...] - ETA: 2s - loss: 0.1644 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.2640 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 0.2649 - accuracy:
0.9619
14/96 [===>...] - ETA: 1s - loss: 0.2567 - accuracy:
0.9626
18/96 [====>...] - ETA: 1s - loss: 0.2802 - accuracy:
0.9603
22/96 [====>...] - ETA: Os - loss: 0.2749 - accuracy:
0.9632
26/96 [======>...] - ETA: Os - loss: 0.2635 - accuracy:
0.9670
31/96 [======>...] - ETA: Os - loss: 0.2725 - accuracy:
0.9647
35/96 [=======>...] - ETA: Os - loss: 0.2681 - accuracy:
0.9646
39/96 [========>...] - ETA: Os - loss: 0.2607 - accuracy:
0.9670
43/96 [========>...] - ETA: Os - loss: 0.2583 - accuracy:
0.9679
0.9696
0.9714
0.9701
0.9713
0.9722
0.9737
0.9747
0.9754
0.9754
```

```
0.9764
0.9760 - val_loss: 1.1015 - val_accuracy: 0.7284
Epoch 23/30
1/96 [...] - ETA: 2s - loss: 0.3549 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 0.2179 - accuracy:
0.9714
9/96 [=>...] - ETA: 1s - loss: 0.2025 - accuracy:
0.9788
13/96 [===>...] - ETA: 1s - loss: 0.2045 - accuracy:
0.9744
17/96 [====>...] - ETA: 1s - loss: 0.1995 - accuracy:
0.9776
22/96 [====>...] - ETA: Os - loss: 0.1919 - accuracy:
0.9805
26/96 [======>...] - ETA: Os - loss: 0.2006 - accuracy:
0.9780
30/96 [======>...] - ETA: Os - loss: 0.2033 - accuracy:
0.9794
34/96 [=======>...] - ETA: Os - loss: 0.2158 - accuracy:
0.9790
38/96 [=======>...] - ETA: Os - loss: 0.2130 - accuracy:
0.9787
42/96 [========>...] - ETA: Os - loss: 0.2144 - accuracy:
0.9785
0.9797
0.9787
0.9800
0.9798
0.9803
0.9807
0.9808
```

```
0.9806
0.9820
0.9820 - val_loss: 1.2565 - val_accuracy: 0.7327
Epoch 24/30
1/96 [...] - ETA: 2s - loss: 0.3816 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 0.1981 - accuracy:
0.9810
9/96 [=>...] - ETA: 1s - loss: 0.1950 - accuracy:
0.9841
13/96 [===>...] - ETA: 1s - loss: 0.1936 - accuracy:
0.9853
17/96 [====>...] - ETA: 1s - loss: 0.1919 - accuracy:
0.9832
20/96 [====>...] - ETA: 1s - loss: 0.1895 - accuracy:
24/96 [=====>...] - ETA: 1s - loss: 0.1819 - accuracy:
0.9861
28/96 [======>...] - ETA: Os - loss: 0.1799 - accuracy:
0.9864
33/96 [=======>...] - ETA: Os - loss: 0.1811 - accuracy:
37/96 [========>...] - ETA: Os - loss: 0.1799 - accuracy:
0.9846
42/96 [========>...] - ETA: Os - loss: 0.1855 - accuracy:
0.9807
47/96 [========>...] - ETA: Os - loss: 0.1812 - accuracy:
0.9818
0.9787
0.9789
0.9751
0.9735
0.9727
```

```
0.9732
0.9709
0.9720
0.9704
0.9705 - val_loss: 1.6266 - val_accuracy: 0.7336
Epoch 25/30
1/96 [...] - ETA: 2s - loss: 0.5192 - accuracy:
0.8571
6/96 [>...] - ETA: Os - loss: 0.2324 - accuracy:
0.9683
12/96 [==>...] - ETA: Os - loss: 0.2433 - accuracy:
0.9643
17/96 [====>...] - ETA: Os - loss: 0.2249 - accuracy:
0.9664
21/96 [====>...] - ETA: Os - loss: 0.2398 - accuracy:
0.9546
26/96 [======>...] - ETA: Os - loss: 0.2463 - accuracy:
0.9542
31/96 [======>...] - ETA: Os - loss: 0.2429 - accuracy:
0.9570
36/96 [=======>...] - ETA: Os - loss: 0.2586 - accuracy:
0.9563
40/96 [========>...] - ETA: Os - loss: 0.2610 - accuracy:
0.9561
0.9574
0.9555
0.9566
0.9569
0.9573
0.9578
```

```
0.9581
0.9585 - val_loss: 2.0961 - val_accuracy: 0.5927
Epoch 26/30
1/96 [...] - ETA: 1s - loss: 0.1939 - accuracy:
7/96 [=>...] - ETA: Os - loss: 0.2630 - accuracy:
0.9660
11/96 [==>...] - ETA: Os - loss: 0.2604 - accuracy:
0.9654
16/96 [====>...] - ETA: Os - loss: 0.2639 - accuracy:
0.9613
21/96 [====>...] - ETA: Os - loss: 0.2491 - accuracy:
0.9660
26/96 [======>...] - ETA: Os - loss: 0.2426 - accuracy:
0.9670
30/96 [======>...] - ETA: Os - loss: 0.2346 - accuracy:
0.9698
35/96 [=======>...] - ETA: Os - loss: 0.2334 - accuracy:
40/96 [=======>...] - ETA: Os - loss: 0.2243 - accuracy:
0.9738
0.9714
0.9714
0.9701
0.9663
0.9683
0.9668
0.9643
0.9625 - val_loss: 1.3552 - val_accuracy: 0.7096
```

```
Epoch 27/30
```

```
1/96 [...] - ETA: 2s - loss: 0.1662 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.1922 - accuracy:
0.9810
11/96 [==>...] - ETA: Os - loss: 0.2555 - accuracy:
0.9654
16/96 [====>...] - ETA: Os - loss: 0.2547 - accuracy:
0.9554
20/96 [====>...] - ETA: Os - loss: 0.2674 - accuracy:
0.9548
24/96 [=====>...] - ETA: Os - loss: 0.2668 - accuracy:
30/96 [======>...] - ETA: Os - loss: 0.2572 - accuracy:
0.9603
35/96 [======>...] - ETA: Os - loss: 0.2468 - accuracy:
0.9646
39/96 [=======>...] - ETA: Os - loss: 0.2463 - accuracy:
0.9634
44/96 [========>...] - ETA: Os - loss: 0.2409 - accuracy:
0.9665
0.9689
0.9665
0.9647
0.9615
0.9628
0.9634
0.9633
0.9648
0.9658
0.9660 - val_loss: 1.3447 - val_accuracy: 0.7327
Epoch 28/30
1/96 [...] - ETA: 1s - loss: 0.1503 - accuracy:
```

```
1.0000
6/96 [>...] - ETA: 1s - loss: 0.1686 - accuracy:
0.9921
11/96 [==>...] - ETA: Os - loss: 0.1942 - accuracy:
0.9827
16/96 [====>...] - ETA: Os - loss: 0.2051 - accuracy:
21/96 [====>...] - ETA: Os - loss: 0.1954 - accuracy:
0.9841
26/96 [======>...] - ETA: Os - loss: 0.1974 - accuracy:
0.9817
31/96 [======>...] - ETA: Os - loss: 0.2036 - accuracy:
0.9785
36/96 [=======>...] - ETA: Os - loss: 0.2138 - accuracy:
0.9775
41/96 [========>...] - ETA: Os - loss: 0.2177 - accuracy:
0.9756
0.9757
0.9734
0.9683
0.9630
0.9624
0.9630
96/96 [============== ] - ETA: Os - loss: 0.2692 - accuracy:
0.9595
96/96 [============ ] - 1s 14ms/step - loss: 0.2692 - accuracy:
0.9595 - val_loss: 1.6120 - val_accuracy: 0.5884
Epoch 29/30
1/96 [...] - ETA: 1s - loss: 0.3174 - accuracy:
6/96 [>...] - ETA: Os - loss: 0.2630 - accuracy:
0.9603
```

```
12/96 [==>...] - ETA: Os - loss: 0.3126 - accuracy:
0.9365
17/96 [====>...] - ETA: Os - loss: 0.2858 - accuracy:
0.9468
22/96 [====>...] - ETA: Os - loss: 0.2844 - accuracy:
0.9524
27/96 [======>...] - ETA: Os - loss: 0.2772 - accuracy:
0.9559
32/96 [=======>...] - ETA: Os - loss: 0.2738 - accuracy:
0.9568
37/96 [======>...] - ETA: Os - loss: 0.2606 - accuracy:
0.9614
42/96 [========>...] - ETA: Os - loss: 0.2614 - accuracy:
0.9585
0.9599
0.9609
0.9610
0.9588
0.9590
0.9604
0.9635
0.9640
0.9645 - val_loss: 1.7201 - val_accuracy: 0.7096
Epoch 1/30
1/96 [...] - ETA: 8:50 - loss: 8.1223 - accuracy:
0.0476
5/96 [>...] - ETA: 1s - loss: 7.8534 - accuracy:
0.1619
9/96 [=>...] - ETA: 1s - loss: 7.5828 - accuracy:
0.1693
12/96 [==>...] - ETA: 1s - loss: 7.4222 - accuracy:
```

```
0.1706
16/96 [====>...] - ETA: 1s - loss: 7.1212 - accuracy:
0.1994
20/96 [====>...] - ETA: 1s - loss: 7.0099 - accuracy:
0.2071
24/96 [=====>...] - ETA: 1s - loss: 6.8610 - accuracy:
28/96 [======>...] - ETA: 1s - loss: 6.7307 - accuracy:
0.2228
32/96 [=======>...] - ETA: Os - loss: 6.5968 - accuracy:
0.2336
36/96 [=======>...] - ETA: Os - loss: 6.5452 - accuracy:
0.2328
40/96 [========>...] - ETA: Os - loss: 6.4345 - accuracy:
44/96 [========>...] - ETA: Os - loss: 6.4067 - accuracy:
0.2316
0.2312
0.2335
0.2459
0.2489
0.2544
0.2600
0.2625
0.2729
0.2835
0.2992
96/96 [============== ] - ETA: Os - loss: 5.4504 - accuracy:
0.2994
0.2994 - val_loss: 6.0736 - val_accuracy: 0.1349
```

Epoch 2/30

```
1/96 [...] - ETA: 1s - loss: 3.5432 - accuracy:
0.4762
5/96 [>...] - ETA: 1s - loss: 3.7289 - accuracy:
0.4952
9/96 [=>...] - ETA: 1s - loss: 3.8388 - accuracy:
0.4709
14/96 [===>...] - ETA: 1s - loss: 3.7764 - accuracy:
0.4932
18/96 [====>...] - ETA: 1s - loss: 3.7916 - accuracy:
0.4841
22/96 [====>...] - ETA: Os - loss: 3.7538 - accuracy:
0.4892
26/96 [======>...] - ETA: Os - loss: 3.7128 - accuracy:
30/96 [======>...] - ETA: Os - loss: 3.7037 - accuracy:
0.5032
34/96 [======>...] - ETA: Os - loss: 3.6653 - accuracy:
0.5070
38/96 [========>...] - ETA: Os - loss: 3.6312 - accuracy:
0.5138
42/96 [========>...] - ETA: Os - loss: 3.5718 - accuracy:
46/96 [=========>...] - ETA: Os - loss: 3.5266 - accuracy:
0.5321
0.5331
0.5394
0.5457
0.5529
0.5587
0.5625
0.5673
0.5709
96/96 [============= ] - 2s 17ms/step - loss: 3.1588 - accuracy:
0.5817 - val_loss: 3.0047 - val_accuracy: 0.5124
```

```
Epoch 3/30
```

```
1/96 [...] - ETA: 2s - loss: 2.3378 - accuracy:
0.7143
4/96 [>...] - ETA: 1s - loss: 2.1792 - accuracy:
8/96 [=>...] - ETA: 1s - loss: 2.2781 - accuracy:
0.7500
12/96 [==>...] - ETA: 1s - loss: 2.3004 - accuracy:
0.7341
16/96 [====>...] - ETA: 1s - loss: 2.2660 - accuracy:
0.7500
20/96 [====>...] - ETA: 1s - loss: 2.2608 - accuracy:
0.7476
25/96 [=====>...] - ETA: Os - loss: 2.2563 - accuracy:
0.7429
29/96 [======>...] - ETA: Os - loss: 2.2417 - accuracy:
0.7455
34/96 [=======>...] - ETA: Os - loss: 2.2356 - accuracy:
0.7437
39/96 [========>...] - ETA: Os - loss: 2.2143 - accuracy:
0.7409
0.7477
0.7498
0.7500
0.7476
0.7455
0.7368
0.7325
0.7313
0.7305
0.7314
```

```
0.7348
0.7346 - val_loss: 3.3306 - val_accuracy: 0.3040
Epoch 4/30
1/96 [...] - ETA: 1s - loss: 1.5318 - accuracy:
0.8571
6/96 [>...] - ETA: 1s - loss: 1.8076 - accuracy:
0.7540
11/96 [==>...] - ETA: Os - loss: 1.7828 - accuracy:
0.7532
16/96 [====>...] - ETA: Os - loss: 1.7300 - accuracy:
0.7798
20/96 [====>...] - ETA: Os - loss: 1.7406 - accuracy:
0.7810
24/96 [=====>...] - ETA: Os - loss: 1.6992 - accuracy:
0.7996
29/96 [======>...] - ETA: Os - loss: 1.6816 - accuracy:
0.7997
34/96 [=======>...] - ETA: Os - loss: 1.6647 - accuracy:
0.7983
38/96 [=======>...] - ETA: Os - loss: 1.6537 - accuracy:
0.7995
42/96 [========>...] - ETA: Os - loss: 1.6358 - accuracy:
0.8050
0.8085
0.8136
0.8125
0.8102
0.8148
0.8133
0.8143
```

```
0.8182
0.8206 - val_loss: 1.9147 - val_accuracy: 0.6789
Epoch 5/30
1/96 [...] - ETA: 2s - loss: 1.0769 - accuracy:
0.9048
4/96 [>...] - ETA: 1s - loss: 1.0437 - accuracy:
0.9167
9/96 [=>...] - ETA: 1s - loss: 1.0611 - accuracy:
0.9048
13/96 [===>...] - ETA: 1s - loss: 1.1287 - accuracy:
0.8828
17/96 [====>...] - ETA: 1s - loss: 1.1355 - accuracy:
0.8824
21/96 [====>...] - ETA: 1s - loss: 1.1459 - accuracy:
0.8685
25/96 [=====>...] - ETA: Os - loss: 1.1536 - accuracy:
0.8667
28/96 [======>...] - ETA: Os - loss: 1.1492 - accuracy:
32/96 [=======>...] - ETA: Os - loss: 1.1732 - accuracy:
0.8616
36/96 [======>...] - ETA: Os - loss: 1.1431 - accuracy:
0.8730
40/96 [========>...] - ETA: Os - loss: 1.1385 - accuracy:
44/96 [========>...] - ETA: Os - loss: 1.1302 - accuracy:
0.8745
0.8750
52/96 [=========>...] - ETA: Os - loss: 1.1108 - accuracy:
0.8764
0.8778
0.8798
0.8789
0.8793
0.8775
```

```
0.8762
0.8776
0.8772
0.8763
96/96 [=============== ] - ETA: Os - loss: 1.0584 - accuracy:
0.8761
0.8761 - val_loss: 1.7361 - val_accuracy: 0.6618
Epoch 6/30
1/96 [...] - ETA: 2s - loss: 0.8583 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 1.0568 - accuracy:
0.8476
9/96 [=>...] - ETA: 1s - loss: 1.0433 - accuracy:
0.8360
12/96 [==>...] - ETA: 1s - loss: 1.0013 - accuracy:
0.8571
16/96 [====>...] - ETA: 1s - loss: 0.9569 - accuracy:
0.8690
19/96 [====>...] - ETA: 1s - loss: 0.9344 - accuracy:
0.8797
22/96 [====>...] - ETA: 1s - loss: 0.9184 - accuracy:
0.8874
26/96 [======>...] - ETA: 1s - loss: 0.9218 - accuracy:
29/96 [======>...] - ETA: 1s - loss: 0.9138 - accuracy:
32/96 [======>...] - ETA: 1s - loss: 0.8980 - accuracy:
0.8943
35/96 [=======>...] - ETA: 1s - loss: 0.9168 - accuracy:
0.8803
38/96 [=======>...] - ETA: 0s - loss: 0.9243 - accuracy:
0.8772
42/96 [========>...] - ETA: Os - loss: 0.9139 - accuracy:
0.8798
45/96 [========>...] - ETA: Os - loss: 0.9208 - accuracy:
0.8741
0.8766
0.8772
```

```
0.8786
0.8785
0.8799
0.8803
0.8788
0.8755
0.8763
0.8752
0.8746 - val_loss: 1.5803 - val_accuracy: 0.6208
Epoch 7/30
1/96 [...] - ETA: 2s - loss: 0.6580 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 0.6926 - accuracy:
0.8952
8/96 [=>...] - ETA: 1s - loss: 0.6782 - accuracy:
0.9048
11/96 [==>...] - ETA: 1s - loss: 0.6835 - accuracy:
0.9134
15/96 [===>...] - ETA: 1s - loss: 0.6672 - accuracy:
18/96 [====>...] - ETA: 1s - loss: 0.6640 - accuracy:
0.9180
21/96 [====>...] - ETA: 1s - loss: 0.6540 - accuracy:
0.9206
24/96 [=====>...] - ETA: 1s - loss: 0.6469 - accuracy:
27/96 [======>...] - ETA: 1s - loss: 0.6417 - accuracy:
30/96 [======>...] - ETA: 1s - loss: 0.6440 - accuracy:
0.9222
33/96 [=======>...] - ETA: 1s - loss: 0.6852 - accuracy:
```

```
0.9091
36/96 [=======>...] - ETA: 1s - loss: 0.6826 - accuracy:
0.9087
39/96 [=======>...] - ETA: 1s - loss: 0.6697 - accuracy:
0.9133
42/96 [========>...] - ETA: Os - loss: 0.6711 - accuracy:
45/96 [=========>...] - ETA: Os - loss: 0.6620 - accuracy:
0.9164
0.9157
0.9150
0.9154
0.9147
0.9149
0.9157
0.9151
0.9165
0.9187
0.9187
96/96 [============= ] - ETA: Os - loss: 0.6332 - accuracy:
0.9200
0.9200 - val_loss: 1.7323 - val_accuracy: 0.7045
Epoch 8/30
1/96 [...] - ETA: 2s - loss: 0.4126 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.4776 - accuracy:
9/96 [=>...] - ETA: 1s - loss: 0.4727 - accuracy:
0.9524
```

```
13/96 [===>...] - ETA: 1s - loss: 0.5066 - accuracy:
0.9414
17/96 [====>...] - ETA: 1s - loss: 0.5149 - accuracy:
0.9468
20/96 [====>...] - ETA: 1s - loss: 0.5007 - accuracy:
0.9524
24/96 [=====>...] - ETA: 1s - loss: 0.5291 - accuracy:
0.9464
28/96 [======>...] - ETA: Os - loss: 0.5224 - accuracy:
0.9490
32/96 [=======>...] - ETA: Os - loss: 0.5353 - accuracy:
0.9390
36/96 [========>...] - ETA: Os - loss: 0.5263 - accuracy:
39/96 [=======>...] - ETA: Os - loss: 0.5183 - accuracy:
0.9463
0.9491
47/96 [========>...] - ETA: Os - loss: 0.5125 - accuracy:
0.9473
0.9468
0.9490
0.9500
0.9530
0.9505
0.9490
0.9453
0.9445
0.9449
96/96 [============ ] - 2s 18ms/step - loss: 0.4969 - accuracy:
0.9445 - val_loss: 1.5881 - val_accuracy: 0.5909
```

```
1/96 [...] - ETA: 2s - loss: 0.3619 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.3912 - accuracy:
0.9714
9/96 [=>...] - ETA: 1s - loss: 0.4148 - accuracy:
13/96 [===>...] - ETA: 1s - loss: 0.4420 - accuracy:
0.9597
17/96 [====>...] - ETA: 1s - loss: 0.4509 - accuracy:
0.9468
21/96 [====>...] - ETA: 1s - loss: 0.4387 - accuracy:
0.9501
25/96 [=====>...] - ETA: Os - loss: 0.4348 - accuracy:
0.9486
29/96 [======>...] - ETA: Os - loss: 0.4300 - accuracy:
0.9491
34/96 [=======>...] - ETA: Os - loss: 0.4254 - accuracy:
0.9510
38/96 [=======>...] - ETA: Os - loss: 0.4313 - accuracy:
0.9474
42/96 [========>...] - ETA: Os - loss: 0.4384 - accuracy:
0.9433
46/96 [========>...] - ETA: Os - loss: 0.4314 - accuracy:
0.9451
0.9467
0.9479
0.9490
0.9500
0.9488
0.9445
0.9446
96/96 [============= ] - ETA: Os - loss: 0.4265 - accuracy:
```

```
0.9430
0.9430 - val_loss: 1.8590 - val_accuracy: 0.6652
Epoch 10/30
1/96 [...] - ETA: 2s - loss: 0.3684 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.4446 - accuracy:
0.9365
10/96 [==>...] - ETA: 1s - loss: 0.4278 - accuracy:
0.9429
14/96 [===>...] - ETA: 1s - loss: 0.4391 - accuracy:
0.9320
18/96 [====>...] - ETA: 1s - loss: 0.4349 - accuracy:
0.9259
22/96 [====>...] - ETA: Os - loss: 0.4541 - accuracy:
0.9264
27/96 [======>...] - ETA: Os - loss: 0.4612 - accuracy:
0.9206
31/96 [======>...] - ETA: Os - loss: 0.4542 - accuracy:
0.9263
35/96 [=======>...] - ETA: Os - loss: 0.4532 - accuracy:
0.9265
39/96 [=======>...] - ETA: Os - loss: 0.4436 - accuracy:
0.9316
44/96 [========>...] - ETA: Os - loss: 0.4361 - accuracy:
0.9359
0.9337
0.9354
0.9365
0.9382
0.9376
0.9378
0.9380
```

```
0.9380
0.9402
0.9410 - val_loss: 1.2132 - val_accuracy: 0.7370
Epoch 11/30
1/96 [...] - ETA: 2s - loss: 0.2172 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.3041 - accuracy:
0.9619
9/96 [=>...] - ETA: 1s - loss: 0.3157 - accuracy:
0.9630
13/96 [===>...] - ETA: 1s - loss: 0.3139 - accuracy:
0.9634
17/96 [====>...] - ETA: 1s - loss: 0.3182 - accuracy:
0.9580
21/96 [====>...] - ETA: Os - loss: 0.3086 - accuracy:
0.9615
25/96 [=====>...] - ETA: Os - loss: 0.3152 - accuracy:
0.9581
28/96 [======>...] - ETA: Os - loss: 0.3108 - accuracy:
0.9592
32/96 [=======>...] - ETA: Os - loss: 0.3059 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.3085 - accuracy:
0.9616
40/96 [========>...] - ETA: Os - loss: 0.3075 - accuracy:
0.9619
0.9610
0.9597
0.9609
0.9595
0.9552
```

```
0.9537
0.9549
0.9560
0.9564
0.9562
0.9560
0.9564
0.9565 - val_loss: 1.5755 - val_accuracy: 0.6413
Epoch 12/30
1/96 [...] - ETA: 1s - loss: 0.2309 - accuracy:
0.9524
6/96 [>...] - ETA: 1s - loss: 0.3660 - accuracy:
0.9365
10/96 [==>...] - ETA: 1s - loss: 0.3123 - accuracy:
0.9571
14/96 [===>...] - ETA: 1s - loss: 0.3324 - accuracy:
0.9524
18/96 [====>...] - ETA: 1s - loss: 0.3018 - accuracy:
0.9630
21/96 [====>...] - ETA: 1s - loss: 0.3005 - accuracy:
0.9615
25/96 [=====>...] - ETA: 1s - loss: 0.2951 - accuracy:
0.9657
29/96 [======>...] - ETA: Os - loss: 0.2883 - accuracy:
0.9655
33/96 [=======>...] - ETA: Os - loss: 0.2846 - accuracy:
0.9668
38/96 [=======>...] - ETA: 0s - loss: 0.2858 - accuracy:
0.9649
42/96 [========>...] - ETA: Os - loss: 0.2801 - accuracy:
0.9671
46/96 [========>...] - ETA: Os - loss: 0.2820 - accuracy:
0.9658
0.9676
0.9696
```

```
0.9700
0.9701
0.9698
0.9695
0.9698
0.9607
96/96 [============== ] - ETA: Os - loss: 0.2859 - accuracy:
0.9615
0.9615 - val_loss: 1.2361 - val_accuracy: 0.7122
Epoch 13/30
1/96 [...] - ETA: 2s - loss: 0.2811 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 0.2496 - accuracy:
0.9619
9/96 [=>...] - ETA: 1s - loss: 0.2144 - accuracy:
0.9788
13/96 [===>...] - ETA: 1s - loss: 0.2195 - accuracy:
0.9744
16/96 [====>...] - ETA: 1s - loss: 0.2192 - accuracy:
0.9762
19/96 [====>...] - ETA: 1s - loss: 0.2123 - accuracy:
23/96 [=====>...] - ETA: 1s - loss: 0.2285 - accuracy:
0.9752
26/96 [======>...] - ETA: 1s - loss: 0.2252 - accuracy:
0.9780
29/96 [======>...] - ETA: 1s - loss: 0.2275 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 0.2219 - accuracy:
0.9798
37/96 [=======>...] - ETA: Os - loss: 0.2386 - accuracy:
0.9743
40/96 [=======>...] - ETA: Os - loss: 0.2363 - accuracy:
```

```
0.9738
44/96 [========>...] - ETA: Os - loss: 0.2484 - accuracy:
0.9729
0.9732
0.9735
0.9749
0.9746
0.9735
0.9744
0.9738
0.9743
0.9718
0.9734
0.9740 - val_loss: 1.4144 - val_accuracy: 0.6849
Epoch 14/30
1/96 [...] - ETA: 2s - loss: 0.1870 - accuracy:
1.0000
4/96 [>...] - ETA: 1s - loss: 0.1961 - accuracy:
0.9881
7/96 [=>...] - ETA: 1s - loss: 0.2358 - accuracy:
0.9660
10/96 [==>...] - ETA: 1s - loss: 0.2456 - accuracy:
0.9667
13/96 [===>...] - ETA: 1s - loss: 0.2752 - accuracy:
16/96 [====>...] - ETA: 1s - loss: 0.2726 - accuracy:
0.9524
```

```
19/96 [====>...] - ETA: 1s - loss: 0.2602 - accuracy:
0.9574
22/96 [====>...] - ETA: 1s - loss: 0.2520 - accuracy:
0.9610
25/96 [=====>...] - ETA: 1s - loss: 0.2456 - accuracy:
0.9619
29/96 [======>...] - ETA: 1s - loss: 0.2636 - accuracy:
0.9589
33/96 [=======>...] - ETA: 1s - loss: 0.2559 - accuracy:
0.9610
37/96 [=======>...] - ETA: 1s - loss: 0.2543 - accuracy:
0.9614
41/96 [=======>...] - ETA: Os - loss: 0.2508 - accuracy:
44/96 [========>...] - ETA: Os - loss: 0.2550 - accuracy:
0.9578
0.9554
0.9552
0.9551
0.9515
0.9524
0.9508
0.9504
0.9517
0.9524
0.9530
0.9547
0.9576
```

```
0.9579
96/96 [============ ] - 2s 25ms/step - loss: 0.2581 - accuracy:
0.9580 - val_loss: 1.3328 - val_accuracy: 0.7284
Epoch 15/30
1/96 [...] - ETA: 2s - loss: 0.1334 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.1973 - accuracy:
0.9619
9/96 [=>...] - ETA: 1s - loss: 0.1849 - accuracy:
0.9683
13/96 [===>...] - ETA: 1s - loss: 0.1751 - accuracy:
0.9780
17/96 [====>...] - ETA: 1s - loss: 0.1765 - accuracy:
0.9776
21/96 [====>...] - ETA: 1s - loss: 0.1741 - accuracy:
0.9773
25/96 [=====>...] - ETA: 1s - loss: 0.1925 - accuracy:
0.9771
29/96 [======>...] - ETA: Os - loss: 0.2005 - accuracy:
0.9737
33/96 [=======>...] - ETA: Os - loss: 0.1938 - accuracy:
0.9769
36/96 [======>...] - ETA: Os - loss: 0.1894 - accuracy:
0.9788
39/96 [========>...] - ETA: Os - loss: 0.1875 - accuracy:
43/96 [========>...] - ETA: Os - loss: 0.1964 - accuracy:
0.9779
46/96 [========>...] - ETA: Os - loss: 0.1925 - accuracy:
0.9793
0.9781
0.9811
0.9808
0.9805
0.9804
```

```
0.9790
0.9800
0.9804
0.9797
0.9805
0.9810 - val_loss: 1.4825 - val_accuracy: 0.7165
Epoch 1/30
1/96 [...] - ETA: 5:03 - loss: 7.7009 - accuracy:
0.0952
6/96 [>...] - ETA: 1s - loss: 7.7449 - accuracy:
0.1508
10/96 [==>...] - ETA: 1s - loss: 7.5132 - accuracy:
0.1524
15/96 [===>...] - ETA: Os - loss: 7.2396 - accuracy:
0.1683
20/96 [====>...] - ETA: Os - loss: 6.9257 - accuracy:
0.2071
26/96 [======>...] - ETA: Os - loss: 6.7001 - accuracy:
0.2198
31/96 [======>...] - ETA: Os - loss: 6.5062 - accuracy:
0.2320
35/96 [=======>...] - ETA: Os - loss: 6.3815 - accuracy:
40/96 [========>...] - ETA: Os - loss: 6.2526 - accuracy:
0.2640
0.2633
0.2704
0.2750
0.2822
0.2897
0.2960
```

```
0.2992
0.3048 - val_loss: 5.8966 - val_accuracy: 0.2340
Epoch 2/30
1/96 [...] - ETA: 2s - loss: 3.8115 - accuracy:
0.5714
6/96 [>...] - ETA: Os - loss: 3.9718 - accuracy:
0.5000
12/96 [==>...] - ETA: Os - loss: 3.8955 - accuracy:
0.5000
17/96 [====>...] - ETA: Os - loss: 3.9097 - accuracy:
0.4818
22/96 [====>...] - ETA: Os - loss: 3.8994 - accuracy:
0.4719
27/96 [======>...] - ETA: Os - loss: 3.8849 - accuracy:
0.4709
32/96 [=======>...] - ETA: Os - loss: 3.8286 - accuracy:
38/96 [========>...] - ETA: Os - loss: 3.8063 - accuracy:
0.4862
0.4806
0.4753
0.4779
0.4837
0.4937
0.4936
0.5011
0.5033
```

```
0.5037 - val_loss: 3.4734 - val_accuracy: 0.3903
Epoch 3/30
1/96 [...] - ETA: 2s - loss: 2.6874 - accuracy:
0.6667
5/96 [>...] - ETA: 1s - loss: 2.9975 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 2.8815 - accuracy:
0.6000
16/96 [====>...] - ETA: Os - loss: 2.7977 - accuracy:
0.6071
22/96 [====>...] - ETA: Os - loss: 2.8467 - accuracy:
0.5909
28/96 [======>...] - ETA: Os - loss: 2.7724 - accuracy:
0.6173
34/96 [======>...] - ETA: Os - loss: 2.7461 - accuracy:
0.6162
40/96 [========>...] - ETA: Os - loss: 2.7260 - accuracy:
0.6190
0.6232
0.6310
0.6301
0.6292
0.6290
0.6334
0.6341
0.6349
0.6340
0.6322
0.6322 - val_loss: 2.9604 - val_accuracy: 0.4415
Epoch 4/30
1/96 [...] - ETA: 2s - loss: 1.9174 - accuracy:
0.7619
5/96 [>...] - ETA: 1s - loss: 1.8445 - accuracy:
```

```
0.7905
10/96 [==>...] - ETA: 1s - loss: 2.0341 - accuracy:
0.7143
14/96 [===>...] - ETA: 1s - loss: 2.0221 - accuracy:
0.7177
18/96 [====>...] - ETA: Os - loss: 2.0227 - accuracy:
23/96 [=====>...] - ETA: Os - loss: 2.0322 - accuracy:
27/96 [======>...] - ETA: Os - loss: 2.0145 - accuracy:
0.7266
31/96 [======>...] - ETA: Os - loss: 2.0204 - accuracy:
0.7250
35/96 [=======>...] - ETA: Os - loss: 2.0196 - accuracy:
40/96 [========>...] - ETA: Os - loss: 2.0116 - accuracy:
0.7155
0.7217
0.7267
0.7238
0.7237
0.7262
0.7245
0.7153
0.7156 - val_loss: 2.3464 - val_accuracy: 0.5389
Epoch 5/30
1/96 [...] - ETA: 2s - loss: 1.6023 - accuracy:
5/96 [>...] - ETA: 1s - loss: 1.6138 - accuracy:
0.7714
```

```
9/96 [=>...] - ETA: 1s - loss: 1.6505 - accuracy:
0.7672
13/96 [===>...] - ETA: 1s - loss: 1.6155 - accuracy:
0.7839
17/96 [====>...] - ETA: 1s - loss: 1.6265 - accuracy:
0.7731
21/96 [====>...] - ETA: 1s - loss: 1.6139 - accuracy:
0.7710
26/96 [======>...] - ETA: Os - loss: 1.6029 - accuracy:
0.7692
30/96 [=======>...] - ETA: Os - loss: 1.5920 - accuracy:
0.7778
34/96 [======>...] - ETA: Os - loss: 1.5758 - accuracy:
39/96 [=======>...] - ETA: Os - loss: 1.5743 - accuracy:
0.7749
0.7659
0.7747
0.7760
0.7758
0.7695
0.7690
0.7664
0.7673
0.7671
0.7676 - val_loss: 2.0635 - val_accuracy: 0.5551
Epoch 6/30
1/96 [...] - ETA: 2s - loss: 1.3218 - accuracy:
0.7619
5/96 [>...] - ETA: 1s - loss: 1.3483 - accuracy:
```

```
0.7810
10/96 [==>...] - ETA: 1s - loss: 1.3791 - accuracy:
0.7619
16/96 [====>...] - ETA: Os - loss: 1.3363 - accuracy:
0.7857
21/96 [====>...] - ETA: Os - loss: 1.3519 - accuracy:
27/96 [======>...] - ETA: Os - loss: 1.3002 - accuracy:
0.8042
31/96 [======>...] - ETA: Os - loss: 1.3159 - accuracy:
0.7972
36/96 [=======>...] - ETA: Os - loss: 1.2869 - accuracy:
0.8095
41/96 [========>...] - ETA: Os - loss: 1.2808 - accuracy:
0.8075
0.8114
0.8129
0.8122
0.8152
0.8169
96/96 [============= ] - ETA: Os - loss: 1.2272 - accuracy:
0.8171
0.8171 - val_loss: 1.8805 - val_accuracy: 0.5816
Epoch 7/30
1/96 [...] - ETA: 2s - loss: 0.8383 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 1.0318 - accuracy:
0.8095
11/96 [==>...] - ETA: Os - loss: 1.0930 - accuracy:
17/96 [====>...] - ETA: Os - loss: 1.1259 - accuracy:
0.8039
```

```
22/96 [====>...] - ETA: Os - loss: 1.1123 - accuracy:
0.8074
28/96 [======>...] - ETA: Os - loss: 1.0852 - accuracy:
0.8180
33/96 [=======>...] - ETA: Os - loss: 1.1011 - accuracy:
0.8081
0.8181
0.8233
0.8210
0.8226
0.8249
0.8275
0.8274
0.8234
0.8242
0.8256 - val_loss: 1.7244 - val_accuracy: 0.5986
Epoch 8/30
1/96 [...] - ETA: 1s - loss: 0.8458 - accuracy:
0.9048
7/96 [=>...] - ETA: Os - loss: 0.9754 - accuracy:
0.8299
13/96 [===>...] - ETA: Os - loss: 0.9934 - accuracy:
18/96 [====>...] - ETA: Os - loss: 0.9809 - accuracy:
0.8254
24/96 [=====>...] - ETA: Os - loss: 0.9765 - accuracy:
0.8313
30/96 [======>...] - ETA: Os - loss: 0.9583 - accuracy:
0.8365
36/96 [======>...] - ETA: Os - loss: 0.9413 - accuracy:
42/96 [========>...] - ETA: Os - loss: 0.9420 - accuracy:
0.8458
```

```
0.8482
0.8501
0.8516
0.8564
0.8592
0.8615
0.8600
0.8556
0.8571 - val_loss: 1.5693 - val_accuracy: 0.6473
Epoch 9/30
1/96 [...] - ETA: 2s - loss: 0.6893 - accuracy:
0.9048
6/96 [>...] - ETA: Os - loss: 0.7593 - accuracy:
0.8968
12/96 [==>...] - ETA: Os - loss: 0.7645 - accuracy:
0.8968
18/96 [====>...] - ETA: Os - loss: 0.7848 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 0.7699 - accuracy:
28/96 [======>...] - ETA: Os - loss: 0.7963 - accuracy:
0.8827
34/96 [======>...] - ETA: Os - loss: 0.7903 - accuracy:
0.8838
38/96 [=======>...] - ETA: 0s - loss: 0.8102 - accuracy:
0.8772
44/96 [========>...] - ETA: Os - loss: 0.8053 - accuracy:
0.8788
0.8824
0.8757
```

```
0.8728
0.8742
0.8752
0.8735
0.8721 - val_loss: 1.6371 - val_accuracy: 0.6695
Epoch 10/30
1/96 [...] - ETA: 1s - loss: 0.7335 - accuracy:
0.8571
6/96 [>...] - ETA: Os - loss: 0.7223 - accuracy:
0.8730
11/96 [==>...] - ETA: Os - loss: 0.7480 - accuracy:
0.8615
17/96 [====>...] - ETA: Os - loss: 0.7729 - accuracy:
0.8515
23/96 [=====>...] - ETA: Os - loss: 0.7611 - accuracy:
0.8613
29/96 [======>...] - ETA: Os - loss: 0.7406 - accuracy:
0.8736
33/96 [=======>...] - ETA: Os - loss: 0.7427 - accuracy:
38/96 [========>...] - ETA: Os - loss: 0.7326 - accuracy:
0.8772
44/96 [========>...] - ETA: Os - loss: 0.7282 - accuracy:
0.8799
0.8857
0.8896
0.8873
0.8872
0.8879
```

```
0.8837
0.8835
96/96 [============= ] - 1s 13ms/step - loss: 0.7128 - accuracy:
0.8836 - val_loss: 1.7729 - val_accuracy: 0.5995
Epoch 11/30
1/96 [...] - ETA: 2s - loss: 0.6927 - accuracy:
0.8095
6/96 [>...] - ETA: Os - loss: 0.5722 - accuracy:
0.9206
11/96 [==>...] - ETA: Os - loss: 0.6327 - accuracy:
0.9048
17/96 [====>...] - ETA: Os - loss: 0.6281 - accuracy:
0.9076
22/96 [====>...] - ETA: Os - loss: 0.6069 - accuracy:
0.9177
27/96 [======>...] - ETA: Os - loss: 0.6063 - accuracy:
0.9171
32/96 [=======>...] - ETA: Os - loss: 0.6040 - accuracy:
0.9196
37/96 [=======>...] - ETA: Os - loss: 0.6091 - accuracy:
0.9151
42/96 [========>...] - ETA: Os - loss: 0.6167 - accuracy:
0.9150
0.9086
0.9083
0.9073
0.9077
0.9064
0.9032
96/96 [============= ] - ETA: Os - loss: 0.6466 - accuracy:
0.9000 - val_loss: 1.9627 - val_accuracy: 0.5858
```

```
Epoch 12/30
```

```
1/96 [...] - ETA: 1s - loss: 0.6402 - accuracy:
0.9048
6/96 [>...] - ETA: Os - loss: 0.6278 - accuracy:
12/96 [==>...] - ETA: Os - loss: 0.6306 - accuracy:
0.9048
17/96 [====>...] - ETA: Os - loss: 0.6198 - accuracy:
0.9020
22/96 [====>...] - ETA: Os - loss: 0.6107 - accuracy:
0.9069
28/96 [======>...] - ETA: Os - loss: 0.6255 - accuracy:
0.9065
33/96 [=======>...] - ETA: Os - loss: 0.6237 - accuracy:
0.9076
38/96 [=======>...] - ETA: Os - loss: 0.6424 - accuracy:
0.8997
44/96 [========>...] - ETA: Os - loss: 0.6494 - accuracy:
0.8983
0.8970
0.9009
0.9033
0.9041
0.9054
0.9025
0.9000
96/96 [============ ] - ETA: Os - loss: 0.6428 - accuracy:
0.8996
0.8996 - val_loss: 1.7340 - val_accuracy: 0.6447
Epoch 13/30
1/96 [...] - ETA: 1s - loss: 0.4768 - accuracy:
7/96 [=>...] - ETA: Os - loss: 0.6110 - accuracy:
0.9116
```

```
12/96 [==>...] - ETA: Os - loss: 0.5772 - accuracy:
0.9206
16/96 [====>...] - ETA: Os - loss: 0.5669 - accuracy:
0.9226
21/96 [====>...] - ETA: Os - loss: 0.5630 - accuracy:
0.9274
27/96 [======>...] - ETA: Os - loss: 0.5681 - accuracy:
0.9277
32/96 [=======>...] - ETA: Os - loss: 0.5563 - accuracy:
0.9315
37/96 [=======>...] - ETA: Os - loss: 0.5563 - accuracy:
0.9318
41/96 [=======>...] - ETA: Os - loss: 0.5548 - accuracy:
0.9244
0.9256
0.9230
0.9221
0.9220
96/96 [============== ] - ETA: Os - loss: 0.5745 - accuracy:
0.9215
0.9215 - val_loss: 1.5994 - val_accuracy: 0.6311
Epoch 14/30
1/96 [...] - ETA: 1s - loss: 0.4395 - accuracy:
0.9524
6/96 [>...] - ETA: 1s - loss: 0.5333 - accuracy:
0.9048
11/96 [==>...] - ETA: Os - loss: 0.5258 - accuracy:
0.9221
16/96 [====>...] - ETA: Os - loss: 0.5238 - accuracy:
0.9167
21/96 [====>...] - ETA: Os - loss: 0.5005 - accuracy:
```

```
0.9297
25/96 [=====>...] - ETA: Os - loss: 0.4991 - accuracy:
0.9314
29/96 [======>...] - ETA: Os - loss: 0.4960 - accuracy:
0.9294
33/96 [=======>...] - ETA: Os - loss: 0.4966 - accuracy:
38/96 [=======>...] - ETA: Os - loss: 0.4986 - accuracy:
0.9286
42/96 [========>...] - ETA: Os - loss: 0.5061 - accuracy:
0.9240
0.9220
0.9176
0.9209
0.9206
0.9216
0.9176
0.9174
0.9108
0.9110 - val_loss: 1.7087 - val_accuracy: 0.6302
Epoch 1/30
1/96 [...] - ETA: 4:07 - loss: 24.5731 - accuracy:
0.0952
6/96 [>...] - ETA: 1s - loss: 32.7322 - accuracy:
0.1508
11/96 [==>...] - ETA: Os - loss: 33.2398 - accuracy:
16/96 [====>...] - ETA: Os - loss: 30.5423 - accuracy:
20/96 [====>...] - ETA: Os - loss: 28.3431 - accuracy:
0.2000
```

```
25/96 [=====>...] - ETA: Os - loss: 26.2303 - accuracy:
0.2133
31/96 [======>...] - ETA: Os - loss: 24.3287 - accuracy:
0.2197
37/96 [========>...] - ETA: Os - loss: 22.9080 - accuracy:
0.2214
43/96 [========>...] - ETA: Os - loss: 21.7501 - accuracy:
0.2303
0.2300
0.2336
0.2478
0.2490
0.2469
0.2536
0.2563
0.2627
96/96 [============= ] - 4s 16ms/step - loss: 16.7149 -
accuracy: 0.2629 - val_loss: 11.8462 - val_accuracy: 0.2058
Epoch 2/30
1/96 [...] - ETA: 1s - loss: 10.9308 - accuracy:
0.2857
7/96 [=>...] - ETA: Os - loss: 10.7075 - accuracy:
0.3537
13/96 [===>...] - ETA: Os - loss: 10.5780 - accuracy:
19/96 [====>...] - ETA: Os - loss: 10.4706 - accuracy:
0.3559
24/96 [=====>...] - ETA: Os - loss: 10.3490 - accuracy:
0.3532
30/96 [======>...] - ETA: Os - loss: 10.2074 - accuracy:
0.3635
36/96 [=======>...] - ETA: Os - loss: 10.0793 - accuracy:
0.3651
43/96 [========>...] - ETA: Os - loss: 9.9291 - accuracy:
0.3721
```

```
0.3712
0.3774
0.3770
0.3824
0.3783
0.3751
0.3728
0.3723 - val_loss: 7.9015 - val_accuracy: 0.2562
Epoch 3/30
1/96 [...] - ETA: 1s - loss: 7.7156 - accuracy:
0.3333
5/96 [>...] - ETA: 1s - loss: 7.3254 - accuracy:
0.3905
11/96 [==>...] - ETA: Os - loss: 7.1511 - accuracy:
0.4156
17/96 [====>...] - ETA: Os - loss: 7.0324 - accuracy:
23/96 [=====>...] - ETA: Os - loss: 6.9728 - accuracy:
28/96 [======>...] - ETA: Os - loss: 6.9054 - accuracy:
0.4082
33/96 [======>...] - ETA: Os - loss: 6.8516 - accuracy:
0.4026
38/96 [=======>...] - ETA: 0s - loss: 6.7776 - accuracy:
0.4085
44/96 [========>...] - ETA: Os - loss: 6.7078 - accuracy:
0.4080
0.4171
0.4269
0.4231
```

```
0.4305
0.4313
0.4347
0.4353 - val_loss: 5.6129 - val_accuracy: 0.3134
Epoch 4/30
1/96 [...] - ETA: 1s - loss: 4.7029 - accuracy:
0.4762
8/96 [=>...] - ETA: Os - loss: 4.9979 - accuracy:
0.4226
14/96 [===>...] - ETA: Os - loss: 4.9153 - accuracy:
0.4456
19/96 [====>...] - ETA: Os - loss: 4.9107 - accuracy:
0.4436
25/96 [=====>...] - ETA: Os - loss: 4.8494 - accuracy:
31/96 [=======>...] - ETA: Os - loss: 4.7787 - accuracy:
0.4654
35/96 [=======>...] - ETA: Os - loss: 4.7477 - accuracy:
0.4707
40/96 [========>...] - ETA: Os - loss: 4.7021 - accuracy:
44/96 [========>...] - ETA: Os - loss: 4.6804 - accuracy:
0.4762
0.4857
0.4872
0.4860
0.4808
0.4774
0.4773
0.4831
```

```
0.4857
96/96 [============= ] - 1s 12ms/step - loss: 4.3662 - accuracy:
0.4853 - val_loss: 4.3956 - val_accuracy: 0.3826
Epoch 5/30
1/96 [...] - ETA: 1s - loss: 4.1805 - accuracy:
0.3810
7/96 [=>...] - ETA: Os - loss: 3.6439 - accuracy:
0.5646
13/96 [===>...] - ETA: Os - loss: 3.6686 - accuracy:
0.5641
18/96 [====>...] - ETA: Os - loss: 3.6701 - accuracy:
0.5423
23/96 [=====>...] - ETA: Os - loss: 3.6392 - accuracy:
0.5445
29/96 [======>...] - ETA: Os - loss: 3.5721 - accuracy:
0.5616
35/96 [=======>...] - ETA: Os - loss: 3.5471 - accuracy:
0.5524
41/96 [=======>...] - ETA: Os - loss: 3.5199 - accuracy:
45/96 [=========>...] - ETA: Os - loss: 3.5013 - accuracy:
0.5481
0.5425
0.5414
0.5486
0.5530
0.5570
0.5535
0.5522 - val_loss: 3.5111 - val_accuracy: 0.4125
Epoch 6/30
1/96 [...] - ETA: 1s - loss: 2.7486 - accuracy:
0.6190
7/96 [=>...] - ETA: Os - loss: 2.7074 - accuracy:
```

```
0.6735
14/96 [===>...] - ETA: Os - loss: 2.8241 - accuracy:
0.6156
20/96 [====>...] - ETA: Os - loss: 2.7935 - accuracy:
0.6214
27/96 [======>...] - ETA: Os - loss: 2.7769 - accuracy:
34/96 [=======>...] - ETA: Os - loss: 2.7456 - accuracy:
0.6022
40/96 [=======>...] - ETA: Os - loss: 2.7341 - accuracy:
0.6024
0.5942
0.5971
0.6021
0.5993
0.5962
0.5979
96/96 [============ ] - 1s 12ms/step - loss: 2.5937 - accuracy:
0.5987 - val_loss: 3.3629 - val_accuracy: 0.3655
Epoch 7/30
1/96 [...] - ETA: 2s - loss: 2.2647 - accuracy:
0.6190
7/96 [=>...] - ETA: 0s - loss: 2.6126 - accuracy:
0.5986
13/96 [===>...] - ETA: Os - loss: 2.4808 - accuracy:
0.6081
19/96 [====>...] - ETA: Os - loss: 2.3529 - accuracy:
0.6241
24/96 [=====>...] - ETA: Os - loss: 2.3275 - accuracy:
0.6270
30/96 [=======>...] - ETA: Os - loss: 2.3570 - accuracy:
0.6095
36/96 [========>...] - ETA: Os - loss: 2.3029 - accuracy:
41/96 [========>...] - ETA: Os - loss: 2.2653 - accuracy:
0.6341
```

```
0.6408
0.6416
0.6410
0.6401
0.6387
0.6386
0.6396
96/96 [============ ] - 1s 14ms/step - loss: 2.1790 - accuracy:
0.6387 - val_loss: 2.6123 - val_accuracy: 0.5098
Epoch 8/30
1/96 [...] - ETA: 1s - loss: 2.1267 - accuracy:
0.6667
5/96 [>...] - ETA: 1s - loss: 1.8719 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 1.9119 - accuracy:
0.6857
15/96 [===>...] - ETA: Os - loss: 1.8528 - accuracy:
0.6794
20/96 [====>...] - ETA: Os - loss: 1.8387 - accuracy:
0.6857
24/96 [=====>...] - ETA: Os - loss: 1.8112 - accuracy:
29/96 [======>...] - ETA: Os - loss: 1.7659 - accuracy:
0.7126
34/96 [=======>...] - ETA: Os - loss: 1.7471 - accuracy:
0.7157
38/96 [========>...] - ETA: Os - loss: 1.7433 - accuracy:
43/96 [========>...] - ETA: Os - loss: 1.7543 - accuracy:
0.7043
0.7063
```

```
0.7008
0.6984
0.7051
0.7077
0.7062
0.7033
0.7012
96/96 [============= ] - ETA: Os - loss: 1.7768 - accuracy:
0.7006
0.7006 - val_loss: 2.7540 - val_accuracy: 0.4902
Epoch 9/30
1/96 [...] - ETA: 1s - loss: 1.4010 - accuracy:
0.7619
5/96 [>...] - ETA: 1s - loss: 1.6229 - accuracy:
0.7429
10/96 [==>...] - ETA: 1s - loss: 1.6616 - accuracy:
15/96 [===>...] - ETA: Os - loss: 1.6982 - accuracy:
0.7079
19/96 [====>...] - ETA: Os - loss: 1.6741 - accuracy:
0.7218
23/96 [=====>...] - ETA: Os - loss: 1.6391 - accuracy:
0.7391
29/96 [======>...] - ETA: Os - loss: 1.6018 - accuracy:
0.7504
34/96 [=======>...] - ETA: Os - loss: 1.5757 - accuracy:
0.7591
39/96 [=======>...] - ETA: Os - loss: 1.5760 - accuracy:
0.7570
44/96 [=======>...] - ETA: Os - loss: 1.5640 - accuracy:
0.7608
0.7584
```

```
0.7586
0.7598
0.7593
0.7514
0.7451
0.7365
0.7381 - val_loss: 2.6609 - val_accuracy: 0.4902
Epoch 10/30
1/96 [...] - ETA: 1s - loss: 3.2064 - accuracy:
7/96 [=>...] - ETA: Os - loss: 1.9268 - accuracy:
0.7143
13/96 [===>...] - ETA: Os - loss: 1.8349 - accuracy:
0.7143
19/96 [====>...] - ETA: Os - loss: 1.7757 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 1.7238 - accuracy:
0.7440
30/96 [======>...] - ETA: Os - loss: 1.7213 - accuracy:
0.7365
36/96 [======>...] - ETA: Os - loss: 1.7129 - accuracy:
0.7421
41/96 [=======>...] - ETA: Os - loss: 1.6975 - accuracy:
46/96 [=========>...] - ETA: Os - loss: 1.7179 - accuracy:
0.7463
58/96 [=========>...] - ETA: Os - loss: 1.6656 - accuracy:
0.7537
0.7604
0.7551
```

```
0.7556
0.7625
0.7603
0.7634
0.7626 - val_loss: 2.5133 - val_accuracy: 0.4953
Epoch 11/30
1/96 [...] - ETA: 2s - loss: 1.5255 - accuracy:
0.7143
7/96 [=>...] - ETA: Os - loss: 1.2878 - accuracy:
0.8095
13/96 [===>...] - ETA: Os - loss: 1.3592 - accuracy:
0.7875
19/96 [====>...] - ETA: Os - loss: 1.3346 - accuracy:
0.7920
25/96 [=====>...] - ETA: Os - loss: 1.3234 - accuracy:
0.7962
29/96 [======>...] - ETA: Os - loss: 1.3038 - accuracy:
0.8013
34/96 [=======>...] - ETA: Os - loss: 1.2858 - accuracy:
0.8095
40/96 [========>...] - ETA: Os - loss: 1.2624 - accuracy:
46/96 [========>...] - ETA: Os - loss: 1.2533 - accuracy:
0.8194
0.8242
0.8181
0.8199
0.8197
0.8169
96/96 [============ ] - 1s 13ms/step - loss: 1.2329 - accuracy:
0.8166 - val_loss: 2.1666 - val_accuracy: 0.5781
```

Epoch 12/30

```
1/96 [...] - ETA: 2s - loss: 1.1007 - accuracy:
0.9048
6/96 [>...] - ETA: Os - loss: 1.1767 - accuracy:
0.8175
11/96 [==>...] - ETA: Os - loss: 1.1134 - accuracy:
17/96 [====>...] - ETA: Os - loss: 1.1238 - accuracy:
0.8291
24/96 [=====>...] - ETA: Os - loss: 1.1086 - accuracy:
0.8353
31/96 [======>...] - ETA: Os - loss: 1.1033 - accuracy:
0.8418
38/96 [=======>...] - ETA: Os - loss: 1.1227 - accuracy:
43/96 [========>...] - ETA: Os - loss: 1.1292 - accuracy:
0.8350
0.8367
0.8398
0.8394
0.8311
0.8247
0.8251 - val loss: 2.4704 - val accuracy: 0.5482
Epoch 13/30
1/96 [...] - ETA: 2s - loss: 1.2371 - accuracy:
0.8571
6/96 [>...] - ETA: Os - loss: 1.1224 - accuracy:
0.8492
12/96 [==>...] - ETA: Os - loss: 1.1278 - accuracy:
18/96 [====>...] - ETA: Os - loss: 1.1084 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 1.1439 - accuracy:
0.8472
```

```
30/96 [======>...] - ETA: Os - loss: 1.1352 - accuracy:
0.8508
36/96 [=======>...] - ETA: Os - loss: 1.1157 - accuracy:
0.8571
42/96 [========>...] - ETA: Os - loss: 1.1180 - accuracy:
0.8526
0.8581
0.8554
0.8571
0.8565
0.8489
0.8436
0.8369
0.8351 - val_loss: 2.8943 - val_accuracy: 0.4868
Epoch 14/30
1/96 [...] - ETA: 1s - loss: 1.2689 - accuracy:
0.8095
8/96 [=>...] - ETA: Os - loss: 1.4161 - accuracy:
0.7857
15/96 [===>...] - ETA: Os - loss: 1.4503 - accuracy:
0.7810
21/96 [====>...] - ETA: Os - loss: 1.4975 - accuracy:
0.7823
27/96 [======>...] - ETA: Os - loss: 1.4686 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 1.4425 - accuracy:
39/96 [========>...] - ETA: Os - loss: 1.3979 - accuracy:
0.8132
44/96 [========>...] - ETA: Os - loss: 1.3958 - accuracy:
0.8160
0.8367
```

```
0.8387
0.8396
0.8417
0.8431
0.8446
0.8441 - val_loss: 2.3509 - val_accuracy: 0.5909
Epoch 15/30
1/96 [...] - ETA: 2s - loss: 1.3305 - accuracy:
0.8095
7/96 [=>...] - ETA: Os - loss: 1.1255 - accuracy:
0.8844
13/96 [===>...] - ETA: Os - loss: 1.1168 - accuracy:
0.8828
19/96 [====>...] - ETA: Os - loss: 1.1197 - accuracy:
0.8772
24/96 [=====>...] - ETA: Os - loss: 1.0807 - accuracy:
0.8869
29/96 [======>...] - ETA: Os - loss: 1.0862 - accuracy:
0.8818
35/96 [=======>...] - ETA: Os - loss: 1.0747 - accuracy:
42/96 [========>...] - ETA: Os - loss: 1.0479 - accuracy:
0.8929
0.8898
0.8862
0.8824
0.8793
0.8810
0.8791
```

```
0.8756 - val_loss: 2.1966 - val_accuracy: 0.5901
Epoch 16/30
1/96 [...] - ETA: 2s - loss: 0.8818 - accuracy:
7/96 [=>...] - ETA: Os - loss: 0.9229 - accuracy:
0.9048
14/96 [===>...] - ETA: Os - loss: 0.9382 - accuracy:
0.9014
19/96 [====>...] - ETA: Os - loss: 0.9566 - accuracy:
0.9023
25/96 [=====>...] - ETA: Os - loss: 0.9448 - accuracy:
31/96 [======>...] - ETA: Os - loss: 0.9584 - accuracy:
0.9078
37/96 [=======>...] - ETA: Os - loss: 0.9513 - accuracy:
0.9125
0.9104
0.9099
0.9086
0.9132
0.9106
0.9108
0.9053
0.9030 - val_loss: 2.2364 - val_accuracy: 0.6029
Epoch 17/30
1/96 [...] - ETA: 1s - loss: 0.8291 - accuracy:
6/96 [>...] - ETA: Os - loss: 0.9523 - accuracy:
0.8810
```

```
12/96 [==>...] - ETA: Os - loss: 0.8852 - accuracy:
0.9008
18/96 [====>...] - ETA: Os - loss: 0.9058 - accuracy:
0.9074
23/96 [=====>...] - ETA: Os - loss: 0.9348 - accuracy:
0.8944
28/96 [======>...] - ETA: Os - loss: 0.9311 - accuracy:
0.8980
34/96 [=======>...] - ETA: Os - loss: 0.9664 - accuracy:
0.8894
40/96 [=======>...] - ETA: Os - loss: 0.9868 - accuracy:
0.8881
45/96 [========>...] - ETA: Os - loss: 1.0027 - accuracy:
0.8895
0.8879
0.8929
0.8900
0.8879
0.8868
0.8875
0.8841 - val_loss: 2.5823 - val_accuracy: 0.5406
Epoch 18/30
1/96 [...] - ETA: 2s - loss: 0.8445 - accuracy:
1.0000
6/96 [>...] - ETA: Os - loss: 1.0279 - accuracy:
0.8889
11/96 [==>...] - ETA: Os - loss: 1.0624 - accuracy:
0.8745
16/96 [====>...] - ETA: Os - loss: 1.0337 - accuracy:
0.8839
21/96 [====>...] - ETA: Os - loss: 0.9967 - accuracy:
26/96 [======>...] - ETA: Os - loss: 0.9914 - accuracy:
0.8974
31/96 [=======>...] - ETA: Os - loss: 0.9735 - accuracy:
```

```
0.9032
36/96 [======>...] - ETA: Os - loss: 0.9713 - accuracy:
0.9048
41/96 [=======>...] - ETA: Os - loss: 0.9572 - accuracy:
0.9094
45/96 [========>...] - ETA: Os - loss: 0.9426 - accuracy:
0.9086
0.9083
0.9080
0.9033
0.8988
0.8947
0.8882
0.8838
96/96 [============== ] - ETA: Os - loss: 1.0338 - accuracy:
0.8831 - val_loss: 2.7759 - val_accuracy: 0.4962
Epoch 19/30
1/96 [...] - ETA: 1s - loss: 0.8808 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 1.1888 - accuracy:
0.8857
10/96 [==>...] - ETA: 1s - loss: 1.2648 - accuracy:
0.8667
15/96 [===>...] - ETA: Os - loss: 1.2802 - accuracy:
0.8635
19/96 [====>...] - ETA: Os - loss: 1.2626 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 1.2793 - accuracy:
28/96 [======>...] - ETA: Os - loss: 1.2755 - accuracy:
0.8622
```

```
33/96 [=======>...] - ETA: Os - loss: 1.2700 - accuracy:
0.8629
37/96 [=======>...] - ETA: Os - loss: 1.2482 - accuracy:
0.8662
0.8560
0.8571
0.8571
0.8589
59/96 [=========>...] - ETA: Os - loss: 1.2586 - accuracy:
0.8592
0.8571
0.8590
0.8600
0.8636
0.8618
0.8632
0.8626 - val_loss: 2.4993 - val_accuracy: 0.5756
Epoch 20/30
1/96 [...] - ETA: 2s - loss: 0.8553 - accuracy:
0.9524
6/96 [>...] - ETA: 1s - loss: 1.1296 - accuracy:
0.8730
11/96 [==>...] - ETA: Os - loss: 1.1008 - accuracy:
0.8745
15/96 [===>...] - ETA: Os - loss: 1.1161 - accuracy:
0.8794
19/96 [====>...] - ETA: Os - loss: 1.1344 - accuracy:
0.8647
24/96 [=====>...] - ETA: Os - loss: 1.1400 - accuracy:
0.8710
29/96 [=======>...] - ETA: Os - loss: 1.1243 - accuracy:
```

```
0.8768
33/96 [======>...] - ETA: Os - loss: 1.1267 - accuracy:
0.8788
38/96 [========>...] - ETA: Os - loss: 1.1322 - accuracy:
0.8759
43/96 [========>...] - ETA: Os - loss: 1.1213 - accuracy:
0.8800
53/96 [=========>...] - ETA: Os - loss: 1.1189 - accuracy:
0.8850
0.8894
0.8879
0.8905
0.8891
0.8924
0.8923
96/96 [============ ] - 1s 13ms/step - loss: 1.0927 - accuracy:
0.8901 - val_loss: 2.8106 - val_accuracy: 0.5816
Epoch 21/30
1/96 [...] - ETA: 1s - loss: 0.8073 - accuracy:
0.9524
8/96 [=>...] - ETA: Os - loss: 1.0587 - accuracy:
0.9226
14/96 [===>...] - ETA: Os - loss: 1.0744 - accuracy:
0.8980
20/96 [====>...] - ETA: Os - loss: 1.0522 - accuracy:
0.9048
25/96 [=====>...] - ETA: Os - loss: 1.0523 - accuracy:
0.8990
31/96 [=======>...] - ETA: Os - loss: 1.0490 - accuracy:
0.9078
37/96 [=======>...] - ETA: Os - loss: 1.0449 - accuracy:
0.9073
43/96 [========>...] - ETA: Os - loss: 1.0463 - accuracy:
0.9068
```

```
0.9084
0.9120
0.9087
0.9104
0.9088
0.9090
0.9085 - val_loss: 2.4071 - val_accuracy: 0.6174
Epoch 22/30
1/96 [...] - ETA: 2s - loss: 0.7350 - accuracy:
1.0000
7/96 [=>...] - ETA: Os - loss: 0.8444 - accuracy:
11/96 [==>...] - ETA: Os - loss: 0.8212 - accuracy:
0.9481
17/96 [====>...] - ETA: Os - loss: 0.7892 - accuracy:
0.9608
23/96 [=====>...] - ETA: Os - loss: 0.7799 - accuracy:
29/96 [======>...] - ETA: Os - loss: 0.7814 - accuracy:
0.9589
35/96 [=======>...] - ETA: Os - loss: 0.7930 - accuracy:
0.9578
40/96 [=======>...] - ETA: Os - loss: 0.7956 - accuracy:
0.9571
47/96 [========>...] - ETA: Os - loss: 0.7976 - accuracy:
0.9524
0.9539
0.9505
```

```
0.9501
0.9513
0.9514
0.9495 - val_loss: 2.2223 - val_accuracy: 0.6097
Epoch 23/30
1/96 [...] - ETA: 1s - loss: 0.7733 - accuracy:
0.9048
7/96 [=>...] - ETA: Os - loss: 0.8048 - accuracy:
0.9252
12/96 [==>...] - ETA: Os - loss: 0.7620 - accuracy:
0.9325
18/96 [====>...] - ETA: Os - loss: 0.8035 - accuracy:
0.9259
24/96 [=====>...] - ETA: Os - loss: 0.8047 - accuracy:
0.9306
30/96 [======>...] - ETA: Os - loss: 0.7836 - accuracy:
0.9365
35/96 [=======>...] - ETA: Os - loss: 0.7692 - accuracy:
0.9388
41/96 [=======>...] - ETA: Os - loss: 0.7595 - accuracy:
0.9396
0.9412
0.9421
0.9456
0.9442
0.9466
0.9448
0.9453
0.9450 - val_loss: 2.4112 - val_accuracy: 0.6080
```

```
1/96 [...] - ETA: 1s - loss: 0.5897 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.7127 - accuracy:
0.9365
12/96 [==>...] - ETA: Os - loss: 0.6901 - accuracy:
18/96 [====>...] - ETA: Os - loss: 0.7217 - accuracy:
0.9497
24/96 [=====>...] - ETA: Os - loss: 0.7425 - accuracy:
0.9385
30/96 [======>...] - ETA: Os - loss: 0.7523 - accuracy:
0.9397
35/96 [=======>...] - ETA: Os - loss: 0.7444 - accuracy:
40/96 [========>...] - ETA: Os - loss: 0.7391 - accuracy:
0.9393
0.9358
0.9377
0.9368
0.9367
0.9332
0.9334
0.9316
0.9325 - val_loss: 2.4519 - val_accuracy: 0.5209
Epoch 25/30
1/96 [...] - ETA: 1s - loss: 0.6173 - accuracy:
1.0000
8/96 [=>...] - ETA: Os - loss: 0.8135 - accuracy:
0.9286
15/96 [===>...] - ETA: Os - loss: 0.7849 - accuracy:
20/96 [====>...] - ETA: Os - loss: 0.7533 - accuracy:
0.9405
```

```
26/96 [======>...] - ETA: Os - loss: 0.7597 - accuracy:
0.9359
32/96 [=======>...] - ETA: Os - loss: 0.7679 - accuracy:
38/96 [=======>...] - ETA: 0s - loss: 0.7674 - accuracy:
0.9373
0.9329
0.9300
0.9268
0.9271
0.9239
0.9212
0.9189
0.9196
96/96 [============= ] - ETA: Os - loss: 0.8405 - accuracy:
0.9185
0.9185 - val_loss: 2.8997 - val_accuracy: 0.5952
Epoch 26/30
1/96 [...] - ETA: 1s - loss: 0.9779 - accuracy:
0.9048
7/96 [=>...] - ETA: Os - loss: 0.8573 - accuracy:
0.9252
12/96 [==>...] - ETA: Os - loss: 0.9319 - accuracy:
18/96 [====>...] - ETA: Os - loss: 0.9247 - accuracy:
0.9180
24/96 [=====>...] - ETA: Os - loss: 0.9391 - accuracy:
0.9127
30/96 [======>...] - ETA: Os - loss: 0.9557 - accuracy:
0.9063
36/96 [======>...] - ETA: Os - loss: 0.9463 - accuracy:
0.9114
42/96 [========>...] - ETA: Os - loss: 0.9375 - accuracy:
0.9150
```

```
0.9189
0.9222
0.9212
0.9195
0.9187
0.9184
0.9190 - val_loss: 2.4784 - val_accuracy: 0.5978
Epoch 1/30
1/96 [...] - ETA: 5:31 - loss: 7.1381 - accuracy:
0.0000e+00
6/96 [>...] - ETA: 1s - loss: 8.0132 - accuracy:
0.1190
10/96 [==>...] - ETA: 1s - loss: 7.5094 - accuracy:
0.1667
15/96 [===>...] - ETA: Os - loss: 7.2116 - accuracy:
19/96 [====>...] - ETA: Os - loss: 6.9752 - accuracy:
0.2005
24/96 [=====>...] - ETA: Os - loss: 6.6975 - accuracy:
0.2302
28/96 [======>...] - ETA: Os - loss: 6.5368 - accuracy:
0.2432
32/96 [=======>...] - ETA: Os - loss: 6.4195 - accuracy:
0.2455
37/96 [========>...] - ETA: Os - loss: 6.2764 - accuracy:
0.2535
41/96 [=======>...] - ETA: Os - loss: 6.1863 - accuracy:
0.2567
0.2624
50/96 [=========>...] - ETA: Os - loss: 5.9831 - accuracy:
0.2719
```

```
0.2768
0.2797
0.2850
0.2857
0.2902
0.2923
0.2921
0.2974
0.2994 - val_loss: 5.6337 - val_accuracy: 0.1896
Epoch 2/30
1/96 [...] - ETA: 1s - loss: 3.8530 - accuracy:
0.5238
6/96 [>...] - ETA: Os - loss: 3.9725 - accuracy:
0.4762
11/96 [==>...] - ETA: Os - loss: 3.9564 - accuracy:
0.4675
15/96 [===>...] - ETA: Os - loss: 3.9220 - accuracy:
0.4603
21/96 [====>...] - ETA: Os - loss: 3.9278 - accuracy:
0.4535
27/96 [======>...] - ETA: Os - loss: 3.8912 - accuracy:
0.4621
32/96 [=======>...] - ETA: Os - loss: 3.8349 - accuracy:
38/96 [=======>...] - ETA: Os - loss: 3.7867 - accuracy:
0.4875
43/96 [========>...] - ETA: Os - loss: 3.7529 - accuracy:
0.4983
0.4947
0.5008
```

```
0.4996
0.5054
0.5067
0.5072
0.5086
0.5108
0.5117 - val_loss: 4.0246 - val_accuracy: 0.2955
Epoch 3/30
1/96 [...] - ETA: 1s - loss: 2.8498 - accuracy:
0.5238
7/96 [=>...] - ETA: Os - loss: 2.9425 - accuracy:
11/96 [==>...] - ETA: Os - loss: 2.9728 - accuracy:
0.5671
16/96 [====>...] - ETA: Os - loss: 2.8669 - accuracy:
0.6101
21/96 [====>...] - ETA: Os - loss: 2.8869 - accuracy:
0.5964
27/96 [======>...] - ETA: Os - loss: 2.8940 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 2.8710 - accuracy:
0.5859
38/96 [=======>...] - ETA: Os - loss: 2.8213 - accuracy:
0.5965
0.5969
0.6012
0.6002
0.6059
0.6062
0.6089
```

```
0.6114
0.6118
0.6154
96/96 [============= ] - ETA: Os - loss: 2.6307 - accuracy:
0.6147
0.6147 - val_loss: 3.4057 - val_accuracy: 0.3390
Epoch 4/30
1/96 [...] - ETA: 1s - loss: 2.0182 - accuracy:
6/96 [>...] - ETA: Os - loss: 2.1415 - accuracy:
0.6905
11/96 [==>...] - ETA: Os - loss: 2.1243 - accuracy:
0.6970
16/96 [====>...] - ETA: Os - loss: 2.1150 - accuracy:
0.6875
21/96 [====>...] - ETA: Os - loss: 2.1773 - accuracy:
26/96 [======>...] - ETA: Os - loss: 2.1653 - accuracy:
0.6777
32/96 [=======>...] - ETA: Os - loss: 2.1384 - accuracy:
0.6845
36/96 [=======>...] - ETA: Os - loss: 2.1403 - accuracy:
42/96 [========>...] - ETA: Os - loss: 2.1381 - accuracy:
0.6859
0.6885
0.6846
0.6786
0.6791
0.6781
0.6821
0.6801
```

```
0.6777
0.6822
0.6822 - val_loss: 2.5908 - val_accuracy: 0.4842
Epoch 5/30
1/96 [...] - ETA: 1s - loss: 1.9411 - accuracy:
0.7143
6/96 [>...] - ETA: 1s - loss: 1.7209 - accuracy:
0.7381
12/96 [==>...] - ETA: Os - loss: 1.6976 - accuracy:
18/96 [====>...] - ETA: Os - loss: 1.6771 - accuracy:
0.7566
22/96 [====>...] - ETA: Os - loss: 1.6879 - accuracy:
0.7468
28/96 [======>...] - ETA: Os - loss: 1.6941 - accuracy:
0.7449
33/96 [=======>...] - ETA: Os - loss: 1.6744 - accuracy:
0.7460
38/96 [=======>...] - ETA: Os - loss: 1.6633 - accuracy:
0.7506
44/96 [========>...] - ETA: Os - loss: 1.6679 - accuracy:
0.7478
0.7486
0.7356
0.7388
0.7396
0.7381
0.7408
0.7424
0.7421 - val_loss: 2.1429 - val_accuracy: 0.5482
```

```
1/96 [...] - ETA: 1s - loss: 1.2143 - accuracy:
0.8571
6/96 [>...] - ETA: Os - loss: 1.2797 - accuracy:
0.8413
12/96 [==>...] - ETA: Os - loss: 1.3830 - accuracy:
0.7897
18/96 [====>...] - ETA: Os - loss: 1.3280 - accuracy:
0.8122
24/96 [=====>...] - ETA: Os - loss: 1.3656 - accuracy:
0.7937
30/96 [======>...] - ETA: Os - loss: 1.3528 - accuracy:
0.7873
34/96 [=======>...] - ETA: Os - loss: 1.3766 - accuracy:
40/96 [========>...] - ETA: Os - loss: 1.3728 - accuracy:
0.7810
0.7733
0.7759
0.7734
0.7741
0.7732
0.7698
0.7704
0.7701 - val_loss: 2.3105 - val_accuracy: 0.4859
Epoch 7/30
1/96 [...] - ETA: 1s - loss: 1.1539 - accuracy:
0.7619
7/96 [=>...] - ETA: Os - loss: 1.2168 - accuracy:
13/96 [===>...] - ETA: Os - loss: 1.1973 - accuracy:
0.8095
```

```
19/96 [====>...] - ETA: Os - loss: 1.1868 - accuracy:
0.8095
25/96 [=====>...] - ETA: Os - loss: 1.1350 - accuracy:
30/96 [=======>...] - ETA: Os - loss: 1.1523 - accuracy:
0.8317
36/96 [=======>...] - ETA: 0s - loss: 1.1190 - accuracy:
0.8439
42/96 [========>...] - ETA: Os - loss: 1.1251 - accuracy:
0.8311
0.8274
0.8313
0.8259
0.8233
0.8250
0.8228
0.8200
96/96 [============== ] - ETA: Os - loss: 1.1317 - accuracy:
0.8206
0.8206 - val_loss: 2.0021 - val_accuracy: 0.5611
Epoch 8/30
1/96 [...] - ETA: 1s - loss: 0.7862 - accuracy:
0.9048
7/96 [=>...] - ETA: Os - loss: 0.9652 - accuracy:
0.8707
13/96 [===>...] - ETA: Os - loss: 0.9594 - accuracy:
0.8571
19/96 [====>...] - ETA: Os - loss: 0.9584 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 0.9434 - accuracy:
28/96 [======>...] - ETA: Os - loss: 0.9324 - accuracy:
0.8622
34/96 [=======>...] - ETA: Os - loss: 0.9434 - accuracy:
```

```
0.8557
40/96 [========>...] - ETA: Os - loss: 0.9454 - accuracy:
0.8536
46/96 [========>...] - ETA: Os - loss: 0.9297 - accuracy:
0.8592
0.8480
0.8440
0.8451
0.8435
0.8409
0.8366 - val_loss: 1.9334 - val_accuracy: 0.5542
Epoch 9/30
1/96 [...] - ETA: 2s - loss: 1.2065 - accuracy:
0.8095
7/96 [=>...] - ETA: Os - loss: 0.9582 - accuracy:
0.8435
13/96 [===>...] - ETA: Os - loss: 0.9201 - accuracy:
0.8498
19/96 [====>...] - ETA: Os - loss: 0.9208 - accuracy:
0.8596
25/96 [=====>...] - ETA: Os - loss: 0.8955 - accuracy:
0.8648
29/96 [======>...] - ETA: Os - loss: 0.8989 - accuracy:
0.8654
35/96 [=======>...] - ETA: Os - loss: 0.9113 - accuracy:
0.8558
41/96 [========>...] - ETA: Os - loss: 0.9177 - accuracy:
0.8525
46/96 [========>...] - ETA: Os - loss: 0.9347 - accuracy:
0.8487
```

```
0.8485
0.8428
0.8375
0.8360
0.8383
0.8362
0.8360
0.8375
0.8386 - val_loss: 1.8184 - val_accuracy: 0.5927
Epoch 10/30
1/96 [...] - ETA: 2s - loss: 0.7198 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 0.7037 - accuracy:
0.9238
9/96 [=>...] - ETA: 1s - loss: 0.8165 - accuracy:
0.8889
14/96 [===>...] - ETA: 1s - loss: 0.8116 - accuracy:
0.8844
18/96 [====>...] - ETA: Os - loss: 0.8382 - accuracy:
0.8730
23/96 [=====>...] - ETA: Os - loss: 0.8252 - accuracy:
27/96 [======>...] - ETA: Os - loss: 0.7961 - accuracy:
0.8801
32/96 [=======>...] - ETA: Os - loss: 0.7938 - accuracy:
0.8839
36/96 [========>...] - ETA: Os - loss: 0.7951 - accuracy:
40/96 [=======>...] - ETA: Os - loss: 0.8081 - accuracy:
0.8794
```

```
0.8805
0.8832
0.8867
0.8842
0.8856
0.8862
0.8793
0.8794
96/96 [============= ] - ETA: Os - loss: 0.7746 - accuracy:
0.8776
0.8776 - val_loss: 1.8252 - val_accuracy: 0.5952
Epoch 11/30
1/96 [...] - ETA: 1s - loss: 0.6751 - accuracy:
0.9048
5/96 [>...] - ETA: 1s - loss: 0.6812 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 0.6799 - accuracy:
0.9238
14/96 [===>...] - ETA: 1s - loss: 0.6731 - accuracy:
0.9184
18/96 [====>...] - ETA: 1s - loss: 0.7130 - accuracy:
0.9021
22/96 [====>...] - ETA: 1s - loss: 0.6985 - accuracy:
0.9069
26/96 [======>...] - ETA: Os - loss: 0.7278 - accuracy:
0.8938
30/96 [=======>...] - ETA: Os - loss: 0.7489 - accuracy:
0.8889
34/96 [=======>...] - ETA: Os - loss: 0.7484 - accuracy:
38/96 [========>...] - ETA: Os - loss: 0.7414 - accuracy:
42/96 [========>...] - ETA: Os - loss: 0.7367 - accuracy:
0.8912
```

```
0.8944
0.8915
0.8875
0.8886
0.8891
0.8871
0.8855
0.8831
0.8815
0.8801 - val_loss: 1.7208 - val_accuracy: 0.5909
Epoch 12/30
1/96 [...] - ETA: 1s - loss: 0.8419 - accuracy:
6/96 [>...] - ETA: Os - loss: 0.7374 - accuracy:
0.8571
11/96 [==>...] - ETA: Os - loss: 0.7090 - accuracy:
0.8658
16/96 [====>...] - ETA: Os - loss: 0.6940 - accuracy:
0.8780
20/96 [====>...] - ETA: Os - loss: 0.6577 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 0.6684 - accuracy:
0.8889
29/96 [======>...] - ETA: Os - loss: 0.6781 - accuracy:
0.8900
35/96 [=======>...] - ETA: Os - loss: 0.6737 - accuracy:
40/96 [========>...] - ETA: Os - loss: 0.6612 - accuracy:
0.8869
0.8799
```

```
0.8819
0.8840
0.8881
0.8867
0.8896
0.8873
0.8929
0.8906 - val_loss: 1.9137 - val_accuracy: 0.5696
Epoch 13/30
1/96 [...] - ETA: 1s - loss: 0.5234 - accuracy:
0.9524
7/96 [=>...] - ETA: Os - loss: 0.5724 - accuracy:
0.9116
13/96 [===>...] - ETA: Os - loss: 0.5633 - accuracy:
0.9194
19/96 [====>...] - ETA: Os - loss: 0.5583 - accuracy:
23/96 [=====>...] - ETA: Os - loss: 0.5466 - accuracy:
0.9255
29/96 [======>...] - ETA: Os - loss: 0.5355 - accuracy:
0.9294
33/96 [=======>...] - ETA: Os - loss: 0.5413 - accuracy:
38/96 [=======>...] - ETA: 0s - loss: 0.5540 - accuracy:
0.9185
43/96 [========>...] - ETA: Os - loss: 0.5736 - accuracy:
0.9125
47/96 [========>...] - ETA: Os - loss: 0.5764 - accuracy:
0.9119
0.9093
58/96 [=========>...] - ETA: Os - loss: 0.5821 - accuracy:
0.9055
```

```
0.9034
0.9030
0.9048
0.9037
96/96 [============= ] - ETA: Os - loss: 0.6129 - accuracy:
0.9005
0.9005 - val_loss: 1.8118 - val_accuracy: 0.6191
Epoch 14/30
1/96 [...] - ETA: 1s - loss: 0.6222 - accuracy:
0.9048
7/96 [=>...] - ETA: 0s - loss: 0.6270 - accuracy:
0.8844
13/96 [===>...] - ETA: Os - loss: 0.5937 - accuracy:
0.8901
19/96 [====>...] - ETA: Os - loss: 0.5683 - accuracy:
0.9048
24/96 [=====>...] - ETA: Os - loss: 0.5782 - accuracy:
0.9087
29/96 [======>...] - ETA: Os - loss: 0.5715 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 0.5659 - accuracy:
0.9129
40/96 [========>...] - ETA: Os - loss: 0.5763 - accuracy:
0.9048
45/96 [========>...] - ETA: Os - loss: 0.5903 - accuracy:
0.9016
0.8968
0.8932
0.8924
```

```
0.8936
0.8942
0.8949
0.8921 - val_loss: 1.9190 - val_accuracy: 0.5713
Epoch 15/30
1/96 [...] - ETA: 1s - loss: 0.5740 - accuracy:
0.9524
7/96 [=>...] - ETA: Os - loss: 0.7550 - accuracy:
0.8503
11/96 [==>...] - ETA: Os - loss: 0.6850 - accuracy:
0.8745
16/96 [====>...] - ETA: Os - loss: 0.6614 - accuracy:
0.8780
21/96 [====>...] - ETA: Os - loss: 0.6739 - accuracy:
0.8753
27/96 [======>...] - ETA: Os - loss: 0.6591 - accuracy:
0.8801
31/96 [======>...] - ETA: Os - loss: 0.6577 - accuracy:
0.8786
36/96 [========>...] - ETA: Os - loss: 0.6693 - accuracy:
0.8770
41/96 [========>...] - ETA: Os - loss: 0.6582 - accuracy:
0.8839
0.8753
0.8746
0.8755
0.8789
0.8775
0.8734
```

```
0.8721 - val_loss: 1.9922 - val_accuracy: 0.5525
Epoch 16/30
1/96 [...] - ETA: 1s - loss: 0.5023 - accuracy:
5/96 [>...] - ETA: 1s - loss: 0.5606 - accuracy:
0.9238
10/96 [==>...] - ETA: 1s - loss: 0.5939 - accuracy:
0.9048
16/96 [====>...] - ETA: Os - loss: 0.5892 - accuracy:
0.9167
21/96 [====>...] - ETA: Os - loss: 0.5887 - accuracy:
0.9093
25/96 [=====>...] - ETA: Os - loss: 0.6033 - accuracy:
0.9048
31/96 [======>...] - ETA: Os - loss: 0.6101 - accuracy:
0.9048
37/96 [=======>...] - ETA: Os - loss: 0.5940 - accuracy:
0.9112
43/96 [========>...] - ETA: Os - loss: 0.5882 - accuracy:
0.9155
0.9146
0.9174
0.9163
0.9162
0.9112
0.9119
0.9120 - val_loss: 1.7298 - val_accuracy: 0.6277
```

Epoch 17/30

```
1/96 [...] - ETA: 2s - loss: 0.6376 - accuracy:
0.8571
6/96 [>...] - ETA: Os - loss: 0.4810 - accuracy:
0.9603
12/96 [==>...] - ETA: Os - loss: 0.4913 - accuracy:
0.9563
17/96 [====>...] - ETA: Os - loss: 0.4763 - accuracy:
0.9636
22/96 [====>...] - ETA: Os - loss: 0.4766 - accuracy:
0.9610
26/96 [======>...] - ETA: Os - loss: 0.4758 - accuracy:
0.9615
32/96 [=======>...] - ETA: Os - loss: 0.4606 - accuracy:
38/96 [=======>...] - ETA: Os - loss: 0.4494 - accuracy:
0.9568
0.9524
0.9515
0.9457
0.9424
0.9384
0.9346
0.9334
0.9341
0.9345 - val_loss: 1.9703 - val_accuracy: 0.6208
Epoch 18/30
1/96 [...] - ETA: 2s - loss: 0.3289 - accuracy:
1.0000
6/96 [>...] - ETA: Os - loss: 0.4825 - accuracy:
0.9603
11/96 [==>...] - ETA: Os - loss: 0.4962 - accuracy:
```

```
0.9481
17/96 [====>...] - ETA: 0s - loss: 0.5020 - accuracy:
0.9468
22/96 [====>...] - ETA: Os - loss: 0.4908 - accuracy:
0.9459
26/96 [======>...] - ETA: Os - loss: 0.4953 - accuracy:
31/96 [======>...] - ETA: Os - loss: 0.4888 - accuracy:
0.9478
37/96 [========>...] - ETA: Os - loss: 0.4919 - accuracy:
0.9434
42/96 [========>...] - ETA: Os - loss: 0.4788 - accuracy:
0.9478
0.9489
0.9491
0.9479
0.9458
0.9398
0.9390 - val loss: 2.0710 - val accuracy: 0.5961
Epoch 19/30
1/96 [...] - ETA: 1s - loss: 0.5000 - accuracy:
0.9524
6/96 [>...] - ETA: 1s - loss: 0.4511 - accuracy:
0.9603
10/96 [==>...] - ETA: 1s - loss: 0.4638 - accuracy:
0.9476
14/96 [===>...] - ETA: 1s - loss: 0.4837 - accuracy:
18/96 [====>...] - ETA: Os - loss: 0.4734 - accuracy:
0.9365
```

```
22/96 [====>...] - ETA: Os - loss: 0.4827 - accuracy:
0.9351
26/96 [======>...] - ETA: Os - loss: 0.4790 - accuracy:
0.9377
30/96 [======>...] - ETA: Os - loss: 0.4809 - accuracy:
0.9349
35/96 [=======>...] - ETA: Os - loss: 0.4811 - accuracy:
0.9347
39/96 [========>...] - ETA: Os - loss: 0.4817 - accuracy:
0.9341
0.9336
47/96 [========>...] - ETA: Os - loss: 0.4885 - accuracy:
0.9323
0.9332
0.9329
0.9299
0.9279
0.9280
0.9262
0.9263
0.9265 - val_loss: 2.2664 - val_accuracy: 0.5816
Epoch 20/30
1/96 [...] - ETA: 1s - loss: 0.4451 - accuracy:
0.9524
6/96 [>...] - ETA: 1s - loss: 0.5143 - accuracy:
0.9127
11/96 [==>...] - ETA: Os - loss: 0.4856 - accuracy:
0.9264
15/96 [===>...] - ETA: Os - loss: 0.4765 - accuracy:
```

```
0.9429
19/96 [====>...] - ETA: Os - loss: 0.4673 - accuracy:
0.9499
23/96 [=====>...] - ETA: Os - loss: 0.4656 - accuracy:
0.9503
27/96 [======>...] - ETA: Os - loss: 0.4728 - accuracy:
31/96 [======>...] - ETA: Os - loss: 0.4670 - accuracy:
0.9493
35/96 [=======>...] - ETA: Os - loss: 0.4654 - accuracy:
0.9497
39/96 [========>...] - ETA: Os - loss: 0.4753 - accuracy:
0.9463
44/96 [========>...] - ETA: Os - loss: 0.4782 - accuracy:
0.9402
0.9421
0.9411
0.9327
0.9346
0.9327
0.9331
0.9334
0.9332
0.9293
0.9290 - val_loss: 2.2257 - val_accuracy: 0.5628
Epoch 21/30
1/96 [...] - ETA: 2s - loss: 0.4818 - accuracy:
0.9048
```

```
5/96 [>...] - ETA: 1s - loss: 0.4315 - accuracy:
0.9429
9/96 [=>...] - ETA: 1s - loss: 0.5165 - accuracy:
0.9259
13/96 [===>...] - ETA: 1s - loss: 0.5315 - accuracy:
0.9231
17/96 [====>...] - ETA: 1s - loss: 0.5301 - accuracy:
0.9300
21/96 [====>...] - ETA: 1s - loss: 0.5474 - accuracy:
0.9297
25/96 [=====>...] - ETA: Os - loss: 0.5547 - accuracy:
0.9276
30/96 [======>...] - ETA: Os - loss: 0.5622 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 0.5716 - accuracy:
40/96 [========>...] - ETA: Os - loss: 0.5703 - accuracy:
0.9202
45/96 [========>...] - ETA: Os - loss: 0.5708 - accuracy:
0.9185
0.9190
0.9167
0.9140
0.9118
0.9123
0.9097
0.9078
0.9075 - val_loss: 2.0852 - val_accuracy: 0.5918
Epoch 1/30
1/96 [...] - ETA: 6:06 - loss: 10.8904 - accuracy:
0.2381
4/96 [>...] - ETA: 1s - loss: 12.8052 - accuracy:
0.2619
7/96 [=>...] - ETA: 1s - loss: 13.4801 - accuracy:
```

```
0.2177
10/96 [==>...] - ETA: 1s - loss: 13.4164 - accuracy:
0.1905
14/96 [===>...] - ETA: 1s - loss: 12.7223 - accuracy:
0.1871
18/96 [====>...] - ETA: 1s - loss: 11.9247 - accuracy:
22/96 [====>...] - ETA: 1s - loss: 11.3585 - accuracy:
0.2078
26/96 [======>...] - ETA: 1s - loss: 11.0350 - accuracy:
0.2198
30/96 [======>...] - ETA: 1s - loss: 10.7226 - accuracy:
0.2317
33/96 [=======>...] - ETA: 1s - loss: 10.5340 - accuracy:
37/96 [=======>...] - ETA: Os - loss: 10.3060 - accuracy:
0.2214
42/96 [========>...] - ETA: Os - loss: 10.0045 - accuracy:
0.2290
0.2519
0.2526
0.2650
0.2701
0.2777
0.2915
0.2985
0.2999 - val_loss: 8.9712 - val_accuracy: 0.1187
```

Epoch 2/30

```
1/96 [...] - ETA: 2s - loss: 5.6964 - accuracy:
0.3810
5/96 [>...] - ETA: 1s - loss: 5.8896 - accuracy:
0.4000
10/96 [==>...] - ETA: 1s - loss: 5.8400 - accuracy:
0.4095
15/96 [===>...] - ETA: Os - loss: 5.7342 - accuracy:
0.4317
19/96 [====>...] - ETA: Os - loss: 5.6720 - accuracy:
0.4386
23/96 [=====>...] - ETA: Os - loss: 5.6309 - accuracy:
0.4493
28/96 [======>...] - ETA: Os - loss: 5.5648 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 5.5210 - accuracy:
38/96 [======>...] - ETA: Os - loss: 5.4771 - accuracy:
0.4599
43/96 [========>...] - ETA: Os - loss: 5.4133 - accuracy:
0.4640
0.4661
0.4641
0.4592
0.4637
0.4724
0.4762
0.4824
0.4901
0.4985
0.5012 - val_loss: 4.4243 - val_accuracy: 0.3877
Epoch 3/30
1/96 [...] - ETA: 1s - loss: 3.6524 - accuracy:
```

```
0.5714
6/96 [>...] - ETA: Os - loss: 3.7179 - accuracy:
0.6032
10/96 [==>...] - ETA: 1s - loss: 3.7179 - accuracy:
0.6048
15/96 [===>...] - ETA: Os - loss: 3.6452 - accuracy:
19/96 [====>...] - ETA: Os - loss: 3.5888 - accuracy:
0.6216
24/96 [=====>...] - ETA: Os - loss: 3.5862 - accuracy:
0.6230
29/96 [======>...] - ETA: Os - loss: 3.5849 - accuracy:
0.6158
34/96 [=======>...] - ETA: Os - loss: 3.5674 - accuracy:
38/96 [=======>...] - ETA: Os - loss: 3.5212 - accuracy:
0.6341
42/96 [========>...] - ETA: Os - loss: 3.4870 - accuracy:
0.6383
0.6460
0.6616
0.6674
0.6763
0.6823
0.6836
0.6890
0.6907
0.6952
96/96 [============= ] - ETA: Os - loss: 3.0899 - accuracy:
0.7006
```

```
0.7006 - val_loss: 2.9366 - val_accuracy: 0.6405
Epoch 4/30
1/96 [...] - ETA: 2s - loss: 2.4761 - accuracy:
0.7619
5/96 [>...] - ETA: 1s - loss: 2.4326 - accuracy:
0.7619
9/96 [=>...] - ETA: 1s - loss: 2.4485 - accuracy:
0.7513
13/96 [===>...] - ETA: 1s - loss: 2.3957 - accuracy:
0.7729
17/96 [====>...] - ETA: 1s - loss: 2.3648 - accuracy:
0.7815
21/96 [====>...] - ETA: 1s - loss: 2.3373 - accuracy:
0.7937
24/96 [=====>...] - ETA: 1s - loss: 2.3018 - accuracy:
0.8056
27/96 [======>...] - ETA: 1s - loss: 2.2898 - accuracy:
0.8060
31/96 [=======>...] - ETA: Os - loss: 2.2800 - accuracy:
0.8034
34/96 [=======>...] - ETA: Os - loss: 2.2869 - accuracy:
0.7955
38/96 [======>...] - ETA: Os - loss: 2.2717 - accuracy:
0.7907
42/96 [========>...] - ETA: Os - loss: 2.2738 - accuracy:
0.7834
0.7866
0.7900
0.7881
0.7808
0.7857
0.7823
0.7857
```

```
0.7863
0.7833
0.7823
0.7794
0.7796 - val_loss: 3.0934 - val_accuracy: 0.4466
Epoch 5/30
1/96 [...] - ETA: 1s - loss: 1.6899 - accuracy:
0.8095
5/96 [>...] - ETA: 1s - loss: 1.7613 - accuracy:
0.8190
9/96 [=>...] - ETA: 1s - loss: 1.6771 - accuracy:
0.8519
13/96 [===>...] - ETA: 1s - loss: 1.6750 - accuracy:
16/96 [====>...] - ETA: 1s - loss: 1.6725 - accuracy:
0.8542
20/96 [====>...] - ETA: 1s - loss: 1.6866 - accuracy:
0.8476
24/96 [=====>...] - ETA: 1s - loss: 1.6787 - accuracy:
28/96 [======>...] - ETA: Os - loss: 1.6925 - accuracy:
0.8435
31/96 [======>...] - ETA: Os - loss: 1.6987 - accuracy:
0.8372
35/96 [=======>...] - ETA: Os - loss: 1.6963 - accuracy:
0.8327
39/96 [========>...] - ETA: Os - loss: 1.6778 - accuracy:
43/96 [========>...] - ETA: Os - loss: 1.6731 - accuracy:
0.8328
47/96 [=========>...] - ETA: Os - loss: 1.6488 - accuracy:
0.8369
0.8361
0.8350
0.8345
```

```
0.8337
0.8344
0.8317
0.8262
0.8256
0.8236 - val_loss: 1.9502 - val_accuracy: 0.6644
Epoch 6/30
1/96 [...] - ETA: 2s - loss: 1.3092 - accuracy:
0.8571
5/96 [>...] - ETA: 1s - loss: 1.3119 - accuracy:
0.8667
10/96 [==>...] - ETA: 1s - loss: 1.3128 - accuracy:
0.8667
15/96 [===>...] - ETA: Os - loss: 1.3024 - accuracy:
0.8730
18/96 [====>...] - ETA: Os - loss: 1.2922 - accuracy:
0.8624
23/96 [=====>...] - ETA: Os - loss: 1.2606 - accuracy:
27/96 [======>...] - ETA: Os - loss: 1.2382 - accuracy:
0.8765
32/96 [======>...] - ETA: Os - loss: 1.2034 - accuracy:
0.8839
37/96 [=======>...] - ETA: 0s - loss: 1.1927 - accuracy:
0.8829
0.8827
45/96 [========>...] - ETA: Os - loss: 1.1872 - accuracy:
0.8815
0.8800
55/96 [=========>...] - ETA: Os - loss: 1.1781 - accuracy:
60/96 [=========>...] - ETA: Os - loss: 1.1750 - accuracy:
0.8711
```

```
0.8709
0.8695
0.8701
0.8659
0.8648
0.8665
96/96 [============= ] - ETA: Os - loss: 1.1733 - accuracy:
0.8656
0.8656 - val_loss: 1.8306 - val_accuracy: 0.6533
Epoch 7/30
1/96 [...] - ETA: 2s - loss: 1.0025 - accuracy:
0.9048
6/96 [>...] - ETA: 1s - loss: 0.9001 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 0.9533 - accuracy:
0.8905
13/96 [===>...] - ETA: 1s - loss: 0.9286 - accuracy:
0.9011
16/96 [====>...] - ETA: 1s - loss: 0.9617 - accuracy:
0.8988
20/96 [====>...] - ETA: 1s - loss: 0.9166 - accuracy:
0.9119
24/96 [=====>...] - ETA: Os - loss: 0.9068 - accuracy:
0.9127
29/96 [======>...] - ETA: Os - loss: 0.9240 - accuracy:
0.9031
33/96 [=======>...] - ETA: Os - loss: 0.9388 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.9387 - accuracy:
41/96 [========>...] - ETA: Os - loss: 0.9514 - accuracy:
0.8908
0.8889
0.8895
```

```
0.8881
0.8852
0.8872
0.8883
0.8853
0.8866
96/96 [============ ] - 2s 17ms/step - loss: 0.9318 - accuracy:
0.8846 - val_loss: 1.6626 - val_accuracy: 0.7011
Epoch 8/30
1/96 [...] - ETA: 2s - loss: 0.6229 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.7422 - accuracy:
0.9333
9/96 [=>...] - ETA: 1s - loss: 0.7176 - accuracy:
0.9418
13/96 [===>...] - ETA: 1s - loss: 0.7818 - accuracy:
16/96 [====>...] - ETA: 1s - loss: 0.7793 - accuracy:
0.9137
19/96 [====>...] - ETA: 1s - loss: 0.7620 - accuracy:
0.9173
23/96 [=====>...] - ETA: 1s - loss: 0.7600 - accuracy:
0.9151
27/96 [======>...] - ETA: Os - loss: 0.7491 - accuracy:
0.9153
32/96 [=======>...] - ETA: Os - loss: 0.7673 - accuracy:
0.9018
36/96 [========>...] - ETA: Os - loss: 0.7630 - accuracy:
0.8995
41/96 [========>...] - ETA: Os - loss: 0.7519 - accuracy:
0.9036
45/96 [========>...] - ETA: Os - loss: 0.7528 - accuracy:
0.9086
```

```
0.9118
0.9072
0.9055
0.9089
0.9080
0.9060
0.8985
0.8952
0.8951 - val_loss: 1.5942 - val_accuracy: 0.6892
Epoch 9/30
1/96 [...] - ETA: 1s - loss: 0.4439 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.5487 - accuracy:
0.9524
9/96 [=>...] - ETA: 1s - loss: 0.6366 - accuracy:
0.9153
13/96 [===>...] - ETA: 1s - loss: 0.6404 - accuracy:
0.9194
18/96 [====>...] - ETA: 1s - loss: 0.6634 - accuracy:
0.9101
23/96 [=====>...] - ETA: Os - loss: 0.7165 - accuracy:
28/96 [======>...] - ETA: Os - loss: 0.7138 - accuracy:
0.8980
32/96 [=======>...] - ETA: Os - loss: 0.7156 - accuracy:
0.9018
36/96 [========>...] - ETA: Os - loss: 0.7226 - accuracy:
41/96 [=======>...] - ETA: Os - loss: 0.7340 - accuracy:
0.8944
```

```
0.8936
0.8909
0.8918
0.8927
0.8940
0.8966
0.8991
0.9004
0.9012
0.9020 - val_loss: 1.5215 - val_accuracy: 0.6900
Epoch 10/30
1/96 [...] - ETA: 1s - loss: 0.4108 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.4541 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 0.4743 - accuracy:
0.9524
14/96 [===>...] - ETA: 1s - loss: 0.4999 - accuracy:
0.9354
18/96 [====>...] - ETA: 1s - loss: 0.4953 - accuracy:
0.9392
22/96 [====>...] - ETA: Os - loss: 0.4987 - accuracy:
0.9351
26/96 [======>...] - ETA: Os - loss: 0.5112 - accuracy:
0.9304
30/96 [=======>...] - ETA: Os - loss: 0.5084 - accuracy:
0.9333
35/96 [=======>...] - ETA: Os - loss: 0.5297 - accuracy:
0.9306
39/96 [=======>...] - ETA: Os - loss: 0.5330 - accuracy:
42/96 [========>...] - ETA: Os - loss: 0.5416 - accuracy:
0.9240
```

```
0.9203
0.9226
0.9246
0.9241
0.9216
0.9243
0.9250
0.9267
0.9278
0.9280 - val_loss: 1.3308 - val_accuracy: 0.7199
Epoch 11/30
1/96 [...] - ETA: 2s - loss: 0.3728 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.4254 - accuracy:
0.9619
9/96 [=>...] - ETA: 1s - loss: 0.4640 - accuracy:
0.9418
13/96 [===>...] - ETA: 1s - loss: 0.4606 - accuracy:
17/96 [====>...] - ETA: 1s - loss: 0.4633 - accuracy:
0.9384
22/96 [====>...] - ETA: Os - loss: 0.4637 - accuracy:
0.9351
27/96 [======>...] - ETA: Os - loss: 0.4559 - accuracy:
31/96 [======>...] - ETA: Os - loss: 0.4721 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 0.4713 - accuracy:
0.9347
40/96 [========>...] - ETA: Os - loss: 0.4679 - accuracy:
```

```
0.9381
45/96 [========>...] - ETA: Os - loss: 0.4805 - accuracy:
0.9344
0.9362
0.9351
0.9350
0.9342
0.9351
0.9359
0.9350
0.9357
0.9350 - val_loss: 1.5149 - val_accuracy: 0.6985
Epoch 12/30
1/96 [...] - ETA: 2s - loss: 0.3430 - accuracy:
0.9524
4/96 [>...] - ETA: 1s - loss: 0.3962 - accuracy:
0.9643
8/96 [=>...] - ETA: 1s - loss: 0.4072 - accuracy:
0.9524
12/96 [==>...] - ETA: 1s - loss: 0.4372 - accuracy:
0.9444
16/96 [====>...] - ETA: 1s - loss: 0.4269 - accuracy:
0.9464
20/96 [====>...] - ETA: 1s - loss: 0.4329 - accuracy:
0.9429
23/96 [=====>...] - ETA: 1s - loss: 0.4324 - accuracy:
26/96 [======>...] - ETA: 1s - loss: 0.4437 - accuracy:
0.9396
```

```
29/96 [======>...] - ETA: 1s - loss: 0.4533 - accuracy:
0.9343
32/96 [=======>...] - ETA: 1s - loss: 0.4466 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.4492 - accuracy:
0.9352
0.9369
44/96 [========>...] - ETA: Os - loss: 0.4625 - accuracy:
0.9297
0.9335
0.9310
0.9301
0.9272
0.9286
0.9289
0.9298
0.9286
96/96 [============= ] - ETA: Os - loss: 0.4679 - accuracy:
0.9280
0.9280 - val_loss: 1.8257 - val_accuracy: 0.5824
Epoch 13/30
1/96 [...] - ETA: 2s - loss: 0.4474 - accuracy:
0.9048
4/96 [>...] - ETA: 1s - loss: 0.4128 - accuracy:
0.9524
7/96 [=>...] - ETA: 1s - loss: 0.4334 - accuracy:
```

```
0.9456
11/96 [==>...] - ETA: 1s - loss: 0.4658 - accuracy:
0.9307
15/96 [===>...] - ETA: 1s - loss: 0.4646 - accuracy:
0.9333
19/96 [====>...] - ETA: 1s - loss: 0.4572 - accuracy:
23/96 [=====>...] - ETA: 1s - loss: 0.4491 - accuracy:
0.9400
27/96 [======>...] - ETA: 1s - loss: 0.4649 - accuracy:
0.9330
30/96 [======>...] - ETA: 1s - loss: 0.4667 - accuracy:
0.9302
33/96 [=======>...] - ETA: 1s - loss: 0.4645 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.4671 - accuracy:
0.9286
39/96 [========>...] - ETA: Os - loss: 0.4643 - accuracy:
0.9267
0.9274
45/96 [=========>...] - ETA: Os - loss: 0.4604 - accuracy:
0.9271
0.9295
0.9317
0.9311
0.9333
0.9320
0.9314
0.9310 - val_loss: 1.7152 - val_accuracy: 0.6268
```

```
Epoch 14/30
```

```
1/96 [...] - ETA: 1s - loss: 0.2420 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.3995 - accuracy:
0.9524
10/96 [==>...] - ETA: 1s - loss: 0.4133 - accuracy:
0.9476
14/96 [===>...] - ETA: 1s - loss: 0.4190 - accuracy:
0.9456
19/96 [====>...] - ETA: Os - loss: 0.3923 - accuracy:
0.9474
23/96 [=====>...] - ETA: Os - loss: 0.3803 - accuracy:
27/96 [======>...] - ETA: Os - loss: 0.3799 - accuracy:
0.9489
32/96 [======>...] - ETA: Os - loss: 0.3803 - accuracy:
0.9479
36/96 [========>...] - ETA: Os - loss: 0.3701 - accuracy:
0.9511
41/96 [=======>...] - ETA: Os - loss: 0.3832 - accuracy:
0.9477
46/96 [=========>...] - ETA: Os - loss: 0.3858 - accuracy:
0.9472
0.9458
0.9455
0.9460
0.9477
0.9474
0.9482
0.9473
0.9460
96/96 [============= ] - 2s 16ms/step - loss: 0.3799 - accuracy:
0.9470 - val_loss: 1.6646 - val_accuracy: 0.6225
```

Epoch 15/30

```
1/96 [...] - ETA: 2s - loss: 0.4177 - accuracy:
0.9524
5/96 [>...] - ETA: 1s - loss: 0.4298 - accuracy:
0.9619
9/96 [=>...] - ETA: 1s - loss: 0.3818 - accuracy:
0.9683
12/96 [==>...] - ETA: 1s - loss: 0.3853 - accuracy:
0.9603
15/96 [===>...] - ETA: 1s - loss: 0.3836 - accuracy:
0.9524
19/96 [====>...] - ETA: 1s - loss: 0.3937 - accuracy:
0.9449
23/96 [=====>...] - ETA: 1s - loss: 0.3860 - accuracy:
0.9420
27/96 [======>...] - ETA: 1s - loss: 0.3695 - accuracy:
0.9471
32/96 [=======>...] - ETA: Os - loss: 0.3584 - accuracy:
0.9509
37/96 [=======>...] - ETA: 0s - loss: 0.3515 - accuracy:
40/96 [=======>...] - ETA: Os - loss: 0.3544 - accuracy:
0.9524
0.9502
0.9496
0.9515
0.9539
0.9531
0.9555
0.9559
0.9568
0.9577
```

```
0.9549
0.9545 - val_loss: 1.7543 - val_accuracy: 0.6149
Epoch 1/30
1/96 [...] - ETA: 7:54 - loss: 14.5510 - accuracy:
0.0476
4/96 [>...] - ETA: 1s - loss: 15.3528 - accuracy:
0.1071
7/96 [=>...] - ETA: 1s - loss: 14.3073 - accuracy:
0.1293
10/96 [==>...] - ETA: 1s - loss: 13.1470 - accuracy:
13/96 [===>...] - ETA: 1s - loss: 12.2725 - accuracy:
0.2051
16/96 [====>...] - ETA: 1s - loss: 11.7620 - accuracy:
0.2143
18/96 [====>...] - ETA: 1s - loss: 11.4904 - accuracy:
0.2249
21/96 [====>...] - ETA: 1s - loss: 11.2041 - accuracy:
0.2177
24/96 [=====>...] - ETA: 1s - loss: 10.9075 - accuracy:
0.2163
27/96 [======>...] - ETA: 1s - loss: 10.6423 - accuracy:
0.2134
30/96 [======>...] - ETA: 1s - loss: 10.3778 - accuracy:
0.2317
33/96 [=======>...] - ETA: 1s - loss: 10.1663 - accuracy:
36/96 [=======>...] - ETA: 1s - loss: 9.9781 - accuracy:
39/96 [========>...] - ETA: 1s - loss: 9.8159 - accuracy:
0.2418
0.2415
0.2455
0.2490
0.2465
0.2549
58/96 [=========>...] - ETA: Os - loss: 8.9825 - accuracy:
0.2615
```

```
0.2689
0.2830
0.2857
0.2870
0.2887
0.2955
0.2993
0.3009
96/96 [=========== ] - 7s 26ms/step - loss: 7.9315 - accuracy:
0.3013 - val_loss: 8.0870 - val_accuracy: 0.1418
Epoch 2/30
1/96 [...] - ETA: 2s - loss: 5.5968 - accuracy:
0.5238
5/96 [>...] - ETA: 1s - loss: 5.7449 - accuracy:
0.4190
9/96 [=>...] - ETA: 1s - loss: 5.5635 - accuracy:
0.4603
14/96 [===>...] - ETA: 1s - loss: 5.5345 - accuracy:
0.4286
18/96 [====>...] - ETA: 1s - loss: 5.5046 - accuracy:
0.4392
22/96 [====>...] - ETA: Os - loss: 5.4640 - accuracy:
26/96 [======>...] - ETA: Os - loss: 5.3974 - accuracy:
30/96 [======>...] - ETA: Os - loss: 5.3610 - accuracy:
0.4476
34/96 [=======>...] - ETA: Os - loss: 5.3194 - accuracy:
39/96 [========>...] - ETA: Os - loss: 5.2681 - accuracy:
43/96 [========>...] - ETA: Os - loss: 5.2094 - accuracy:
0.4629
```

```
0.4792
0.4853
0.4954
0.5137
0.5150
0.5167
0.5306
0.5334
0.5397
0.5437 - val_loss: 4.3745 - val_accuracy: 0.4005
Epoch 3/30
1/96 [...] - ETA: 1s - loss: 3.5298 - accuracy:
0.6190
5/96 [>...] - ETA: 1s - loss: 3.7067 - accuracy:
10/96 [==>...] - ETA: 1s - loss: 3.5974 - accuracy:
0.6714
14/96 [===>...] - ETA: 1s - loss: 3.4990 - accuracy:
0.6973
17/96 [====>...] - ETA: 1s - loss: 3.5003 - accuracy:
0.7003
21/96 [====>...] - ETA: 1s - loss: 3.5332 - accuracy:
0.6848
25/96 [=====>...] - ETA: Os - loss: 3.4849 - accuracy:
0.6933
29/96 [======>...] - ETA: Os - loss: 3.4546 - accuracy:
0.6962
33/96 [=======>...] - ETA: Os - loss: 3.4626 - accuracy:
37/96 [========>...] - ETA: Os - loss: 3.4275 - accuracy:
41/96 [========>...] - ETA: Os - loss: 3.4011 - accuracy:
0.6969
```

```
0.6921
0.6933
0.6987
0.6933
0.6977
0.6958
0.7003
0.7043
0.7074
0.7118
96/96 [============= ] - 2s 16ms/step - loss: 3.0974 - accuracy:
0.7116 - val_loss: 3.1082 - val_accuracy: 0.5858
Epoch 4/30
1/96 [...] - ETA: 2s - loss: 2.4568 - accuracy:
0.7619
6/96 [>...] - ETA: 1s - loss: 2.5889 - accuracy:
0.7222
9/96 [=>...] - ETA: 1s - loss: 2.5569 - accuracy:
0.7513
14/96 [===>...] - ETA: 1s - loss: 2.4901 - accuracy:
19/96 [====>...] - ETA: Os - loss: 2.4175 - accuracy:
0.7769
24/96 [=====>...] - ETA: Os - loss: 2.3984 - accuracy:
0.7817
28/96 [======>...] - ETA: Os - loss: 2.3432 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 2.3439 - accuracy:
0.7908
36/96 [=======>...] - ETA: Os - loss: 2.3389 - accuracy:
0.7923
41/96 [========>...] - ETA: Os - loss: 2.3132 - accuracy:
```

```
0.7909
46/96 [========>...] - ETA: Os - loss: 2.3009 - accuracy:
0.7950
0.7937
0.7937
0.8000
0.7959
0.7973
0.7937
0.7944
0.7935
96/96 [============== ] - ETA: Os - loss: 2.1680 - accuracy:
0.7901
96/96 [============ ] - 2s 16ms/step - loss: 2.1680 - accuracy:
0.7901 - val_loss: 2.5184 - val_accuracy: 0.5816
Epoch 5/30
1/96 [...] - ETA: 2s - loss: 1.6855 - accuracy:
0.8571
5/96 [>...] - ETA: 1s - loss: 1.9052 - accuracy:
0.7810
9/96 [=>...] - ETA: 1s - loss: 1.8890 - accuracy:
0.7937
13/96 [===>...] - ETA: 1s - loss: 1.8972 - accuracy:
0.7912
17/96 [====>...] - ETA: 1s - loss: 1.8808 - accuracy:
0.7983
22/96 [====>...] - ETA: 1s - loss: 1.8283 - accuracy:
0.8030
27/96 [======>...] - ETA: Os - loss: 1.8032 - accuracy:
0.8042
32/96 [======>...] - ETA: Os - loss: 1.7793 - accuracy:
37/96 [========>...] - ETA: Os - loss: 1.7567 - accuracy:
0.8160
```

```
42/96 [========>...] - ETA: Os - loss: 1.7776 - accuracy:
0.8016
0.8045
0.8059
0.8053
0.8111
0.8206
0.8225
0.8246
0.8259
0.8288
0.8311 - val_loss: 2.6540 - val_accuracy: 0.5167
Epoch 6/30
1/96 [...] - ETA: 2s - loss: 1.1024 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 1.3645 - accuracy:
0.8286
9/96 [=>...] - ETA: 1s - loss: 1.3618 - accuracy:
13/96 [===>...] - ETA: 1s - loss: 1.3620 - accuracy:
0.8498
17/96 [====>...] - ETA: 1s - loss: 1.3418 - accuracy:
0.8515
22/96 [====>...] - ETA: Os - loss: 1.3222 - accuracy:
25/96 [=====>...] - ETA: Os - loss: 1.3140 - accuracy:
29/96 [======>...] - ETA: Os - loss: 1.3149 - accuracy:
0.8555
33/96 [=======>...] - ETA: Os - loss: 1.3051 - accuracy:
```

```
0.8571
38/96 [=======>...] - ETA: Os - loss: 1.2888 - accuracy:
0.8647
42/96 [========>...] - ETA: Os - loss: 1.2871 - accuracy:
0.8662
46/96 [========>...] - ETA: Os - loss: 1.2819 - accuracy:
0.8638
0.8632
0.8627
0.8668
0.8689
0.8681
0.8659
0.8676
0.8677
0.8676 - val_loss: 1.8036 - val_accuracy: 0.6926
Epoch 7/30
1/96 [...] - ETA: 1s - loss: 1.3105 - accuracy:
0.7619
6/96 [>...] - ETA: 1s - loss: 0.9772 - accuracy:
0.9127
11/96 [==>...] - ETA: Os - loss: 0.9492 - accuracy:
0.9264
16/96 [====>...] - ETA: Os - loss: 0.9349 - accuracy:
0.9226
20/96 [====>...] - ETA: Os - loss: 0.9445 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 0.9778 - accuracy:
29/96 [======>...] - ETA: Os - loss: 0.9637 - accuracy:
0.9113
```

```
34/96 [=======>...] - ETA: Os - loss: 0.9539 - accuracy:
0.9118
39/96 [========>...] - ETA: Os - loss: 0.9505 - accuracy:
0.9138
0.9079
0.9113
0.9091
59/96 [=========>...] - ETA: Os - loss: 0.9374 - accuracy:
0.9122
0.9133
0.9122
0.9099
0.9066
0.9076
0.9053
96/96 [============== ] - ETA: Os - loss: 0.9372 - accuracy:
0.9020 - val_loss: 1.7160 - val_accuracy: 0.6661
Epoch 8/30
1/96 [...] - ETA: 2s - loss: 0.9133 - accuracy:
0.8095
6/96 [>...] - ETA: 1s - loss: 0.8669 - accuracy:
0.8810
10/96 [==>...] - ETA: 1s - loss: 0.9063 - accuracy:
0.8762
13/96 [===>...] - ETA: 1s - loss: 0.8756 - accuracy:
0.8828
16/96 [====>...] - ETA: 1s - loss: 0.8752 - accuracy:
0.8839
20/96 [====>...] - ETA: 1s - loss: 0.8699 - accuracy:
```

```
0.8810
23/96 [=====>...] - ETA: 1s - loss: 0.8462 - accuracy:
0.8882
26/96 [======>...] - ETA: 1s - loss: 0.8599 - accuracy:
0.8846
30/96 [=======>...] - ETA: 1s - loss: 0.8388 - accuracy:
0.8905
33/96 [=======>...] - ETA: Os - loss: 0.8290 - accuracy:
0.8903
37/96 [========>...] - ETA: Os - loss: 0.8150 - accuracy:
0.8945
41/96 [========>...] - ETA: Os - loss: 0.8137 - accuracy:
0.8955
0.8998
0.9010
0.9004
0.9016
0.9017
0.9069
0.9101
0.9105
0.9103
0.9112
0.9078
0.9080 - val_loss: 1.4763 - val_accuracy: 0.7011
```

Epoch 9/30

```
1/96 [...] - ETA: 2s - loss: 0.5362 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.6639 - accuracy:
0.9143
9/96 [=>...] - ETA: 1s - loss: 0.6420 - accuracy:
0.9206
13/96 [===>...] - ETA: 1s - loss: 0.6403 - accuracy:
0.9231
17/96 [====>...] - ETA: 1s - loss: 0.6610 - accuracy:
0.9104
21/96 [====>...] - ETA: 1s - loss: 0.6457 - accuracy:
0.9161
25/96 [=====>...] - ETA: Os - loss: 0.6602 - accuracy:
29/96 [======>...] - ETA: Os - loss: 0.6567 - accuracy:
0.9195
32/96 [======>...] - ETA: Os - loss: 0.6536 - accuracy:
0.9167
35/96 [=======>...] - ETA: Os - loss: 0.6539 - accuracy:
0.9156
38/96 [=======>...] - ETA: Os - loss: 0.6538 - accuracy:
0.9148
42/96 [========>...] - ETA: Os - loss: 0.6586 - accuracy:
0.9127
45/96 [=========>...] - ETA: Os - loss: 0.6575 - accuracy:
0.9101
0.9067
0.8992
0.9002
0.9019
0.9000
0.8990
0.8987
0.9020
```

```
0.9016
0.9010 - val_loss: 1.7488 - val_accuracy: 0.6678
Epoch 10/30
1/96 [...] - ETA: 2s - loss: 0.4637 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.5725 - accuracy:
0.9429
10/96 [==>...] - ETA: 1s - loss: 0.5534 - accuracy:
0.9381
14/96 [===>...] - ETA: 1s - loss: 0.5517 - accuracy:
0.9286
18/96 [====>...] - ETA: Os - loss: 0.5345 - accuracy:
0.9339
22/96 [====>...] - ETA: Os - loss: 0.5430 - accuracy:
0.9329
26/96 [======>...] - ETA: Os - loss: 0.5607 - accuracy:
0.9286
30/96 [=======>...] - ETA: Os - loss: 0.5615 - accuracy:
0.9286
34/96 [=======>...] - ETA: Os - loss: 0.5478 - accuracy:
0.9314
38/96 [=======>...] - ETA: Os - loss: 0.5612 - accuracy:
42/96 [========>...] - ETA: Os - loss: 0.5609 - accuracy:
0.9286
46/96 [========>...] - ETA: Os - loss: 0.5488 - accuracy:
0.9327
0.9349
0.9316
0.9322
0.9318
```

```
0.9292
0.9294
0.9310
0.9304
0.9293
0.9290 - val_loss: 1.5161 - val_accuracy: 0.6704
Epoch 11/30
1/96 [...] - ETA: 2s - loss: 0.4524 - accuracy:
0.9524
5/96 [>...] - ETA: 1s - loss: 0.6487 - accuracy:
0.9143
10/96 [==>...] - ETA: 1s - loss: 0.7383 - accuracy:
0.8714
14/96 [===>...] - ETA: 1s - loss: 0.7217 - accuracy:
0.8878
18/96 [====>...] - ETA: 1s - loss: 0.7018 - accuracy:
0.8915
22/96 [====>...] - ETA: Os - loss: 0.6912 - accuracy:
0.8896
26/96 [======>...] - ETA: Os - loss: 0.6912 - accuracy:
0.8846
31/96 [======>...] - ETA: Os - loss: 0.6646 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.6544 - accuracy:
40/96 [========>...] - ETA: Os - loss: 0.6376 - accuracy:
0.8976
0.8963
0.8941
0.8949
0.8939
0.8977
0.8992
```

```
0.8989
0.9011
0.9030
0.9053
0.9058
0.9048
0.9055 - val_loss: 2.3300 - val_accuracy: 0.5047
Epoch 12/30
1/96 [...] - ETA: 1s - loss: 0.9528 - accuracy:
0.8095
6/96 [>...] - ETA: 1s - loss: 0.5016 - accuracy:
0.9286
11/96 [==>...] - ETA: Os - loss: 0.4270 - accuracy:
0.9610
16/96 [====>...] - ETA: Os - loss: 0.4235 - accuracy:
0.9613
20/96 [====>...] - ETA: Os - loss: 0.4158 - accuracy:
0.9619
24/96 [=====>...] - ETA: Os - loss: 0.4061 - accuracy:
28/96 [======>...] - ETA: Os - loss: 0.4096 - accuracy:
0.9626
32/96 [======>...] - ETA: Os - loss: 0.4171 - accuracy:
0.9613
37/96 [=======>...] - ETA: Os - loss: 0.4052 - accuracy:
0.9640
42/96 [========>...] - ETA: Os - loss: 0.4085 - accuracy:
46/96 [=========>...] - ETA: Os - loss: 0.4059 - accuracy:
0.9607
0.9524
0.9524
0.9494
```

```
0.9496
0.9491
0.9487
0.9453
0.9441
0.9445 - val_loss: 1.3819 - val_accuracy: 0.7071
Epoch 13/30
1/96 [...] - ETA: 1s - loss: 0.3950 - accuracy:
0.9524
5/96 [>...] - ETA: 1s - loss: 0.3290 - accuracy:
0.9714
9/96 [=>...] - ETA: 1s - loss: 0.4325 - accuracy:
0.9365
14/96 [===>...] - ETA: 1s - loss: 0.3945 - accuracy:
0.9490
18/96 [====>...] - ETA: 1s - loss: 0.3888 - accuracy:
0.9497
22/96 [====>...] - ETA: Os - loss: 0.4054 - accuracy:
0.9459
26/96 [======>...] - ETA: Os - loss: 0.3885 - accuracy:
30/96 [======>...] - ETA: Os - loss: 0.3897 - accuracy:
33/96 [======>...] - ETA: Os - loss: 0.3844 - accuracy:
0.9553
38/96 [=======>...] - ETA: 0s - loss: 0.3895 - accuracy:
0.9512
47/96 [========>...] - ETA: Os - loss: 0.3985 - accuracy:
0.9493
0.9468
58/96 [=========>...] - ETA: Os - loss: 0.4242 - accuracy:
0.9448
```

```
0.9433
0.9439
0.9436
0.9446
0.9429
0.9423
0.9422
0.9415 - val_loss: 1.4485 - val_accuracy: 0.6567
Epoch 14/30
1/96 [...] - ETA: 2s - loss: 0.2266 - accuracy:
1.0000
5/96 [>...] - ETA: 1s - loss: 0.3848 - accuracy:
0.9429
10/96 [==>...] - ETA: 1s - loss: 0.3704 - accuracy:
0.9381
14/96 [===>...] - ETA: 1s - loss: 0.3501 - accuracy:
0.9524
18/96 [====>...] - ETA: Os - loss: 0.3292 - accuracy:
0.9603
21/96 [====>...] - ETA: Os - loss: 0.3250 - accuracy:
0.9660
26/96 [======>...] - ETA: Os - loss: 0.3305 - accuracy:
0.9670
30/96 [======>...] - ETA: Os - loss: 0.3417 - accuracy:
0.9667
35/96 [=======>...] - ETA: Os - loss: 0.3430 - accuracy:
39/96 [========>...] - ETA: Os - loss: 0.3370 - accuracy:
0.9670
43/96 [========>...] - ETA: Os - loss: 0.3368 - accuracy:
0.9679
47/96 [========>...] - ETA: Os - loss: 0.3396 - accuracy:
0.9656
0.9664
0.9685
```

```
0.9657
0.9646
0.9645
0.9643
0.9643
0.9649
0.9637
0.9643
0.9635 - val_loss: 1.4799 - val_accuracy: 0.6678
Epoch 15/30
1/96 [...] - ETA: 1s - loss: 0.2749 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.3099 - accuracy:
0.9683
11/96 [==>...] - ETA: Os - loss: 0.2825 - accuracy:
0.9740
15/96 [===>...] - ETA: Os - loss: 0.2836 - accuracy:
0.9746
19/96 [====>...] - ETA: Os - loss: 0.2969 - accuracy:
24/96 [=====>...] - ETA: Os - loss: 0.2934 - accuracy:
0.9683
29/96 [======>...] - ETA: Os - loss: 0.2847 - accuracy:
0.9721
33/96 [=======>...] - ETA: Os - loss: 0.2785 - accuracy:
0.9740
37/96 [=======>...] - ETA: 0s - loss: 0.2816 - accuracy:
0.9730
40/96 [=======>...] - ETA: Os - loss: 0.2869 - accuracy:
0.9714
45/96 [========>...] - ETA: Os - loss: 0.2867 - accuracy:
0.9714
0.9714
0.9690
```

```
0.9702
0.9706
0.9696
0.9706
0.9712
0.9714
0.9721
0.9720 - val_loss: 1.4378 - val_accuracy: 0.6900
Epoch 16/30
1/96 [...] - ETA: 2s - loss: 0.2080 - accuracy:
1.0000
6/96 [>...] - ETA: 1s - loss: 0.2464 - accuracy:
0.9762
9/96 [=>...] - ETA: 1s - loss: 0.2435 - accuracy:
0.9735
13/96 [===>...] - ETA: 1s - loss: 0.2379 - accuracy:
0.9744
17/96 [====>...] - ETA: 1s - loss: 0.2289 - accuracy:
0.9776
21/96 [====>...] - ETA: 1s - loss: 0.2297 - accuracy:
0.9773
25/96 [=====>...] - ETA: Os - loss: 0.2234 - accuracy:
0.9810
30/96 [======>...] - ETA: Os - loss: 0.2293 - accuracy:
33/96 [=======>...] - ETA: Os - loss: 0.2417 - accuracy:
0.9726
37/96 [======>...] - ETA: Os - loss: 0.2368 - accuracy:
0.9743
42/96 [=======>...] - ETA: Os - loss: 0.2510 - accuracy:
0.9671
47/96 [========>...] - ETA: Os - loss: 0.2524 - accuracy:
0.9676
0.9686
```

```
0.9691
0.9704
0.9677
0.9654
0.9646
0.9640
0.9654
0.9620
0.9589
0.9576
96/96 [============= ] - ETA: Os - loss: 0.3063 - accuracy:
0.9560 - val_loss: 2.8661 - val_accuracy: 0.5568
Epoch 17/30
1/96 [...] - ETA: 2s - loss: 0.4784 - accuracy:
0.9048
4/96 [>...] - ETA: 1s - loss: 0.4106 - accuracy:
0.9048
7/96 [=>...] - ETA: 1s - loss: 0.3337 - accuracy:
0.9456
11/96 [==>...] - ETA: 1s - loss: 0.4019 - accuracy:
0.9394
15/96 [===>...] - ETA: 1s - loss: 0.4090 - accuracy:
0.9397
18/96 [====>...] - ETA: 1s - loss: 0.3935 - accuracy:
0.9444
22/96 [====>...] - ETA: 1s - loss: 0.3819 - accuracy:
0.9437
26/96 [======>...] - ETA: 1s - loss: 0.3916 - accuracy:
29/96 [======>...] - ETA: 1s - loss: 0.3811 - accuracy:
32/96 [=======>...] - ETA: 1s - loss: 0.3837 - accuracy:
0.9435
```

```
35/96 [=======>...] - ETA: Os - loss: 0.3812 - accuracy:
0.9429
38/96 [=======>...] - ETA: Os - loss: 0.3887 - accuracy:
41/96 [========>...] - ETA: Os - loss: 0.3964 - accuracy:
0.9419
0.9448
0.9454
0.9424
0.9429
0.9446
0.9449
0.9433
0.9455
0.9458
0.9435 - val_loss: 2.2033 - val_accuracy: 0.6849
Epoch 1/30
1/96 [...] - ETA: 4:56 - loss: 6.7465 - accuracy:
0.0952
5/96 [>...] - ETA: 1s - loss: 7.6976 - accuracy:
0.1619
10/96 [==>...] - ETA: 1s - loss: 7.3093 - accuracy:
0.1762
15/96 [===>...] - ETA: Os - loss: 6.9207 - accuracy:
```

```
0.1937
20/96 [====>...] - ETA: Os - loss: 6.6683 - accuracy:
0.2071
25/96 [=====>...] - ETA: Os - loss: 6.4988 - accuracy:
0.2171
30/96 [=======>...] - ETA: Os - loss: 6.3773 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 6.2271 - accuracy:
0.2435
41/96 [=======>...] - ETA: Os - loss: 6.0870 - accuracy:
0.2555
0.2584
0.2734
0.2790
0.2808
0.2988
0.3075
0.3083 - val_loss: 5.1498 - val_accuracy: 0.2340
Epoch 2/30
1/96 [...] - ETA: 2s - loss: 4.1134 - accuracy:
0.4762
6/96 [>...] - ETA: 1s - loss: 4.2132 - accuracy:
0.4365
11/96 [==>...] - ETA: Os - loss: 4.0589 - accuracy:
0.4805
16/96 [====>...] - ETA: Os - loss: 4.0363 - accuracy:
0.4821
19/96 [====>...] - ETA: Os - loss: 4.0517 - accuracy:
25/96 [=====>...] - ETA: Os - loss: 4.0207 - accuracy:
0.4629
```

```
31/96 [======>...] - ETA: Os - loss: 3.9775 - accuracy:
0.4731
37/96 [=======>...] - ETA: Os - loss: 3.9567 - accuracy:
0.4718
0.4702
0.4661
0.4778
0.4895
0.4886
0.4917
0.4907
0.4919
0.4990
0.4988 - val_loss: 3.6451 - val_accuracy: 0.3826
Epoch 3/30
1/96 [...] - ETA: 1s - loss: 2.8321 - accuracy:
0.6190
6/96 [>...] - ETA: Os - loss: 3.0048 - accuracy:
0.5714
12/96 [==>...] - ETA: Os - loss: 2.8571 - accuracy:
18/96 [====>...] - ETA: Os - loss: 2.8443 - accuracy:
0.6270
23/96 [=====>...] - ETA: Os - loss: 2.8552 - accuracy:
0.6108
27/96 [======>...] - ETA: Os - loss: 2.8213 - accuracy:
0.6155
31/96 [======>...] - ETA: Os - loss: 2.8330 - accuracy:
0.6114
36/96 [=======>...] - ETA: Os - loss: 2.8403 - accuracy:
0.6071
41/96 [========>...] - ETA: Os - loss: 2.8315 - accuracy:
```

```
0.6051
46/96 [=========>...] - ETA: Os - loss: 2.8152 - accuracy:
0.6035
0.6026
0.5968
0.6048
0.6065
0.6115
0.6131
0.6114
0.6137 - val_loss: 3.0409 - val_accuracy: 0.3954
Epoch 4/30
1/96 [...] - ETA: 2s - loss: 2.3855 - accuracy:
0.7143
6/96 [>...] - ETA: 1s - loss: 2.4080 - accuracy:
12/96 [==>...] - ETA: Os - loss: 2.2502 - accuracy:
0.7103
17/96 [====>...] - ETA: Os - loss: 2.2228 - accuracy:
0.7143
21/96 [====>...] - ETA: Os - loss: 2.2336 - accuracy:
0.6984
27/96 [======>...] - ETA: Os - loss: 2.2289 - accuracy:
0.6949
32/96 [=======>...] - ETA: Os - loss: 2.2597 - accuracy:
0.6726
37/96 [========>...] - ETA: Os - loss: 2.2369 - accuracy:
0.6731
42/96 [========>...] - ETA: Os - loss: 2.2290 - accuracy:
47/96 [========>...] - ETA: Os - loss: 2.2016 - accuracy:
0.6877
```

```
0.6934
0.6874
0.6873
0.6817
0.6804
0.6874
0.6867 - val_loss: 2.5227 - val_accuracy: 0.5098
Epoch 5/30
1/96 [...] - ETA: 1s - loss: 1.7510 - accuracy:
6/96 [>...] - ETA: Os - loss: 1.6693 - accuracy:
0.7857
11/96 [==>...] - ETA: Os - loss: 1.8175 - accuracy:
0.7446
15/96 [===>...] - ETA: Os - loss: 1.7748 - accuracy:
20/96 [====>...] - ETA: Os - loss: 1.7983 - accuracy:
0.7429
25/96 [=====>...] - ETA: Os - loss: 1.7924 - accuracy:
0.7410
29/96 [======>...] - ETA: Os - loss: 1.7652 - accuracy:
0.7455
33/96 [=======>...] - ETA: Os - loss: 1.7909 - accuracy:
37/96 [=======>...] - ETA: Os - loss: 1.7848 - accuracy:
41/96 [========>...] - ETA: Os - loss: 1.7677 - accuracy:
0.7375
0.7376
0.7376
0.7443
```

```
0.7397
0.7404
0.7359
0.7388
0.7429
0.7408
0.7367
0.7363
0.7383
0.7386 - val_loss: 2.2706 - val_accuracy: 0.5474
Epoch 6/30
1/96 [...] - ETA: 1s - loss: 1.4004 - accuracy:
0.8571
6/96 [>...] - ETA: Os - loss: 1.4896 - accuracy:
0.7460
10/96 [==>...] - ETA: Os - loss: 1.4691 - accuracy:
14/96 [===>...] - ETA: Os - loss: 1.4297 - accuracy:
0.7585
18/96 [====>...] - ETA: Os - loss: 1.3850 - accuracy:
0.7884
23/96 [=====>...] - ETA: Os - loss: 1.4113 - accuracy:
0.7867
28/96 [======>...] - ETA: Os - loss: 1.4068 - accuracy:
0.7925
33/96 [=======>...] - ETA: Os - loss: 1.4196 - accuracy:
0.7879
37/96 [========>...] - ETA: Os - loss: 1.4148 - accuracy:
0.7876
42/96 [=======>...] - ETA: Os - loss: 1.4300 - accuracy:
0.7834
47/96 [========>...] - ETA: Os - loss: 1.4395 - accuracy:
0.7852
```

```
0.7827
0.7823
0.7808
0.7765
0.7767
0.7705
0.7702
0.7684
0.7676 - val_loss: 1.9843 - val_accuracy: 0.5611
Epoch 7/30
1/96 [...] - ETA: 2s - loss: 1.5149 - accuracy:
0.6667
5/96 [>...] - ETA: 1s - loss: 1.3315 - accuracy:
0.7714
9/96 [=>...] - ETA: 1s - loss: 1.2159 - accuracy:
0.8201
14/96 [===>...] - ETA: 1s - loss: 1.2369 - accuracy:
0.8027
18/96 [====>...] - ETA: 1s - loss: 1.2354 - accuracy:
22/96 [====>...] - ETA: Os - loss: 1.2531 - accuracy:
0.7922
27/96 [======>...] - ETA: Os - loss: 1.2319 - accuracy:
0.7989
32/96 [=======>...] - ETA: Os - loss: 1.2125 - accuracy:
38/96 [=======>...] - ETA: Os - loss: 1.2090 - accuracy:
43/96 [========>...] - ETA: Os - loss: 1.1990 - accuracy:
0.8040
```

```
0.8027
0.8000
0.8006
0.8065
0.8074
0.8076
0.8036
0.8048
0.8031 - val_loss: 1.9256 - val_accuracy: 0.5602
Epoch 8/30
1/96 [...] - ETA: 1s - loss: 0.7625 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 1.0586 - accuracy:
0.8492
11/96 [==>...] - ETA: Os - loss: 1.0452 - accuracy:
0.8442
16/96 [====>...] - ETA: Os - loss: 1.0712 - accuracy:
0.8274
21/96 [====>...] - ETA: Os - loss: 1.0858 - accuracy:
0.8231
26/96 [======>...] - ETA: Os - loss: 1.0857 - accuracy:
0.8278
31/96 [======>...] - ETA: Os - loss: 1.0740 - accuracy:
0.8341
37/96 [=======>...] - ETA: 0s - loss: 1.0728 - accuracy:
0.8275
42/96 [========>...] - ETA: Os - loss: 1.0558 - accuracy:
0.8311
0.8383
0.8404
0.8378
```

```
0.8371
0.8356
0.8301
0.8298
0.8274
0.8261 - val_loss: 2.0714 - val_accuracy: 0.5619
Epoch 9/30
1/96 [...] - ETA: 1s - loss: 1.0436 - accuracy:
0.9048
7/96 [=>...] - ETA: Os - loss: 0.9967 - accuracy:
0.8435
12/96 [==>...] - ETA: Os - loss: 0.9966 - accuracy:
0.8333
17/96 [====>...] - ETA: Os - loss: 0.9595 - accuracy:
0.8487
23/96 [=====>...] - ETA: Os - loss: 0.9358 - accuracy:
0.8489
29/96 [======>...] - ETA: Os - loss: 0.9605 - accuracy:
0.8440
34/96 [=======>...] - ETA: Os - loss: 0.9526 - accuracy:
40/96 [=======>...] - ETA: Os - loss: 0.9498 - accuracy:
0.8488
46/96 [========>...] - ETA: Os - loss: 0.9410 - accuracy:
0.8520
0.8524
0.8543
0.8518
0.8577
```

```
0.8588
0.8587
0.8566 - val_loss: 1.9315 - val_accuracy: 0.5226
Epoch 10/30
1/96 [...] - ETA: 1s - loss: 0.9836 - accuracy:
0.8095
7/96 [=>...] - ETA: Os - loss: 0.8501 - accuracy:
0.8639
12/96 [==>...] - ETA: Os - loss: 0.7902 - accuracy:
18/96 [====>...] - ETA: Os - loss: 0.8213 - accuracy:
0.8624
23/96 [=====>...] - ETA: Os - loss: 0.8434 - accuracy:
0.8551
28/96 [======>...] - ETA: Os - loss: 0.8484 - accuracy:
0.8520
34/96 [=======>...] - ETA: Os - loss: 0.8595 - accuracy:
0.8515
39/96 [========>...] - ETA: Os - loss: 0.8616 - accuracy:
0.8462
45/96 [=========>...] - ETA: Os - loss: 0.8601 - accuracy:
0.8455
0.8491
0.8518
0.8547
0.8537
0.8539
0.8525
0.8531 - val_loss: 1.7459 - val_accuracy: 0.5892
```

```
1/96 [...] - ETA: 1s - loss: 0.6659 - accuracy:
0.9048
7/96 [=>...] - ETA: Os - loss: 0.7028 - accuracy:
0.9048
13/96 [===>...] - ETA: Os - loss: 0.7071 - accuracy:
0.8974
19/96 [====>...] - ETA: Os - loss: 0.7253 - accuracy:
0.8947
23/96 [=====>...] - ETA: Os - loss: 0.7364 - accuracy:
0.8986
28/96 [======>...] - ETA: Os - loss: 0.7351 - accuracy:
0.8997
33/96 [=======>...] - ETA: Os - loss: 0.7451 - accuracy:
38/96 [=======>...] - ETA: Os - loss: 0.7482 - accuracy:
0.8910
43/96 [========>...] - ETA: Os - loss: 0.7507 - accuracy:
0.8848
0.8899
0.8847
0.8837
0.8867
0.8866
0.8836
0.8807
96/96 [============ ] - 1s 15ms/step - loss: 0.7592 - accuracy:
0.8806 - val_loss: 1.7338 - val_accuracy: 0.6371
Epoch 12/30
1/96 [...] - ETA: 1s - loss: 0.5402 - accuracy:
1.0000
```

```
6/96 [>...] - ETA: Os - loss: 0.6896 - accuracy:
0.8968
12/96 [==>...] - ETA: Os - loss: 0.6703 - accuracy:
0.8889
17/96 [====>...] - ETA: Os - loss: 0.6841 - accuracy:
0.8936
22/96 [====>...] - ETA: Os - loss: 0.7087 - accuracy:
0.8853
28/96 [======>...] - ETA: Os - loss: 0.7270 - accuracy:
0.8810
32/96 [=======>...] - ETA: Os - loss: 0.7291 - accuracy:
0.8839
37/96 [=======>...] - ETA: Os - loss: 0.7220 - accuracy:
42/96 [========>...] - ETA: Os - loss: 0.7150 - accuracy:
0.8934
0.8956
0.8964
0.8954
0.8971
0.8950
0.8917
0.8955
0.8935
0.8926 - val_loss: 1.7647 - val_accuracy: 0.6132
Epoch 13/30
1/96 [...] - ETA: 2s - loss: 0.6963 - accuracy:
0.8571
6/96 [>...] - ETA: Os - loss: 0.7030 - accuracy:
12/96 [==>...] - ETA: Os - loss: 0.6292 - accuracy:
0.8929
18/96 [====>...] - ETA: Os - loss: 0.6139 - accuracy:
```

```
0.9074
24/96 [=====>...] - ETA: Os - loss: 0.6210 - accuracy:
0.9008
30/96 [=======>...] - ETA: Os - loss: 0.6258 - accuracy:
0.9000
36/96 [========>...] - ETA: Os - loss: 0.6305 - accuracy:
0.9021
41/96 [========>...] - ETA: Os - loss: 0.6311 - accuracy:
0.9001
46/96 [=========>...] - ETA: Os - loss: 0.6464 - accuracy:
0.8975
0.9011
0.8988
0.9014
0.8991
0.8991
0.8986 - val_loss: 2.0739 - val_accuracy: 0.5995
Epoch 14/30
1/96 [...] - ETA: 1s - loss: 0.5311 - accuracy:
0.9524
6/96 [>...] - ETA: Os - loss: 0.5089 - accuracy:
0.9683
12/96 [==>...] - ETA: Os - loss: 0.5777 - accuracy:
0.9206
18/96 [====>...] - ETA: Os - loss: 0.5738 - accuracy:
0.9286
23/96 [=====>...] - ETA: Os - loss: 0.5722 - accuracy:
0.9296
28/96 [======>...] - ETA: Os - loss: 0.5684 - accuracy:
0.9337
34/96 [======>...] - ETA: Os - loss: 0.5897 - accuracy:
40/96 [========>...] - ETA: Os - loss: 0.5921 - accuracy:
0.9167
```

```
0.9120
0.9139
0.9130
0.9124
0.9132
0.9157
0.9189
0.9192
0.9173
0.9170 - val_loss: 2.0244 - val_accuracy: 0.6003
Epoch 15/30
1/96 [...] - ETA: 2s - loss: 0.6961 - accuracy:
0.8571
5/96 [>...] - ETA: 1s - loss: 0.5518 - accuracy:
0.9333
9/96 [=>...] - ETA: 1s - loss: 0.6047 - accuracy:
0.9259
14/96 [===>...] - ETA: 1s - loss: 0.5941 - accuracy:
0.9218
19/96 [====>...] - ETA: Os - loss: 0.5881 - accuracy:
0.9148
24/96 [=====>...] - ETA: Os - loss: 0.5950 - accuracy:
28/96 [======>...] - ETA: Os - loss: 0.5817 - accuracy:
0.9235
32/96 [=======>...] - ETA: Os - loss: 0.5717 - accuracy:
0.9256
36/96 [========>...] - ETA: Os - loss: 0.5646 - accuracy:
40/96 [========>...] - ETA: Os - loss: 0.5616 - accuracy:
0.9274
44/96 [========>...] - ETA: Os - loss: 0.5699 - accuracy:
0.9221
```

```
0.9206
0.9194
0.9192
0.9189
0.9153
0.9127
0.9111
0.9120
0.9100
96/96 [============== ] - ETA: Os - loss: 0.5985 - accuracy:
0.9095 - val_loss: 2.0237 - val_accuracy: 0.5662
Epoch 16/30
1/96 [...] - ETA: 2s - loss: 0.3662 - accuracy:
1.0000
6/96 [>...] - ETA: Os - loss: 0.5138 - accuracy:
0.9524
11/96 [==>...] - ETA: Os - loss: 0.5675 - accuracy:
0.9177
16/96 [====>...] - ETA: Os - loss: 0.5544 - accuracy:
0.9226
21/96 [====>...] - ETA: Os - loss: 0.5381 - accuracy:
0.9297
26/96 [======>...] - ETA: Os - loss: 0.5325 - accuracy:
0.9322
31/96 [=======>...] - ETA: Os - loss: 0.5360 - accuracy:
0.9278
36/96 [========>...] - ETA: Os - loss: 0.5285 - accuracy:
0.9325
40/96 [=======>...] - ETA: Os - loss: 0.5303 - accuracy:
44/96 [========>...] - ETA: Os - loss: 0.5282 - accuracy:
0.9275
```

```
0.9315
0.9328
0.9360
0.9387
0.9396
0.9371
0.9385
0.9362
0.9365 - val_loss: 1.8655 - val_accuracy: 0.5824
Epoch 1/30
1/96 [...] - ETA: 7:19 - loss: 3.5132 - accuracy:
0.0476
6/96 [>...] - ETA: 1s - loss: 3.3309 - accuracy:
0.1587
9/96 [=>...] - ETA: 1s - loss: 3.2590 - accuracy:
0.1693
14/96 [===>...] - ETA: 1s - loss: 3.1443 - accuracy:
0.1871
18/96 [====>...] - ETA: 1s - loss: 3.0816 - accuracy:
22/96 [====>...] - ETA: Os - loss: 3.0343 - accuracy:
0.2165
27/96 [======>...] - ETA: Os - loss: 2.9976 - accuracy:
0.2046
32/96 [=======>...] - ETA: Os - loss: 2.9336 - accuracy:
37/96 [=======>...] - ETA: Os - loss: 2.8783 - accuracy:
41/96 [========>...] - ETA: Os - loss: 2.8493 - accuracy:
0.2300
```

```
0.2381
0.2371
0.2427
0.2446
0.2597
0.2720
0.2744
0.2767
0.2873
0.2874 - val_loss: 3.8389 - val_accuracy: 0.1486
Epoch 2/30
1/96 [...] - ETA: 2s - loss: 1.9843 - accuracy:
0.4762
4/96 [>...] - ETA: 1s - loss: 2.0673 - accuracy:
0.4524
7/96 [=>...] - ETA: 1s - loss: 2.1530 - accuracy:
0.4014
10/96 [==>...] - ETA: 1s - loss: 2.1745 - accuracy:
0.3905
13/96 [===>...] - ETA: 1s - loss: 2.1470 - accuracy:
16/96 [====>...] - ETA: 1s - loss: 2.1664 - accuracy:
19/96 [====>...] - ETA: 1s - loss: 2.1725 - accuracy:
0.3684
```

```
22/96 [====>...] - ETA: 1s - loss: 2.1628 - accuracy:
0.3874
25/96 [=====>...] - ETA: 1s - loss: 2.1609 - accuracy:
0.3943
28/96 [======>...] - ETA: 1s - loss: 2.1387 - accuracy:
0.3963
31/96 [======>...] - ETA: 1s - loss: 2.1288 - accuracy:
0.4040
34/96 [=======>...] - ETA: 1s - loss: 2.1068 - accuracy:
0.4146
38/96 [=======>...] - ETA: 1s - loss: 2.0995 - accuracy:
0.4085
41/96 [======>...] - ETA: 1s - loss: 2.1001 - accuracy:
44/96 [========>...] - ETA: Os - loss: 2.0715 - accuracy:
0.4242
47/96 [=========>...] - ETA: Os - loss: 2.0468 - accuracy:
0.4367
0.4362
0.4394
0.4452
0.4490
0.4576
0.4594
0.4588
0.4613
0.4667
0.4722
0.4788
```

```
0.4792
0.4793 - val_loss: 3.6492 - val_accuracy: 0.1759
Epoch 3/30
1/96 [...] - ETA: 2s - loss: 1.2184 - accuracy:
6/96 [>...] - ETA: 1s - loss: 1.6363 - accuracy:
0.5794
11/96 [==>...] - ETA: Os - loss: 1.4809 - accuracy:
0.6277
16/96 [====>...] - ETA: Os - loss: 1.4607 - accuracy:
0.6190
20/96 [====>...] - ETA: Os - loss: 1.4777 - accuracy:
0.6238
23/96 [=====>...] - ETA: Os - loss: 1.4728 - accuracy:
0.6232
27/96 [======>...] - ETA: Os - loss: 1.4664 - accuracy:
0.6173
31/96 [======>...] - ETA: Os - loss: 1.4559 - accuracy:
35/96 [=======>...] - ETA: Os - loss: 1.4589 - accuracy:
0.6163
40/96 [========>...] - ETA: Os - loss: 1.4577 - accuracy:
0.6190
44/96 [========>...] - ETA: Os - loss: 1.4635 - accuracy:
0.6160
0.6152
0.6217
0.6168
0.6183
0.6190
0.6197
0.6190
```

```
0.6173
0.6168
0.6164
0.6170
0.6162 - val_loss: 2.5648 - val_accuracy: 0.2912
Epoch 4/30
1/96 [...] - ETA: 2s - loss: 1.1117 - accuracy:
0.7619
5/96 [>...] - ETA: 1s - loss: 1.2734 - accuracy:
0.6762
9/96 [=>...] - ETA: 1s - loss: 1.3177 - accuracy:
0.6561
13/96 [===>...] - ETA: 1s - loss: 1.2585 - accuracy:
0.6593
17/96 [====>...] - ETA: 1s - loss: 1.2565 - accuracy:
0.6499
21/96 [====>...] - ETA: 1s - loss: 1.2376 - accuracy:
0.6576
25/96 [=====>...] - ETA: Os - loss: 1.2186 - accuracy:
0.6629
29/96 [======>...] - ETA: Os - loss: 1.2183 - accuracy:
0.6601
32/96 [=======>...] - ETA: Os - loss: 1.2311 - accuracy:
37/96 [=======>...] - ETA: Os - loss: 1.2266 - accuracy:
0.6705
41/96 [========>...] - ETA: Os - loss: 1.2207 - accuracy:
0.6748
0.6775
0.6815
0.6810
0.6850
0.6878
```

```
0.6889
0.6885
0.6872
0.6852
0.6839
0.6857 - val_loss: 1.8344 - val_accuracy: 0.4833
Epoch 5/30
1/96 [...] - ETA: 2s - loss: 1.1073 - accuracy:
0.7619
5/96 [>...] - ETA: 1s - loss: 1.1755 - accuracy:
0.6762
9/96 [=>...] - ETA: 1s - loss: 1.0624 - accuracy:
14/96 [===>...] - ETA: 1s - loss: 1.0419 - accuracy:
0.7279
17/96 [====>...] - ETA: 1s - loss: 1.0835 - accuracy:
0.7171
20/96 [====>...] - ETA: 1s - loss: 1.0303 - accuracy:
24/96 [=====>...] - ETA: 1s - loss: 1.0259 - accuracy:
0.7480
28/96 [======>...] - ETA: Os - loss: 0.9999 - accuracy:
0.7602
32/96 [=======>...] - ETA: Os - loss: 0.9978 - accuracy:
0.7634
36/96 [=======>...] - ETA: Os - loss: 1.0014 - accuracy:
40/96 [========>...] - ETA: Os - loss: 1.0053 - accuracy:
43/96 [========>...] - ETA: Os - loss: 0.9967 - accuracy:
0.7608
0.7548
0.7470
0.7489
```

```
0.7498
0.7468
0.7456
0.7435
0.7410
0.7424
0.7435
0.7399
0.7386 - val_loss: 1.5410 - val_accuracy: 0.5628
Epoch 6/30
1/96 [...] - ETA: 1s - loss: 1.0661 - accuracy:
0.7619
5/96 [>...] - ETA: 1s - loss: 0.8574 - accuracy:
0.7810
10/96 [==>...] - ETA: 1s - loss: 0.9420 - accuracy:
0.7667
15/96 [===>...] - ETA: Os - loss: 0.8879 - accuracy:
20/96 [====>...] - ETA: Os - loss: 0.8738 - accuracy:
0.7881
25/96 [=====>...] - ETA: Os - loss: 0.8688 - accuracy:
0.7905
29/96 [======>...] - ETA: Os - loss: 0.8766 - accuracy:
0.7915
32/96 [=======>...] - ETA: Os - loss: 0.8730 - accuracy:
0.7902
37/96 [========>...] - ETA: Os - loss: 0.8563 - accuracy:
0.7979
41/96 [=======>...] - ETA: Os - loss: 0.8565 - accuracy:
0.7944
0.7919
0.7968
```

```
0.7944
0.7854
0.7883
0.7848
0.7814
0.7810
0.7806 - val_loss: 1.6913 - val_accuracy: 0.5465
Epoch 7/30
1/96 [...] - ETA: 1s - loss: 1.0285 - accuracy:
0.7143
5/96 [>...] - ETA: 1s - loss: 0.7163 - accuracy:
0.8190
10/96 [==>...] - ETA: 1s - loss: 0.7578 - accuracy:
0.7905
15/96 [===>...] - ETA: Os - loss: 0.7833 - accuracy:
0.7841
20/96 [====>...] - ETA: Os - loss: 0.8006 - accuracy:
0.7786
25/96 [=====>...] - ETA: Os - loss: 0.8211 - accuracy:
0.7771
29/96 [======>...] - ETA: Os - loss: 0.7946 - accuracy:
34/96 [=======>...] - ETA: Os - loss: 0.7914 - accuracy:
39/96 [========>...] - ETA: Os - loss: 0.8381 - accuracy:
0.7705
44/96 [========>...] - ETA: Os - loss: 0.8376 - accuracy:
0.7847
0.7907
```

```
0.7903
0.7931
0.8001
0.8035
0.8076
0.8095
0.8095
0.8091 - val_loss: 1.6851 - val_accuracy: 0.5534
Epoch 8/30
1/96 [...] - ETA: 1s - loss: 0.4579 - accuracy:
0.8571
6/96 [>...] - ETA: 1s - loss: 0.5642 - accuracy:
0.8651
10/96 [==>...] - ETA: 1s - loss: 0.6036 - accuracy:
0.8667
14/96 [===>...] - ETA: 1s - loss: 0.5809 - accuracy:
0.8776
19/96 [====>...] - ETA: 1s - loss: 0.6070 - accuracy:
0.8647
23/96 [=====>...] - ETA: Os - loss: 0.6413 - accuracy:
0.8571
28/96 [======>...] - ETA: Os - loss: 0.6389 - accuracy:
0.8605
33/96 [=======>...] - ETA: Os - loss: 0.6260 - accuracy:
0.8571
38/96 [========>...] - ETA: Os - loss: 0.6303 - accuracy:
0.8584
42/96 [========>...] - ETA: Os - loss: 0.6409 - accuracy:
0.8571
0.8486
```

```
0.8454
0.8478
0.8410
0.8413
0.8433
0.8419
0.8401 - val_loss: 1.5028 - val_accuracy: 0.6277
Epoch 9/30
1/96 [...] - ETA: 2s - loss: 0.6941 - accuracy:
6/96 [>...] - ETA: 1s - loss: 0.5388 - accuracy:
0.8651
11/96 [==>...] - ETA: Os - loss: 0.5711 - accuracy:
0.8485
16/96 [====>...] - ETA: Os - loss: 0.5646 - accuracy:
20/96 [====>...] - ETA: Os - loss: 0.5563 - accuracy:
0.8548
24/96 [=====>...] - ETA: Os - loss: 0.5497 - accuracy:
0.8631
29/96 [======>...] - ETA: Os - loss: 0.5420 - accuracy:
0.8686
32/96 [=======>...] - ETA: Os - loss: 0.5518 - accuracy:
36/96 [=======>...] - ETA: Os - loss: 0.5707 - accuracy:
0.8624
40/96 [========>...] - ETA: Os - loss: 0.5626 - accuracy:
0.8679
43/96 [========>...] - ETA: Os - loss: 0.5616 - accuracy:
46/96 [========>...] - ETA: Os - loss: 0.5591 - accuracy:
0.8716
0.8746
```

```
0.8727
0.8684
0.8719
0.8701
0.8703
0.8704
0.8662
0.8629
0.8654
0.8602
0.8596 - val_loss: 1.3935 - val_accuracy: 0.6516
Epoch 10/30
1/96 [...] - ETA: 2s - loss: 0.3345 - accuracy:
0.8571
4/96 [>...] - ETA: 1s - loss: 0.4014 - accuracy:
0.9048
8/96 [=>...] - ETA: 1s - loss: 0.5715 - accuracy:
0.8571
11/96 [==>...] - ETA: 1s - loss: 0.5390 - accuracy:
0.8701
14/96 [===>...] - ETA: 1s - loss: 0.5490 - accuracy:
0.8707
18/96 [====>...] - ETA: 1s - loss: 0.5253 - accuracy:
0.8836
22/96 [====>...] - ETA: 1s - loss: 0.5413 - accuracy:
26/96 [======>...] - ETA: 1s - loss: 0.5502 - accuracy:
0.8773
```

```
30/96 [======>...] - ETA: 1s - loss: 0.5346 - accuracy:
0.8825
34/96 [=======>...] - ETA: Os - loss: 0.5355 - accuracy:
37/96 [=======>...] - ETA: 0s - loss: 0.5335 - accuracy:
0.8842
0.8821
43/96 [========>...] - ETA: Os - loss: 0.5484 - accuracy:
0.8826
0.8820
0.8776
0.8768
0.8740
0.8718
0.8690
0.8704
0.8698
0.8692
0.8676
0.8682
0.8676 - val_loss: 1.4281 - val_accuracy: 0.6490
Epoch 11/30
1/96 [...] - ETA: 3s - loss: 0.3437 - accuracy:
0.9048
4/96 [>...] - ETA: 1s - loss: 0.5351 - accuracy:
```

```
0.8452
8/96 [=>...] - ETA: 1s - loss: 0.6316 - accuracy:
0.8036
11/96 [==>...] - ETA: 1s - loss: 0.6076 - accuracy:
0.8225
15/96 [===>...] - ETA: 1s - loss: 0.5730 - accuracy:
20/96 [====>...] - ETA: 1s - loss: 0.5354 - accuracy:
0.8476
24/96 [=====>...] - ETA: 1s - loss: 0.5292 - accuracy:
0.8512
28/96 [======>...] - ETA: Os - loss: 0.5206 - accuracy:
0.8554
32/96 [=======>...] - ETA: Os - loss: 0.5162 - accuracy:
37/96 [=======>...] - ETA: Os - loss: 0.5050 - accuracy:
0.8674
40/96 [========>...] - ETA: Os - loss: 0.5010 - accuracy:
0.8690
0.8723
0.8773
0.8741
0.8746
0.8760
0.8763
0.8783
0.8743
0.8767
0.8749
0.8741 - val_loss: 1.5376 - val_accuracy: 0.5764
```

```
Epoch 12/30
```

```
1/96 [...] - ETA: 2s - loss: 0.7687 - accuracy:
0.8095
4/96 [>...] - ETA: 1s - loss: 0.4949 - accuracy:
0.8571
8/96 [=>...] - ETA: 1s - loss: 0.4447 - accuracy:
0.8929
13/96 [===>...] - ETA: 1s - loss: 0.4514 - accuracy:
0.9011
16/96 [====>...] - ETA: 1s - loss: 0.4530 - accuracy:
0.9018
21/96 [====>...] - ETA: 1s - loss: 0.4774 - accuracy:
26/96 [======>...] - ETA: Os - loss: 0.4811 - accuracy:
0.8864
30/96 [======>...] - ETA: Os - loss: 0.4803 - accuracy:
0.8841
35/96 [=======>...] - ETA: Os - loss: 0.4701 - accuracy:
0.8871
40/96 [=======>...] - ETA: Os - loss: 0.4627 - accuracy:
0.8905
44/96 [========>...] - ETA: Os - loss: 0.4555 - accuracy:
0.8929
0.8909
0.8859
0.8836
0.8887
0.8904
0.8893
0.8877
0.8865
0.8841 - val_loss: 1.4914 - val_accuracy: 0.6405
```

```
Epoch 13/30
```

```
1/96 [...] - ETA: 2s - loss: 0.5967 - accuracy:
0.8571
4/96 [>...] - ETA: 1s - loss: 0.4349 - accuracy:
8/96 [=>...] - ETA: 1s - loss: 0.4365 - accuracy:
0.8810
13/96 [===>...] - ETA: 1s - loss: 0.4369 - accuracy:
0.8901
18/96 [====>...] - ETA: 1s - loss: 0.4190 - accuracy:
0.8942
23/96 [=====>...] - ETA: Os - loss: 0.4135 - accuracy:
0.9006
27/96 [======>...] - ETA: Os - loss: 0.4127 - accuracy:
0.8995
31/96 [======>...] - ETA: Os - loss: 0.4140 - accuracy:
0.9002
35/96 [=======>...] - ETA: Os - loss: 0.4119 - accuracy:
0.8993
39/96 [========>...] - ETA: Os - loss: 0.4079 - accuracy:
0.9023
44/96 [========>...] - ETA: Os - loss: 0.4086 - accuracy:
0.9026
0.9027
0.9066
0.9065
0.9055
0.9091
0.9086
0.9036
0.9042
0.9031
0.9037
```

```
0.9013
0.9010 - val_loss: 1.4376 - val_accuracy: 0.6465
Epoch 14/30
1/96 [...] - ETA: 2s - loss: 0.3227 - accuracy:
0.9048
4/96 [>...] - ETA: 1s - loss: 0.3076 - accuracy:
0.9286
8/96 [=>...] - ETA: 1s - loss: 0.3086 - accuracy:
0.9464
12/96 [==>...] - ETA: 1s - loss: 0.3275 - accuracy:
15/96 [===>...] - ETA: 1s - loss: 0.3330 - accuracy:
0.9365
18/96 [====>...] - ETA: 1s - loss: 0.3238 - accuracy:
0.9418
23/96 [=====>...] - ETA: 1s - loss: 0.3213 - accuracy:
0.9358
27/96 [======>...] - ETA: Os - loss: 0.3362 - accuracy:
0.9312
31/96 [======>...] - ETA: Os - loss: 0.3353 - accuracy:
0.9339
36/96 [=======>...] - ETA: Os - loss: 0.3351 - accuracy:
0.9339
40/96 [========>...] - ETA: Os - loss: 0.3412 - accuracy:
44/96 [========>...] - ETA: Os - loss: 0.3365 - accuracy:
0.9332
0.9362
0.9349
0.9315
0.9255
0.9228
0.9162
```

The Bayesian optimization gives us the best hyperparameters for the advanced model as: *num_layers: 4 * num_filters: 256

Overall, the accuracy of the advanced model is about as similar as Model Averaging at 0.72. This does not seem to find the best parameters as indicated by the increased accuracy of Random Search. This can be due to overtraining the model leading to a overfitted model.

Task 5: Data Augmentation, Model Validation of Personal Voice Recordings, and Transfer Learning First we record our own voice for about 10 times in two classes 'yes' and 'no'. We also ensure that the recordings are of the same length as the dataset recordings, which is of 1s each.

We augment the data by adding random noise to the recordings to generate more data. We randomly add 5% noise to the recordings.

[14]: !unzip /content/personalRecordings.zip

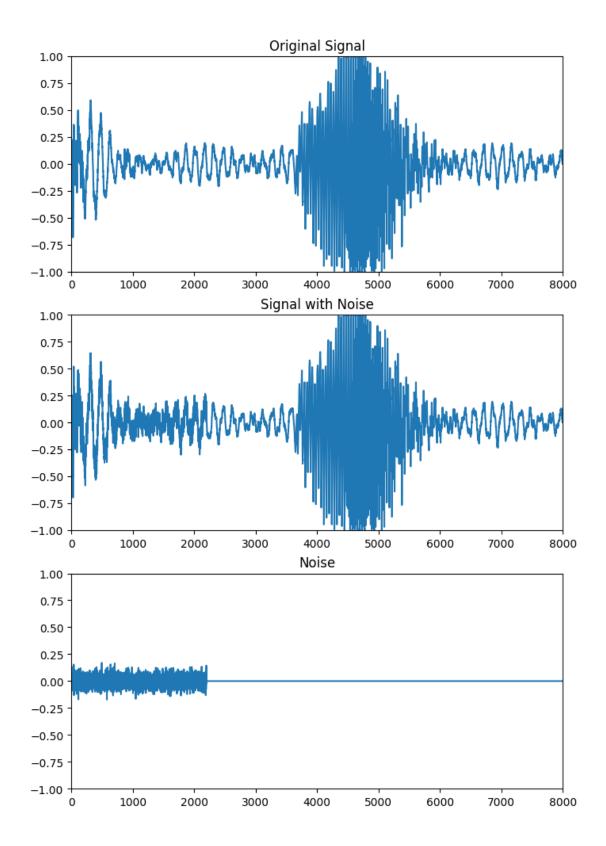
```
Archive: /content/personalRecordings.zip
  creating: personalRecordings/
  creating: personalRecordings/no/
  inflating: personalRecordings/no/0.wav
  inflating: personalRecordings/no/1.wav
  inflating: personalRecordings/no/2.wav
  inflating: personalRecordings/no/3.wav
  inflating: personalRecordings/no/4.wav
  inflating: personalRecordings/no/5.wav
  inflating: personalRecordings/no/6.wav
  inflating: personalRecordings/no/7.wav
  inflating: personalRecordings/no/8.wav
  inflating: personalRecordings/no/9.wav
  inflating: personalRecordings/no/0 with noise 0.wav
  inflating: personalRecordings/no/0_with_noise_1.wav
  inflating: personalRecordings/no/1 with noise 0.wav
  inflating: personalRecordings/no/1_with_noise_1.wav
  inflating: personalRecordings/no/2 with noise 0.wav
  inflating: personalRecordings/no/2_with_noise_1.wav
  inflating: personalRecordings/no/3_with_noise_0.wav
  inflating: personalRecordings/no/3_with_noise_1.wav
```

```
inflating: personalRecordings/no/4_with_noise_0.wav
       inflating: personalRecordings/no/4_with_noise_1.wav
       inflating: personalRecordings/no/5_with_noise_0.wav
       inflating: personalRecordings/no/5 with noise 1.wav
       inflating: personalRecordings/no/6 with noise 0.wav
       inflating: personalRecordings/no/6 with noise 1.wav
       inflating: personalRecordings/no/7 with noise 0.wav
       inflating: personalRecordings/no/7 with noise 1.wav
       inflating: personalRecordings/no/8 with noise 0.wav
       inflating: personalRecordings/no/8_with_noise_1.wav
       inflating: personalRecordings/no/9_with_noise_0.wav
       inflating: personalRecordings/no/9_with_noise_1.wav
        creating: personalRecordings/yes/
       inflating: personalRecordings/yes/0.wav
       inflating: personalRecordings/yes/1.wav
       inflating: personalRecordings/yes/2.wav
       inflating: personalRecordings/yes/3.wav
       inflating: personalRecordings/yes/4.wav
       inflating: personalRecordings/yes/5.wav
       inflating: personalRecordings/yes/6.wav
       inflating: personalRecordings/yes/7.wav
       inflating: personalRecordings/yes/8.wav
       inflating: personalRecordings/yes/9.wav
       inflating: personalRecordings/yes/0 with noise 0.wav
       inflating: personalRecordings/yes/0_with_noise_1.wav
       inflating: personalRecordings/yes/1_with_noise_0.wav
       inflating: personalRecordings/yes/1_with_noise_1.wav
       inflating: personalRecordings/yes/2_with_noise_0.wav
       inflating: personalRecordings/yes/2 with noise 1.way
       inflating: personalRecordings/yes/3_with_noise_0.wav
       inflating: personalRecordings/yes/3_with_noise_1.wav
       inflating: personalRecordings/yes/4_with_noise_0.wav
       inflating: personalRecordings/yes/4_with_noise_1.wav
       inflating: personalRecordings/yes/5 with noise 0.wav
       inflating: personalRecordings/yes/5 with noise 1.wav
       inflating: personalRecordings/yes/6 with noise 0.wav
       inflating: personalRecordings/yes/6 with noise 1.wav
       inflating: personalRecordings/yes/7_with_noise_0.wav
       inflating: personalRecordings/yes/7_with_noise_1.wav
       inflating: personalRecordings/yes/8_with_noise_0.wav
       inflating: personalRecordings/yes/8_with_noise_1.wav
       inflating: personalRecordings/yes/9_with_noise_0.wav
       inflating: personalRecordings/yes/9_with_noise_1.wav
[20]: recorded_classes = ['no', 'yes']
      num_recordings = 10
      num_synthetic_recordings = 2
```

```
noise_percentage = 0.01
viewing_label = True
for word in recorded_classes:
   for i in range(num_recordings):
        sound_path = f'personalRecordings/{word}/{i}.wav'
        # trim the audio file to 1 second
        sound = AudioSegment.from_file(sound_path)
        sound = sound[:1000]
        sound.export(sound_path, format="wav")
        # capture the signal and sample rate
       signal, sr = librosa.load(sound_path, sr=8000)
        for j in range(num_synthetic_recordings):
            # add noise to the signal to only a part of the signal
           random_index = random.randint(0, len(signal) - 1)
            noise = np.random.normal(0, noise_percentage, random_index)
            signal_with_noise = signal.copy()
            signal_with_noise[:random_index] += noise
            signal_noise = signal_with_noise - signal
            # export the signal with noise
            sound_with_noise_path = f'personalRecordings/{word}/
 sf.write(sound_with_noise_path, signal_with_noise, sr)
        if viewing_label:
            viewing_label = False
            # use subplots to plot the original signal, signal with noise and \Box
 ⇒noise
            fig, axs = plt.subplots(3, figsize=(8, 12))
            fig.suptitle(f'Original Signal, Signal with Noise, and Noise - U

√{word}')
            # set x-axis and y-axis limits
            axs[0].set_xlim(0, len(signal))
            axs[0].set_ylim(-1, 1)
            axs[1].set_xlim(0, len(signal))
            axs[1].set_ylim(-1, 1)
            axs[2].set_xlim(0, len(signal))
            axs[2].set_ylim(-1, 1)
```

```
axs[0].plot(signal)
axs[0].set_title('Original Signal')
axs[1].plot(signal_with_noise)
axs[1].set_title('Signal with Noise')
axs[2].plot(signal_noise)
axs[2].set_title('Noise')
plt.show()
```



As you can see, random indices are chosen to add noise to the recordings. So, for every iteration, the noise added to the recordings is different.

We now convert the recordings to spectrograms and use the advanced model to predict the classes of the recordings. This is done by the helper script provided in MATLAB.

First let's analyze the confusion matrix of the model on the personal recordings.

T5.1 Analysis of Confusion matrix on 'yes' and 'no' classes on Advanced Model: We now use the random search model hyperparameters to baseoff our personal recordings against.

```
[46]: task5_model = t1_model(len([2, 3, 4, 5]), [128, 128, 128, 128], upassthrough=True) task5_model.summary()
```

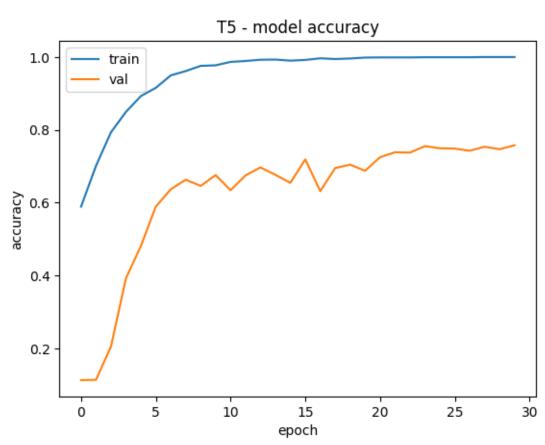
Model: "sequential_38"

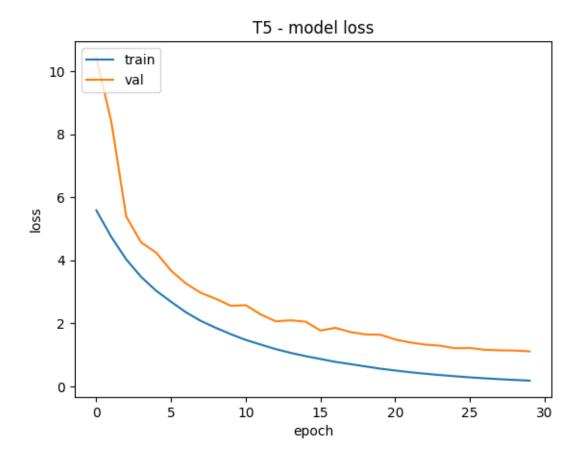
	Layer (type)	Output	-			Param #
•	conv2d_146 (Conv2D)	(None,				320
	<pre>batch_normalization_146 (B atchNormalization)</pre>	(None,	98,	50,	32)	128
	conv2d_147 (Conv2D)	(None,	98,	50,	128)	36992
	<pre>batch_normalization_147 (B atchNormalization)</pre>	(None,	98,	50,	128)	512
	<pre>max_pooling2d_167 (MaxPool ing2D)</pre>	(None,	49,	25,	128)	0
	conv2d_148 (Conv2D)	(None,	49,	25,	128)	147584
	batch_normalization_148 (B atchNormalization)	(None,	49,	25,	128)	512
	<pre>max_pooling2d_168 (MaxPool ing2D)</pre>	(None,	25,	13,	128)	0
	conv2d_149 (Conv2D)	(None,	25,	13,	128)	147584
	<pre>batch_normalization_149 (B atchNormalization)</pre>	(None,	25,	13,	128)	512
	max_pooling2d_169 (MaxPool	(None,	13,	7,	128)	0

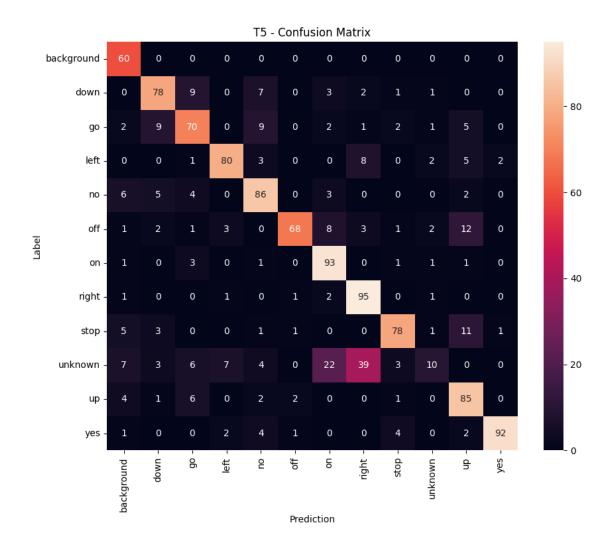
```
ing2D)
      conv2d_150 (Conv2D)
                                   (None, 13, 7, 128)
                                                             147584
      batch_normalization_150 (B (None, 13, 7, 128)
                                                             512
      atchNormalization)
      max_pooling2d_170 (MaxPool (None, 7, 4, 128)
      ing2D)
      max_pooling2d_171 (MaxPool (None, 1, 4, 128)
                                                             0
      ing2D)
      flatten_38 (Flatten)
                                   (None, 512)
      dense_76 (Dense)
                                   (None, 1024)
                                                             525312
      dropout_38 (Dropout)
                                  (None, 1024)
      dense 77 (Dense)
                                   (None, 12)
                                                             12300
     Total params: 1019852 (3.89 MB)
     Trainable params: 1018764 (3.89 MB)
     Non-trainable params: 1088 (4.25 KB)
[48]: task5_history = task5_model.fit(x_train, y_train, batch_size=BATCH_SIZE,__
       →epochs=30, validation_data=(x_val, y_val))
      # summarize history for accuracy
      plt.plot(task5_history.history['accuracy'])
      plt.plot(task5_history.history['val_accuracy'])
      plt.title('T5 - model accuracy')
      plt.ylabel('accuracy')
      plt.xlabel('epoch')
      plt.legend(['train', 'val'], loc='upper left')
      plt.show()
      # summarize history for loss
      plt.plot(task5_history.history['loss'])
      plt.plot(task5_history.history['val_loss'])
      plt.title('T5 - model loss')
      plt.ylabel('loss')
      plt.xlabel('epoch')
      plt.legend(['train', 'val'], loc='upper left')
      plt.show()
```

```
# Print accuracy
score = task5_model.evaluate(x_val, y_val, verbose=0)
print('T5 - validation accuracy:', score[1])
# Print confusion matrix
y_pred = task5_model.predict(x_val)
y_pred = np.argmax(y_pred, axis=1)
y_true = np.argmax(y_val, axis=1)
class_labels = list(val_ds.class_names)
confusion_mtx = confusion_matrix(y_true, y_pred)
plt.figure(figsize=(10, 8))
sns.heatmap(confusion_mtx, xticklabels=class_labels, yticklabels=class_labels,_u
 ⇔annot=True, fmt='d')
plt.xlabel('Prediction')
plt.ylabel('Label')
plt.title('T5 - Confusion Matrix')
plt.show()
Epoch 1/30
accuracy: 0.5892 - val_loss: 10.4304 - val_accuracy: 0.1126
Epoch 2/30
accuracy: 0.7014 - val_loss: 8.3976 - val_accuracy: 0.1135
Epoch 3/30
accuracy: 0.7934 - val_loss: 5.3865 - val_accuracy: 0.2049
Epoch 4/30
accuracy: 0.8492 - val_loss: 4.5652 - val_accuracy: 0.3920
Epoch 5/30
accuracy: 0.8927 - val_loss: 4.2441 - val_accuracy: 0.4809
accuracy: 0.9155 - val_loss: 3.6706 - val_accuracy: 0.5893
Epoch 7/30
accuracy: 0.9496 - val_loss: 3.2630 - val_accuracy: 0.6367
accuracy: 0.9614 - val_loss: 2.9644 - val_accuracy: 0.6630
Epoch 9/30
accuracy: 0.9758 - val_loss: 2.7778 - val_accuracy: 0.6461
Epoch 10/30
```

```
accuracy: 0.9773 - val_loss: 2.5560 - val_accuracy: 0.6757
Epoch 11/30
accuracy: 0.9867 - val_loss: 2.5759 - val_accuracy: 0.6342
Epoch 12/30
accuracy: 0.9891 - val_loss: 2.2815 - val_accuracy: 0.6749
Epoch 13/30
accuracy: 0.9926 - val_loss: 2.0652 - val_accuracy: 0.6969
Epoch 14/30
accuracy: 0.9931 - val_loss: 2.0963 - val_accuracy: 0.6765
Epoch 15/30
accuracy: 0.9901 - val_loss: 2.0541 - val_accuracy: 0.6545
Epoch 16/30
16/16 [============= ] - 2s 136ms/step - loss: 0.8689 -
accuracy: 0.9921 - val_loss: 1.7727 - val_accuracy: 0.7189
Epoch 17/30
accuracy: 0.9965 - val_loss: 1.8559 - val_accuracy: 0.6317
Epoch 18/30
accuracy: 0.9946 - val_loss: 1.7239 - val_accuracy: 0.6952
Epoch 19/30
accuracy: 0.9960 - val_loss: 1.6473 - val_accuracy: 0.7045
Epoch 20/30
accuracy: 0.9985 - val_loss: 1.6394 - val_accuracy: 0.6876
Epoch 21/30
accuracy: 0.9990 - val loss: 1.4870 - val accuracy: 0.7248
Epoch 22/30
accuracy: 0.9990 - val_loss: 1.3935 - val_accuracy: 0.7384
Epoch 23/30
accuracy: 0.9990 - val_loss: 1.3256 - val_accuracy: 0.7375
Epoch 24/30
accuracy: 0.9995 - val_loss: 1.2920 - val_accuracy: 0.7553
Epoch 25/30
accuracy: 0.9995 - val_loss: 1.2090 - val_accuracy: 0.7494
Epoch 26/30
```







As, we can observe: * The model is able to predict the 'yes' class with an accuracy of 0.8. * The model is able to predict the 'no' class with an accuracy of 0.6.

T5.2 Training the Advanced Model on the Personal Recordings: We now move the spectrograms to the training and validation directories and train the advanced model on the personal recordings.

We create new instance of x_train, y_train, x_val, y_val and train the advanced model on the personal recordings.

```
[50]: # Load the data
      train_ds = tf.keras.utils.image_dataset_from_directory(
          directory='speechImageData/TrainData',
          labels='inferred',
          color_mode="grayscale",
          label_mode='categorical',
          batch_size=BATCH_SIZE,
          image_size=IMG_SIZE
      )
      val_ds = tf.keras.utils.image_dataset_from_directory(
          directory='speechImageData/ValData',
          labels='inferred',
          color_mode="grayscale",
          label_mode='categorical',
          batch_size=BATCH_SIZE,
          image_size=IMG_SIZE
      )
      # Extract the training input images and output class labels
      x_train = []
      y_train = []
      for images, labels in train_ds.take(-1):
          x_train.append(images.numpy())
          y_train.append(labels.numpy())
      x_train_recrd = np.concatenate(x_train, axis=0)
      y_train_recrd = np.concatenate(y_train, axis=0)
      print(y_train_recrd)
      # Extract the validation input images and output class labels
      x_val = []
      y_val = []
```

```
for images, labels in val_ds.take(-1):
    x_val.append(images.numpy())
    y_val.append(labels.numpy())

x_val_recrd = np.concatenate(x_val, axis=0)
y_val_recrd = np.concatenate(y_val, axis=0)

print(y_val_recrd)
```

```
Found 2023 files belonging to 12 classes.

Found 1181 files belonging to 12 classes.

[[0. 0. 0. ... 1. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 1.]

...

[0. 0. 0. ... 1. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]

[1. 0. 0. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 0.]
```

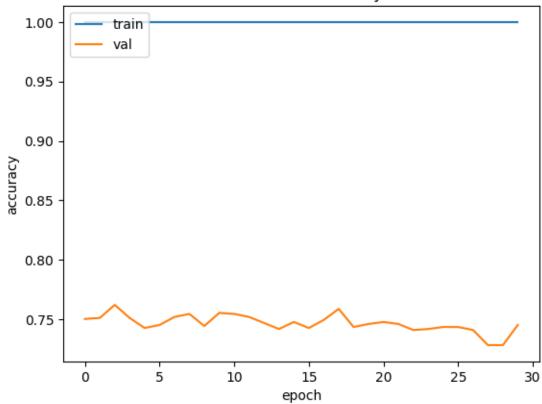
We now plot the training history and confusion matrix of the advanced model on the personal recordings.

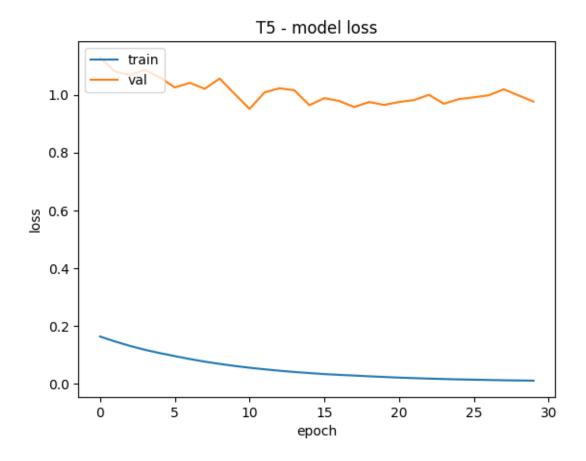
```
[51]: task5 history = task5 model.fit(x_train_recrd, y_train_recrd,__
       ⇔batch_size=BATCH_SIZE, epochs=30, validation_data=(x_val_recrd, y_val_recrd))
      # summarize history for accuracy
      plt.plot(task5 history.history['accuracy'])
      plt.plot(task5_history.history['val_accuracy'])
      plt.title('T5 - model accuracy')
      plt.ylabel('accuracy')
      plt.xlabel('epoch')
      plt.legend(['train', 'val'], loc='upper left')
      plt.show()
      # summarize history for loss
      plt.plot(task5_history.history['loss'])
      plt.plot(task5_history.history['val_loss'])
      plt.title('T5 - model loss')
      plt.ylabel('loss')
      plt.xlabel('epoch')
      plt.legend(['train', 'val'], loc='upper left')
```

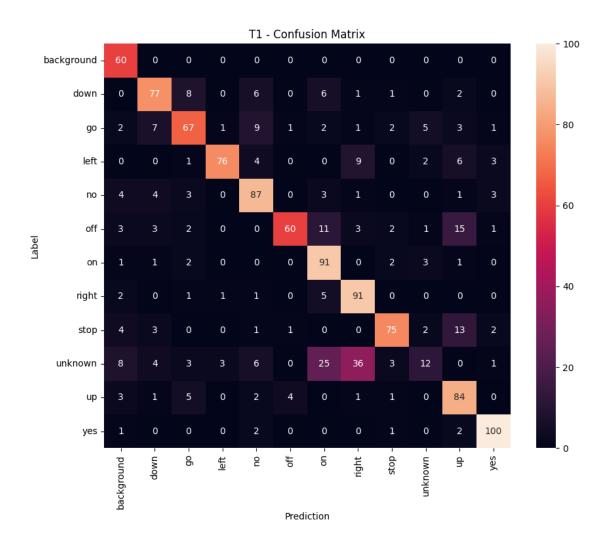
```
plt.show()
# Print accuracy
score = task5_model.evaluate(x_val_recrd, y_val_recrd, verbose=0)
print('T5 - validation accuracy:', score[1])
# Print confusion matrix
y_pred = task5_model.predict(x_val_recrd)
y pred = np.argmax(y pred, axis=1)
y_true = np.argmax(y_val_recrd, axis=1)
class_labels = list(val_ds.class_names)
confusion_mtx = confusion_matrix(y_true, y_pred)
plt.figure(figsize=(10, 8))
→annot=True, fmt='d')
plt.xlabel('Prediction')
plt.ylabel('Label')
plt.title('T1 - Confusion Matrix')
plt.show()
Epoch 1/30
accuracy: 1.0000 - val_loss: 1.1288 - val_accuracy: 0.7502
accuracy: 1.0000 - val_loss: 1.0800 - val_accuracy: 0.7511
accuracy: 1.0000 - val_loss: 1.0681 - val_accuracy: 0.7621
Epoch 4/30
16/16 [============ ] - 2s 134ms/step - loss: 0.1177 -
accuracy: 1.0000 - val_loss: 1.0858 - val_accuracy: 0.7511
Epoch 5/30
accuracy: 1.0000 - val_loss: 1.0604 - val_accuracy: 0.7426
Epoch 6/30
accuracy: 1.0000 - val loss: 1.0253 - val accuracy: 0.7451
Epoch 7/30
accuracy: 1.0000 - val_loss: 1.0414 - val_accuracy: 0.7519
Epoch 8/30
accuracy: 1.0000 - val_loss: 1.0206 - val_accuracy: 0.7544
Epoch 9/30
16/16 [============ ] - 2s 136ms/step - loss: 0.0690 -
accuracy: 1.0000 - val_loss: 1.0563 - val_accuracy: 0.7443
```

```
Epoch 10/30
accuracy: 1.0000 - val_loss: 1.0029 - val_accuracy: 0.7553
Epoch 11/30
accuracy: 1.0000 - val_loss: 0.9511 - val_accuracy: 0.7544
Epoch 12/30
accuracy: 1.0000 - val_loss: 1.0081 - val_accuracy: 0.7519
Epoch 13/30
accuracy: 1.0000 - val_loss: 1.0225 - val_accuracy: 0.7468
Epoch 14/30
accuracy: 1.0000 - val_loss: 1.0156 - val_accuracy: 0.7417
Epoch 15/30
accuracy: 1.0000 - val_loss: 0.9640 - val_accuracy: 0.7477
Epoch 16/30
accuracy: 1.0000 - val_loss: 0.9882 - val_accuracy: 0.7426
Epoch 17/30
accuracy: 1.0000 - val_loss: 0.9784 - val_accuracy: 0.7494
Epoch 18/30
accuracy: 1.0000 - val_loss: 0.9575 - val_accuracy: 0.7587
Epoch 19/30
accuracy: 1.0000 - val_loss: 0.9750 - val_accuracy: 0.7434
Epoch 20/30
accuracy: 1.0000 - val_loss: 0.9649 - val_accuracy: 0.7460
Epoch 21/30
accuracy: 1.0000 - val_loss: 0.9750 - val_accuracy: 0.7477
Epoch 22/30
accuracy: 1.0000 - val_loss: 0.9816 - val_accuracy: 0.7460
Epoch 23/30
accuracy: 1.0000 - val_loss: 1.0001 - val_accuracy: 0.7409
accuracy: 1.0000 - val_loss: 0.9689 - val_accuracy: 0.7417
Epoch 25/30
accuracy: 1.0000 - val_loss: 0.9849 - val_accuracy: 0.7434
```

T5 - model accuracy







As observed, there is an increased match in both the classes, 'yes' and 'no' with the model's predictions. There is no dramatic drop of accuracy in these classes, and hence the data augmentation has helped in improving the model's performance.

T5.3 Transfer Learning using the Advanced Model We now use the popular model mobilenetv2 to perform transfer learning on the advanced model. Since the model requires the image size to be either 96x96, 128x128, 224x224, we resize the images to 96x96, since our original image size is 98x50.

We also reimport our training and validation data as three channeled rgb images instead of grayscale.

```
color_mode="rgb",
    label_mode='categorical',
    batch_size=BATCH_SIZE,
    image_size=IMG_SIZE
)
val_ds = tf.keras.utils.image_dataset_from_directory(
    directory='speechImageData/ValData',
    labels='inferred',
    color_mode="rgb",
    label_mode='categorical',
    batch_size=BATCH_SIZE,
    image_size=IMG_SIZE
)
# Extract the training input images and output class labels
x_train = []
y_train = []
for images, labels in train_ds.take(-1):
    x_train.append(images.numpy())
    y_train.append(labels.numpy())
x_train_recrd = np.concatenate(x_train, axis=0)
y_train_recrd = np.concatenate(y_train, axis=0)
print(y_train_recrd)
# Extract the validation input images and output class labels
x_val = []
y_val = []
for images, labels in val_ds.take(-1):
    x_val.append(images.numpy())
    y_val.append(labels.numpy())
x_val_recrd = np.concatenate(x_val, axis=0)
y_val_recrd = np.concatenate(y_val, axis=0)
print(y_val_recrd)
Found 2023 files belonging to 12 classes.
Found 1181 files belonging to 12 classes.
[[0. 0. 0. ... 0. 1. 0.]
[0. 0. 1. ... 0. 0. 0.]
 [0. 0. 0. ... 1. 0. 0.]
 [0. 0. 0. ... 0. 0. 1.]
 [1. 0. 0. ... 0. 0. 0.]
```

```
[1. 0. 0. ... 0. 0. 0.]]
     [[0. 0. 0. ... 0. 0. 0.]
      [0. 0. 0. ... 1. 0. 0.]
      [0. 0. 0. ... 0. 0. 1.]
      [0. 0. 0. ... 0. 1. 0.]
      [0. 0. 0. ... 0. 0. 0.]
      [0. 0. 0. ... 0. 0. 0.]]
[56]: # mobilenet image size
     MOBILENET_IMG_SIZE = (96, 96)
     # resize the images
     x_train_recrd = tf.image.resize(x_train_recrd, (MOBILENET_IMG_SIZE[0],__
       →MOBILENET_IMG_SIZE[1]))
     x_val_recrd = tf.image.resize(x_val_recrd, (MOBILENET_IMG_SIZE[0],_
       →MOBILENET_IMG_SIZE[1]))
     base_model = MobileNetV2(input_shape=(MOBILENET_IMG_SIZE[0],_
       →MOBILENET_IMG_SIZE[1], 3), include_top=False, weights='imagenet')
     x_train_recrd = preprocess_input(x_train_recrd)
     x_val_recrd = preprocess_input(x_val_recrd)
     # extract features
     train_features = base_model.predict(x_train_recrd)
     val_features = base_model.predict(x_val_recrd)
     64/64 [======== ] - 2s 12ms/step
     37/37 [======== ] - Os 11ms/step
```

We now define the base model for training. We just append the fully connected layer to the base model and output layer to the model.

```
mobilenet_model = Sequential()
mobilenet_model.add(Flatten(input_shape=train_features.shape[1:]))
mobilenet_model.add(Dense(1024, kernel_regularizer=regularizers.12(0.1),
activation='relu'))
mobilenet_model.add(Dropout(0.2))
mobilenet_model.add(Dense(NUM_CLASSES, activation='softmax'))
mobilenet_model.compile(optimizer=optimizers.Adam(learning_rate=0.0001),
aloss='categorical_crossentropy', metrics=['accuracy'])
mobilenet_model.summary()
```

Epoch 1/100

T5.A - Model Training Now, we proceed to train the model with the training and validation features extracted by the base model.

```
[63]: mobilenet_model.history = mobilenet_model.fit(train_features, y_train_recrd,__ batch_size=BATCH_SIZE, epochs=100, validation_data=(val_features,__ y_val_recrd))
```

```
accuracy: 0.3381 - val_loss: 160.5809 - val_accuracy: 0.3404
Epoch 2/100
accuracy: 0.5971 - val_loss: 134.0095 - val_accuracy: 0.3903
Epoch 3/100
accuracy: 0.6945 - val_loss: 111.3102 - val_accuracy: 0.4039
Epoch 4/100
accuracy: 0.7904 - val_loss: 92.1280 - val_accuracy: 0.4479
Epoch 5/100
accuracy: 0.8492 - val_loss: 76.0616 - val_accuracy: 0.4335
accuracy: 0.8962 - val_loss: 62.7798 - val_accuracy: 0.4403
Epoch 7/100
accuracy: 0.9288 - val_loss: 51.8201 - val_accuracy: 0.4505
Epoch 8/100
16/16 [============== ] - Os 14ms/step - loss: 46.0271 -
accuracy: 0.9545 - val_loss: 42.7247 - val_accuracy: 0.4640
Epoch 9/100
accuracy: 0.9619 - val_loss: 35.2867 - val_accuracy: 0.4555
```

```
Epoch 10/100
accuracy: 0.9758 - val_loss: 29.2539 - val_accuracy: 0.4589
Epoch 11/100
accuracy: 0.9698 - val_loss: 24.2336 - val_accuracy: 0.4835
Epoch 12/100
accuracy: 0.9778 - val_loss: 20.1942 - val_accuracy: 0.4666
Epoch 13/100
accuracy: 0.9748 - val_loss: 16.9831 - val_accuracy: 0.4488
Epoch 14/100
accuracy: 0.9797 - val_loss: 14.2950 - val_accuracy: 0.4742
Epoch 15/100
accuracy: 0.9842 - val_loss: 12.1549 - val_accuracy: 0.4818
Epoch 16/100
0.9832 - val_loss: 10.4383 - val_accuracy: 0.4564
Epoch 17/100
0.9847 - val_loss: 9.0073 - val_accuracy: 0.4733
Epoch 18/100
0.9871 - val_loss: 7.8104 - val_accuracy: 0.4801
Epoch 19/100
16/16 [============= ] - Os 13ms/step - loss: 5.8249 - accuracy:
0.9827 - val_loss: 6.8671 - val_accuracy: 0.4970
Epoch 20/100
16/16 [============= ] - Os 13ms/step - loss: 4.9839 - accuracy:
0.9797 - val_loss: 6.1881 - val_accuracy: 0.4970
Epoch 21/100
0.9842 - val_loss: 5.6243 - val_accuracy: 0.4843
Epoch 22/100
0.9807 - val_loss: 5.0497 - val_accuracy: 0.5224
Epoch 23/100
0.9876 - val_loss: 4.6097 - val_accuracy: 0.4953
Epoch 24/100
16/16 [============= ] - Os 13ms/step - loss: 2.8151 - accuracy:
0.9891 - val_loss: 4.2320 - val_accuracy: 0.4843
Epoch 25/100
0.9862 - val_loss: 3.9228 - val_accuracy: 0.4903
```

```
Epoch 26/100
0.9921 - val_loss: 3.7327 - val_accuracy: 0.5038
Epoch 27/100
0.9871 - val_loss: 3.6404 - val_accuracy: 0.4581
Epoch 28/100
0.9748 - val_loss: 3.2967 - val_accuracy: 0.4936
Epoch 29/100
0.9713 - val_loss: 3.1933 - val_accuracy: 0.4784
Epoch 30/100
16/16 [============= ] - Os 17ms/step - loss: 1.5158 - accuracy:
0.9728 - val_loss: 3.0982 - val_accuracy: 0.4894
Epoch 31/100
16/16 [============= ] - Os 18ms/step - loss: 1.3894 - accuracy:
0.9758 - val_loss: 3.1291 - val_accuracy: 0.4530
Epoch 32/100
0.9812 - val_loss: 2.9828 - val_accuracy: 0.4632
Epoch 33/100
0.9857 - val_loss: 2.8020 - val_accuracy: 0.4852
Epoch 34/100
16/16 [============= ] - Os 18ms/step - loss: 1.0554 - accuracy:
0.9916 - val_loss: 2.6798 - val_accuracy: 0.5004
Epoch 35/100
16/16 [============= ] - Os 17ms/step - loss: 0.9742 - accuracy:
0.9886 - val_loss: 2.7112 - val_accuracy: 0.4860
Epoch 36/100
16/16 [============= ] - Os 14ms/step - loss: 0.9195 - accuracy:
0.9867 - val_loss: 2.6424 - val_accuracy: 0.4699
Epoch 37/100
0.9857 - val_loss: 2.5473 - val_accuracy: 0.5021
Epoch 38/100
0.9842 - val_loss: 2.5702 - val_accuracy: 0.4632
Epoch 39/100
0.9743 - val_loss: 2.5593 - val_accuracy: 0.4843
Epoch 40/100
0.9812 - val_loss: 2.5418 - val_accuracy: 0.4488
Epoch 41/100
16/16 [============= ] - Os 13ms/step - loss: 0.7808 - accuracy:
0.9639 - val_loss: 2.6410 - val_accuracy: 0.4767
```

```
Epoch 42/100
16/16 [============= ] - Os 13ms/step - loss: 0.7993 - accuracy:
0.9481 - val_loss: 2.7536 - val_accuracy: 0.4615
Epoch 43/100
0.9565 - val_loss: 2.7792 - val_accuracy: 0.4742
Epoch 44/100
0.9664 - val_loss: 2.5656 - val_accuracy: 0.4809
Epoch 45/100
0.9763 - val_loss: 2.4344 - val_accuracy: 0.4860
Epoch 46/100
16/16 [============= ] - Os 13ms/step - loss: 0.6151 - accuracy:
0.9857 - val_loss: 2.4452 - val_accuracy: 0.4860
Epoch 47/100
16/16 [============= ] - Os 13ms/step - loss: 0.5920 - accuracy:
0.9827 - val_loss: 2.4310 - val_accuracy: 0.4860
Epoch 48/100
0.9797 - val_loss: 2.3777 - val_accuracy: 0.4632
Epoch 49/100
0.9669 - val_loss: 2.4848 - val_accuracy: 0.4716
Epoch 50/100
0.9600 - val_loss: 2.4090 - val_accuracy: 0.4835
Epoch 51/100
16/16 [============= ] - Os 13ms/step - loss: 0.5877 - accuracy:
0.9689 - val_loss: 2.6146 - val_accuracy: 0.4750
Epoch 52/100
16/16 [============= ] - Os 15ms/step - loss: 0.5909 - accuracy:
0.9679 - val_loss: 2.3782 - val_accuracy: 0.4852
Epoch 53/100
0.9718 - val_loss: 2.4317 - val_accuracy: 0.4623
Epoch 54/100
0.9713 - val_loss: 2.4058 - val_accuracy: 0.4818
Epoch 55/100
0.9708 - val_loss: 2.3547 - val_accuracy: 0.4776
0.9689 - val_loss: 2.3681 - val_accuracy: 0.4640
Epoch 57/100
16/16 [============= ] - Os 13ms/step - loss: 0.5425 - accuracy:
0.9664 - val_loss: 2.3271 - val_accuracy: 0.4843
```

```
Epoch 58/100
0.9664 - val_loss: 2.3809 - val_accuracy: 0.4767
Epoch 59/100
0.9733 - val_loss: 2.5178 - val_accuracy: 0.4843
Epoch 60/100
0.9664 - val_loss: 2.4543 - val_accuracy: 0.4649
Epoch 61/100
0.9674 - val_loss: 2.3922 - val_accuracy: 0.4809
Epoch 62/100
16/16 [============= ] - Os 14ms/step - loss: 0.5156 - accuracy:
0.9708 - val_loss: 2.4112 - val_accuracy: 0.4835
Epoch 63/100
16/16 [============= ] - Os 14ms/step - loss: 0.5285 - accuracy:
0.9654 - val_loss: 2.4383 - val_accuracy: 0.4496
Epoch 64/100
0.9585 - val_loss: 2.4447 - val_accuracy: 0.4996
Epoch 65/100
0.9609 - val_loss: 2.5964 - val_accuracy: 0.4649
Epoch 66/100
16/16 [============= ] - Os 15ms/step - loss: 0.5701 - accuracy:
0.9511 - val_loss: 2.5188 - val_accuracy: 0.4674
Epoch 67/100
16/16 [============= ] - Os 13ms/step - loss: 0.5355 - accuracy:
0.9654 - val_loss: 2.3874 - val_accuracy: 0.5055
Epoch 68/100
16/16 [============= ] - Os 13ms/step - loss: 0.4868 - accuracy:
0.9862 - val_loss: 2.4060 - val_accuracy: 0.4615
Epoch 69/100
0.9733 - val_loss: 2.4249 - val_accuracy: 0.4632
Epoch 70/100
0.9605 - val_loss: 2.3747 - val_accuracy: 0.4682
Epoch 71/100
0.9649 - val_loss: 2.4551 - val_accuracy: 0.4801
Epoch 72/100
0.9694 - val_loss: 2.3764 - val_accuracy: 0.4860
Epoch 73/100
16/16 [============= ] - Os 14ms/step - loss: 0.4658 - accuracy:
0.9787 - val_loss: 2.2814 - val_accuracy: 0.5080
```

```
Epoch 74/100
0.9787 - val_loss: 2.2893 - val_accuracy: 0.4869
Epoch 75/100
0.9609 - val_loss: 2.3356 - val_accuracy: 0.4877
Epoch 76/100
0.9639 - val_loss: 2.2954 - val_accuracy: 0.4776
Epoch 77/100
0.9525 - val_loss: 2.4062 - val_accuracy: 0.4691
Epoch 78/100
16/16 [============= ] - Os 14ms/step - loss: 0.5115 - accuracy:
0.9605 - val_loss: 2.3608 - val_accuracy: 0.4877
Epoch 79/100
16/16 [============= ] - Os 13ms/step - loss: 0.4858 - accuracy:
0.9664 - val_loss: 2.2733 - val_accuracy: 0.4801
Epoch 80/100
0.9753 - val_loss: 2.3907 - val_accuracy: 0.4649
Epoch 81/100
0.9713 - val_loss: 2.4049 - val_accuracy: 0.4733
Epoch 82/100
16/16 [============= ] - Os 16ms/step - loss: 0.4673 - accuracy:
0.9718 - val_loss: 2.3584 - val_accuracy: 0.4649
Epoch 83/100
16/16 [============= ] - Os 17ms/step - loss: 0.4798 - accuracy:
0.9664 - val_loss: 2.4860 - val_accuracy: 0.4809
Epoch 84/100
16/16 [============= ] - Os 17ms/step - loss: 0.4653 - accuracy:
0.9773 - val_loss: 2.3721 - val_accuracy: 0.4564
Epoch 85/100
0.9783 - val_loss: 2.3558 - val_accuracy: 0.4530
Epoch 86/100
0.9456 - val_loss: 2.5672 - val_accuracy: 0.4623
Epoch 87/100
0.9471 - val_loss: 2.4081 - val_accuracy: 0.4776
16/16 [============= ] - Os 16ms/step - loss: 0.5174 - accuracy:
0.9614 - val_loss: 2.3689 - val_accuracy: 0.4674
Epoch 89/100
16/16 [============= ] - Os 13ms/step - loss: 0.4801 - accuracy:
0.9718 - val_loss: 2.4201 - val_accuracy: 0.5021
```

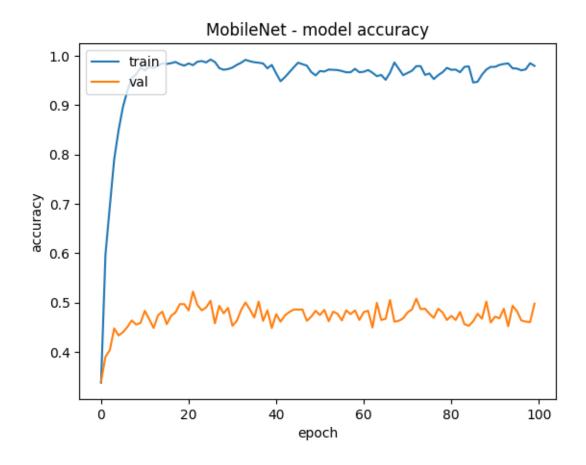
```
Epoch 90/100
0.9773 - val_loss: 2.3257 - val_accuracy: 0.4598
Epoch 91/100
0.9773 - val_loss: 2.3323 - val_accuracy: 0.4716
Epoch 92/100
0.9812 - val_loss: 2.2895 - val_accuracy: 0.4682
Epoch 93/100
0.9832 - val_loss: 2.2334 - val_accuracy: 0.4877
Epoch 94/100
0.9842 - val_loss: 2.3693 - val_accuracy: 0.4522
Epoch 95/100
0.9743 - val_loss: 2.3085 - val_accuracy: 0.4936
Epoch 96/100
0.9738 - val_loss: 2.4025 - val_accuracy: 0.4826
Epoch 97/100
16/16 [============= ] - Os 13ms/step - loss: 0.4512 - accuracy:
0.9703 - val_loss: 2.2379 - val_accuracy: 0.4640
Epoch 98/100
16/16 [============= ] - Os 13ms/step - loss: 0.4471 - accuracy:
0.9723 - val_loss: 2.2590 - val_accuracy: 0.4615
Epoch 99/100
0.9847 - val_loss: 2.3434 - val_accuracy: 0.4606
Epoch 100/100
0.9792 - val_loss: 2.2964 - val_accuracy: 0.4979
```

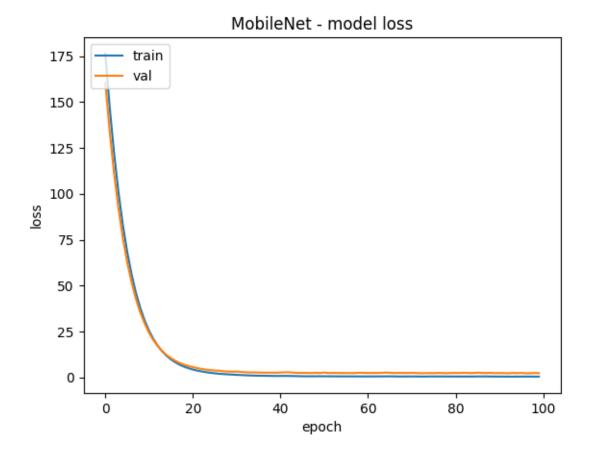
T5.B - Plot Training History and Confusion Matrix Here, we plot the training history and confusion matrix of the advanced model with transfer learning.

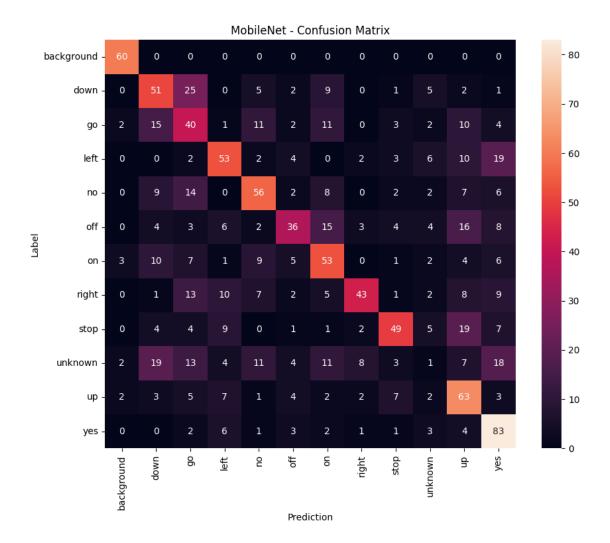
```
[64]: # summarize history for accuracy
plt.plot(mobilenet_model.history.history['accuracy'])
plt.plot(mobilenet_model.history.history['val_accuracy'])
plt.title('MobileNet - model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'val'], loc='upper left')
plt.show()

# summarize history for loss
```

```
plt.plot(mobilenet_model.history.history['loss'])
plt.plot(mobilenet_model.history.history['val_loss'])
plt.title('MobileNet - model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'val'], loc='upper left')
plt.show()
# Print accuracy
score = mobilenet_model.evaluate(val_features, y_val_recrd, verbose=0)
print('MobileNet - validation accuracy:', score[1])
# Print confusion matrix
y_pred = mobilenet_model.predict(val_features)
y_pred = np.argmax(y_pred, axis=1)
y_true = np.argmax(y_val_recrd, axis=1)
class_labels = list(val_ds.class_names)
confusion_mtx = confusion_matrix(y_true, y_pred)
plt.figure(figsize=(10, 8))
sns.heatmap(confusion_mtx, xticklabels=class_labels, yticklabels=class_labels,_u
 ⇔annot=True, fmt='d')
plt.xlabel('Prediction')
plt.ylabel('Label')
plt.title('MobileNet - Confusion Matrix')
plt.show()
```







T5.C - Conclusion The transfer learning model has a poor accuracy of about 0.5. This is because the model is not able to learn the features of the personal recordings well due to the resizing constraint of the model. The resizing jump from 98x50 to 96x96 has caused the model to lose the features of the spectrograms, and hence the model is not able to predict the classes well.

Future steps to improve the model would be taking inspiration of the model's architecture in the transfer learning model and training a new model with the same architecture as the base model without the resizing constraint.

0.2 Overall Conclusion

The advanced model has an accuracy of about 0.84, which is better than the baseline model's accuracy of 0.6. The advanced model is able to predict the classes of the personal recordings with an accuracy of 0.8 for the 'yes' class and 0.6 for the 'no' class.

Additionally, the transfer learning model has a poor accuracy of about 0.5. This is because the model is not able to learn the features of the personal recordings well due to the resizing constraint

of the model.