

Implementation of HPC for Music Genres Classification

for music recommendation systems

DSC 520-01: High Perform Scientific Compute

NIKHILESH NARKHEDE (02196141)



Problem Statement

Music genres: Build a DL model to classify music into the top 10 genres.

Audio Data — Time series data.

If we go with traditional time series architecture like RNN, the LSTM model turns out to be bulky and faces problems like vanishing gradient.

But we can convert Audio data in image format using a spectrogram with a library like Librosa.

Its representation of the spectrum of frequency in sound recording as they vary over time.

Dark Area: Low-intensity frequency.

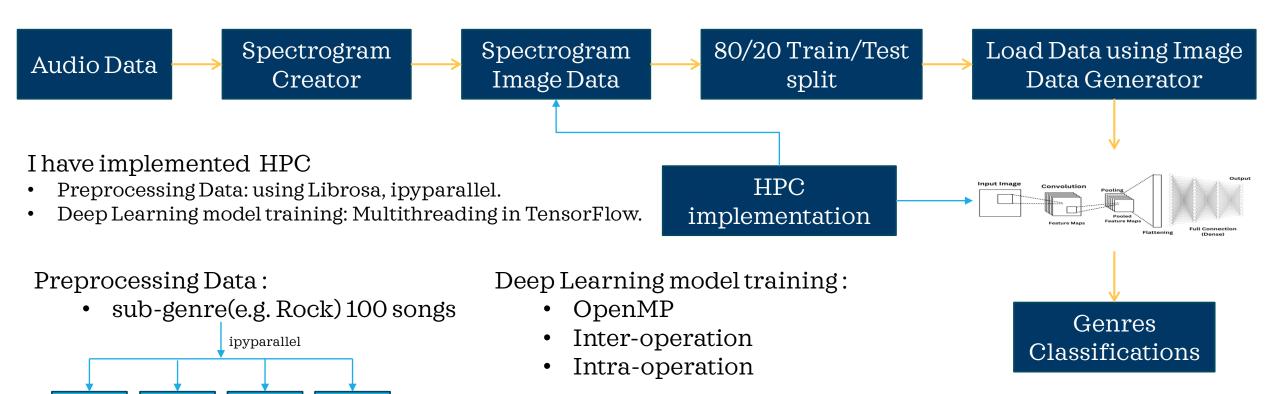
Orange and yellow Areas: High intensities frequency.

https://academo.org/demos/spectrum-analyzer/





Architecture



25 songs for each worker.

Worker 2

Worker 4

Worker 1



ipyparallel for Distributed Computing

What is ipyparallel?

• A Python library that enables parallel task execution using multiple workers, ideal for distributed computing.

Key Components:

- Controller: Distributes tasks and maintains the state.
- **Engines:** Execute tasks independently.

Example Use Case:

• Parallel processing of audio files into spectrogram images to reduce computation time.

Benefits:

• Improves efficiency, scales to multi-core systems, and integrates seamlessly with Python and Jupyter.

Deep Learning Model Training Optimization

OpenMP

- •OpenMP optimizes CPU-based computations by distributing work across multiple cores.
- •Usage in this Model:
- $\hbox{-} Convolution \, layers \, (Conv2D) \, and \, matrix \, operations \, in \, fully \, connected \, layers \, (Dense) \, are \, parallelized. \\$
- •OpenMP ensures that operations like matrix multiplications for the filters in the convolution layers and the weight updates in the dense layers are executed across multiple CPU threads.

Inter-Operation Parallelism

- •Manages parallel execution across different operations in the computational graph.
- \bullet If two Conv2D layers are computationally independent (e.g., during forward propagation), TensorFlow can process them concurrently.
- $\bullet Pooling \ layers \ and \ dropout \ calculations \ can \ also \ be \ executed \ in \ parallel \ with \ certain independent \ layers.$

Intra-Operation Parallelism

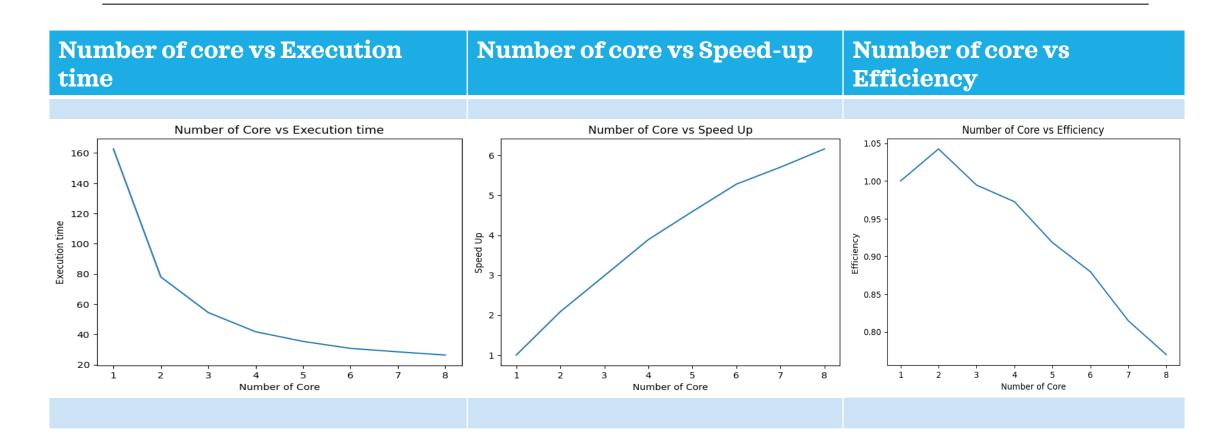
- •Optimizes parallelism within a single operation.
- •Each Conv2D operation (a matrix multiplication between input data and convolution filters) is broken into smaller tasks.
- •Pooling and activation computations (like ReLU) for large feature maps are divided among threads.

Key Benefits

- •Improved efficiency and reduced execution time.
- •Scalability across multi-core systems.
- $\bullet Better\,utilization\,of\,computational\,resources.$

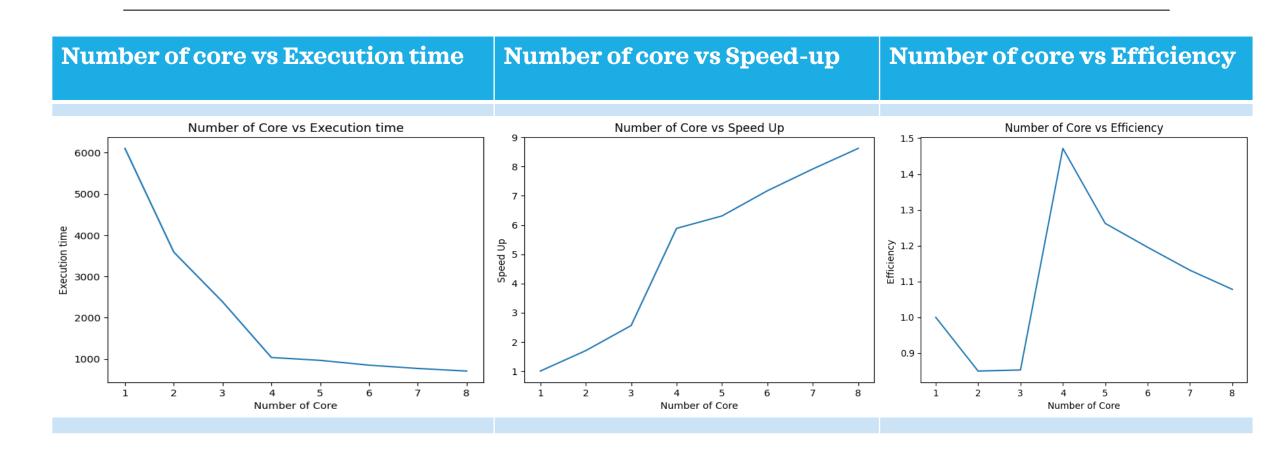


Result (Preprocessing)





Result (DL Model Training)





Conclusion

Preprocessing Data:

- 2 workers are most preferable in this context if efficiency is given more weight
- · 4 workers are most preferable in this context if execution time is given more weight

Deep Learning Model Training:

• 4 Multithreading (Worker) is most preferable in this context in terms of both execution time and efficiency.

Appendix

Import Lib

```
import os
from glob import glob
import librosa
import librosa.display
import numpy as np
import matplotlib.pyplot as plt
import ipyparallel as ipp
import time
from warnings import filterwarnings
filterwarnings('ignore')
```

Function to create a spectrogram from an audio file

```
In [2]: def create_spectrogram(filename, name, store_path):
            plt.interactive(False)
            clip, sample_rate = librosa.load(filename, sr=None)
            fig = plt.figure(figsize=[0.72, 0.72])
            ax = fig.add_subplot(111)
            ax.axes.get_xaxis().set_visible(False)
            ax.axes.get_yaxis().set_visible(False)
            ax.set_frame_on(False)
            S = librosa.feature.melspectrogram(y=clip, sr=sample_rate)
            librosa.display.specshow(librosa.power_to_db(S, ref=np.max))
            filename = os.path.join(store_path, name + '.jpg')
            plt.savefig(filename, dpi=400, bbox_inches='tight', pad_inches=0)
            plt.close()
            fig.clf()
            plt.close(fig)
            plt.close('all')
            del filename, name, clip, sample_rate, fig, ax, S
```

Helper function to process a single audio file

```
In [3]: def process_audio_file(file, folder_path):
    name = os.path.basename(file)
    create_spectrogram(file, name, folder_path)
```

Convert Audio files to spectrogram JPG files with ipyparallel

```
In [4]: def Convert_Audio_File_to_jpg_file(filename, store_filepath, number_of_worker):
            Convert audio files to spectrogram JPG files using ipyparallel.
            Parameters:
               filename (str): Path to the main directory containing audio files and number of worker.
            # Set up ipyparallel cluster
            Execution_time = 0
            rc = ipp.Cluster(n=number_of_worker).start_and_connect_sync() # Start and connect cluster with 4 workers
            dview = rc[:] # Create a direct view to all workers
            dview.push({"create_spectrogram": create_spectrogram})
            dview.push({"process_audio_file": process_audio_file})
            dview.execute("import os")
            dview.execute("import librosa")
            dview.execute("import librosa.display")
            dview.execute("import numpy as np")
            dview.execute("import matplotlib.pyplot as plt")
            dview.execute("import time")
            start_time = time.time()
            # Make a dictionary for given class and their file path names
            file_list = glob(os.path.join(filename, "*"))
            file_dic = {}
            for file in file_list:
                all_files = []
```

```
for root, dirs, files in os.walk(file):
        for file_ in files:
            # Join the root directory with the file name to get the full path
            all_files.append(os.path.join(root, file_))
    file_dic[file] = all_files
# Create file directories to store the converted audio files into JPG
file_path = []
for folder in file_dic.keys():
    folder_rot = os.path.join(store_filepath, os.path.basename(folder))
   file_path.append(folder_rot)
    os.makedirs(folder_rot, exist_ok=True)
# Delegate work to workers using dview.map_sync()
for i, folder in enumerate(file_dic.keys()):
   music_files = file_dic[folder]
    folder_path = file_path[i]
    # Map the processing function across music files with ipyparallel
    dview.map_sync(lambda file: process_audio_file(file, folder_path), music_files)
end_time = time.time()
elapsed_time = end_time - start_time
Execution_time = elapsed_time
print(f"Execution time: {elapsed_time:.2f} seconds")
# Shut down the cluster after processing
rc.shutdown()
return Execution_time
```

Convert Audio File to jpg file

Experiment with Core

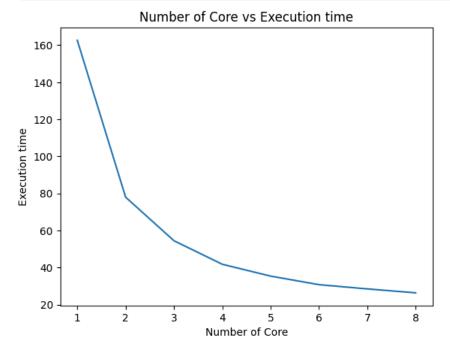
| 0/5 [00:00<?, ?engine/s]

```
In [6]: Execution_time = []
        Cores = range(1,9)
        for number of worker in Cores:
           Execution_time.append(Convert_Audio_File_to_jpg_file(Audio_Data_Path, Spectro_jpg_Path, number_of_worker = number_of_worker))
       Starting 1 engines with <class 'ipyparallel.cluster.launcher.LocalEngineSetLauncher'>
                      | 0/1 [00:00<?, ?engine/s]
       Execution time: 162.66 seconds
       engine set stopped 1733365915: {'engines': {'0': {'exit_code': 0, 'pid': 14416, 'identifier': '0'}}, 'exit_code': 0}
       Starting 2 engines with <class 'ipyparallel.cluster.launcher.LocalEngineSetLauncher'>
                      | 0/2 [00:00<?, ?engine/s]
        0% l
       Execution time: 78.02 seconds
       Failed to remove C:\Users\nikhi\.ipython\profile_default\log\ipengine-1733366084-vstu-1733366085-1.log: [WinError 5] Access is denied:
       'C:\\Users\\nikhi\\.ipython\\profile_default\\log\\ipengine-1733366084-vstu-1733366085-1.log'
       engine set stopped 1733366085: {'engines': {'0': {'exit_code': 0, 'pid': 21460, 'identifier': '0'}, '1': {'exit_code': 0, 'pid': 27340,
       'identifier': '1'}}, 'exit_code': 0}
      Starting 3 engines with <class 'ipyparallel.cluster.launcher.LocalEngineSetLauncher'>
        0%|
                      | 0/3 [00:00<?, ?engine/s]
       Execution time: 54.51 seconds
       engine set stopped 1733366171: {'engines': {'2': {'exit_code': 0, 'pid': 25728, 'identifier': '2'}, '1': {'exit_code': 0, 'pid': 20336,
       'identifier': '1'}, '0': {'exit_code': 0, 'pid': 25376, 'identifier': '0'}}, 'exit_code': 0}
       Starting 4 engines with <class 'ipyparallel.cluster.launcher.LocalEngineSetLauncher'>
                     | 0/4 [00:00<?, ?engine/s]
       Execution time: 41.81 seconds
       engine set stopped 1733366233: {'engines': {'0': {'exit_code': 0, 'pid': 18900, 'identifier': '0'}, '2': {'exit_code': 0, 'pid': 20572,
       'identifier': '2'}, '3': {'exit_code': 0, 'pid': 22864, 'identifier': '3'}, '1': {'exit_code': 0, 'pid': 23844, 'identifier': '1'}}, 'e
       xit code': 0}
       Starting 5 engines with <class 'ipyparallel.cluster.launcher.LocalEngineSetLauncher'>
```

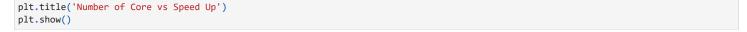
```
Execution time: 35.43 seconds
engine set stopped 1733366282: {'engines': {'3': {'exit_code': 0, 'pid': 25924, 'identifier': '3'}, '0': {'exit_code': 0, 'pid': 24580,
'identifier': '0'}, '2': {'exit_code': 0, 'pid': 27620, 'identifier': '2'}, '1': {'exit_code': 0, 'pid': 21080, 'identifier': '1'},
'4': {'exit_code': 0, 'pid': 19184, 'identifier': '4'}}, 'exit_code': 0}
Starting 6 engines with <class 'ipyparallel.cluster.launcher.LocalEngineSetLauncher'>
               | 0/6 [00:00<?, ?engine/s]
Execution time: 30.82 seconds
engine set stopped 1733366326: {'engines': {'2': {'exit_code': 0, 'pid': 20140, 'identifier': '2'}, '0': {'exit_code': 0, 'pid': 23164,
'identifier': '0'}, '1': {'exit_code': 0, 'pid': 11856, 'identifier': '1'}, '5': {'exit_code': 0, 'pid': 26344, 'identifier': '5'},
'4': {'exit_code': 0, 'pid': 2152, 'identifier': '4'}, '3': {'exit_code': 0, 'pid': 27152, 'identifier': '3'}}, 'exit_code': 0}
Starting 7 engines with <class 'ipyparallel.cluster.launcher.LocalEngineSetLauncher'>
 0%|
               | 0/7 [00:00<?, ?engine/s]
Execution time: 28.52 seconds
Failed to remove C:\Users\nikhi\.ipython\profile_default\log\ipengine-1733366363-hjat-1733366364-2.log: [WinError 2] The system cannot
find the file specified: 'C:\\Users\\nikhi\\.ipython\\profile_default\\log\\ipengine-1733366363-hjat-1733366364-2.log'
Failed to remove C:\Users\nikhi\.ipython\profile_default\log\ipengine-1733366363-hjat-1733366364-3.log: [WinError 32] The process canno
t access the file because it is being used by another process: 'C:\\Users\\nikhi\\.ipython\\profile_default\\log\\ipengine-1733366363-h
iat-1733366364-3.log'
engine set stopped 1733366364: {'engines': {'1': {'exit_code': 0, 'pid': 23124, 'identifier': '1'}, '2': {'exit_code': 0, 'pid': 23284,
'identifier': '2'}, '3': {'exit_code': 0, 'pid': 10412, 'identifier': '3'}, '4': {'exit_code': 0, 'pid': 11464, 'identifier': '4'},
'5': {'exit_code': 0, 'pid': 1996, 'identifier': '5'}, '0': {'exit_code': 0, 'pid': 13308, 'identifier': '0'}, '6': {'exit_code': 0, 'p
id': 26908, 'identifier': '6'}}, 'exit_code': 0}
Starting 8 engines with <class 'ipyparallel.cluster.launcher.LocalEngineSetLauncher'>
               | 0/8 [00:00<?, ?engine/s]
Execution time: 26.41 seconds
Failed to remove C:\Users\nikhi\.ipython\profile_default\log\ipengine-1733366400-7yh5-1733366401-1.log: [WinError 2] The system cannot
find the file specified: 'C:\\Users\\nikhi\\.ipython\\profile_default\\log\\ipengine-1733366400-7yh5-1733366401-1.log'
Failed to remove C:\Users\nikhi\.ipython\profile_default\log\ipengine-1733366400-7yh5-1733366401-6.log: [WinError 32] The process canno
t access the file because it is being used by another process: 'C:\\Users\\nikhi\\.ipython\\profile_default\\log\\ipengine-1733366400-7
yh5-1733366401-6.log'
engine set stopped 1733366401: {'engines': {'0': {'exit_code': 0, 'pid': 24628, 'identifier': '0'}, '1': {'exit_code': 0, 'pid': 14832,
'identifier': '1'}, '7': {'exit_code': 0, 'pid': 15012, 'identifier': '7'}, '5': {'exit_code': 0, 'pid': 27244, 'identifier': '5'}, '3': {'exit_code': 0, 'pid': 15744, 'identifier': '3'}, '6': {'exit_code': 0, 'pid': 11516, 'identifier': '6'}, '4': {'exit_code': 0,
'pid': 4368, 'identifier': '4'}, '2': {'exit_code': 0, 'pid': 18236, 'identifier': '2'}}, 'exit_code': 0}
```

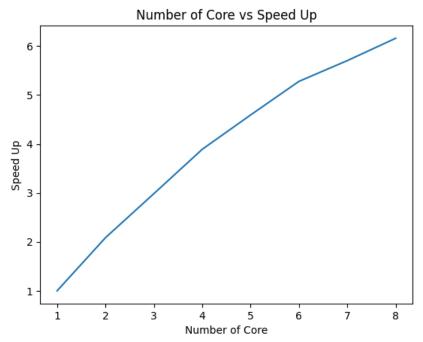
Number of Core vs Execution time

```
In [7]: plt.plot(Cores, Execution_time)
  plt.xlabel('Number of Core')
  plt.ylabel('Execution time')
  plt.title('Number of Core vs Execution time')
  plt.show()
```

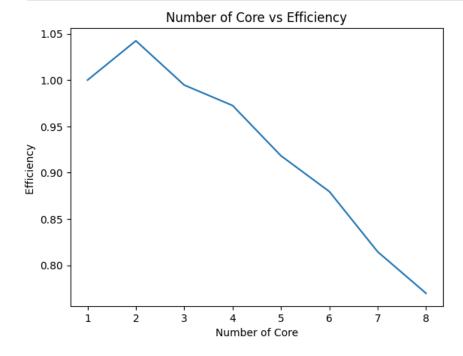


Number of Core vs Speed Up





Number of Core vs Efficiency

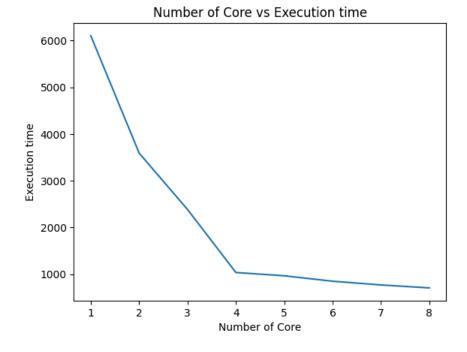


In []:

```
In [1]: import matplotlib.pyplot as plt
         import pandas as pd
In [2]: data = pd.read_csv(r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DEEP LEARNING WITH HPSC\core_data.txt")
In [3]: data.sort_values(by = 'Number of Cores', inplace = True,ignore_index= True)
In [4]: data
Out[4]:
           Number of Cores Elapsed Time
         0
                              6102.521532
                              3590.608773
         1
         2
                          3
                              2385.136909
         3
                              1036.823782
         4
                          5
                               967.008510
         5
                               850.766087
         6
                               770.620165
                               707.659396
```

Number of Core vs Execution time

```
In [5]: plt.plot(data['Number of Cores'], data['Elapsed Time'])
  plt.xlabel('Number of Core')
  plt.ylabel('Execution time')
  plt.title('Number of Core vs Execution time')
  plt.show()
```

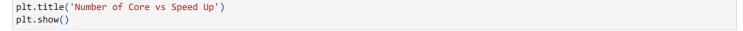


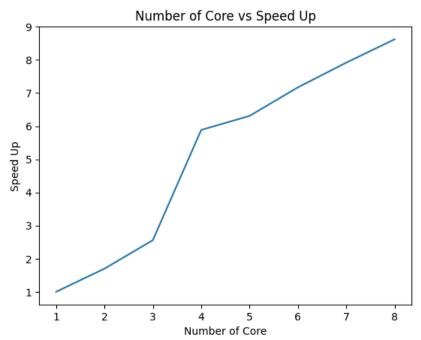
Number of Core vs Speed Up

```
In [6]: data['Elapsed Time'][7]
Out[6]: 707.6593961715698

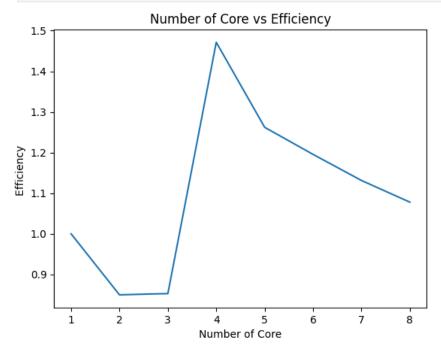
In [7]: speed_up = []
    for i in data['Number of Cores']:
        speed_up.append(data['Elapsed Time'][0] / data['Elapsed Time'][i - 1])

    plt.plot(data['Number of Cores'], speed_up)
    plt.xlabel('Number of Core')
    plt.ylabel('Speed Up')
```





Number of Core vs Efficiency



In []:

Import Req Lib

```
In [1]: %matplotlib inline
        import shutil
        import random
        import numpy as np
        from warnings import filterwarnings
        filterwarnings('ignore')
        from tensorflow.keras import layers, regularizers, optimizers
        from tensorflow.keras import models
        from tensorflow.keras.models import Sequential, Model
        from tensorflow.keras.layers import LeakyReLU,Dense, Activation, Flatten, Dropout, BatchNormalization,Conv2D, MaxPooling2D
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import tensorflow as tf
        import os
        import time
        import csv
        from matplotlib import figure
```

Define 1 worker

```
In [2]: # Set the number of threads
number_of_worker = 1
os.environ['OMP_NUM_THREADS'] = '1' # OpenMP threads for parallelism
os.environ['TF_NUM_INTEROP_THREADS'] = '1' # Threads for inter-operation parallelism
os.environ['TF_NUM_INTRAOP_THREADS'] = '1' # Threads for intra-operation parallelism
# Confirm TensorFlow is using the specified number of threads
tf.config.threading.set_inter_op_parallelism_threads(number_of_worker)
tf.config.threading.set_intra_op_parallelism_threads(number_of_worker)
```

Train Val data Split

```
In [3]: source_dir = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DATA\Convert_Audio_File_to_jpg_file"
        target_dir = r'genres_train_val_split_data'
        split_ratio = 0.8
        def Train_Test_Split(source_dir,target_dir,split_ratio):
            # Define source and target directories
            train_dir = os.path.join(target_dir, 'train')
            val_dir = os.path.join(target_dir, 'val')
            # Create target directories if they don't exist
            os.makedirs(train_dir, exist_ok=True)
            os.makedirs(val_dir, exist_ok=True)
            # Get the list of class directories
            classes = [d for d in os.listdir(source_dir) if os.path.isdir(os.path.join(source_dir, d))]
            for class name in classes:
                # Create class directories in train and val folders
                os.makedirs(os.path.join(train_dir, class_name), exist_ok=True)
                os.makedirs(os.path.join(val_dir, class_name), exist_ok=True)
                # Get list of images in the class directory
                class_dir = os.path.join(source_dir, class_name)
                images = [f for f in os.listdir(class_dir) if os.path.isfile(os.path.join(class_dir, f))]
                # Shuffle the images
                random.shuffle(images)
                # Compute the split point
                split_point = int(len(images) * split_ratio)
                # Split the images into training and validation sets
                train images = images[:split point]
                val_images = images[split_point:]
```

In [4]: Train_Test_Split(source_dir,target_dir,split_ratio)

Data split completed successfully!

Load the Data

```
In [5]: WIDTH = 64
        HEIGHT = 64
        BATCH_SIZE = 32
        TRAIN_DIR=r'genres_train_val_split_data/train'
        val_dir = r'genres_train_val_split_data/val'
        # data prep
        train_datagen = ImageDataGenerator(
            rescale=1./255.,validation_split=0.25)
        train_generator = train_datagen.flow_from_directory(
            TRAIN_DIR,
            target size=(HEIGHT, WIDTH),
                batch_size=BATCH_SIZE,
                class_mode='categorical')
        validation_gen = train_datagen.flow_from_directory(
            val_dir,target_size = (HEIGHT,WIDTH),
            batch_size = BATCH_SIZE,
            class_mode = 'categorical'
```

Found 800 images belonging to 10 classes. Found 200 images belonging to 10 classes.

Model Architecture

```
In [6]: model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                         input_shape=(64,64,3)))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(128, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(128, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(10, activation='softmax'))
        model.compile(optimizers.RMSprop(learning_rate=0.0005, decay=1e-6),loss="categorical_crossentropy",metrics=["accuracy"])
        model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 64, 64, 32)	896
activation (Activation)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 62, 62, 64)	18,496
activation_1 (Activation)	(None, 62, 62, 64)	0
max_pooling2d (MaxPooling2D)	(None, 31, 31, 64)	0
dropout (Dropout)	(None, 31, 31, 64)	0
conv2d_2 (Conv2D)	(None, 31, 31, 64)	36,928
activation_2 (Activation)	(None, 31, 31, 64)	0
conv2d_3 (Conv2D)	(None, 29, 29, 64)	36,928
activation_3 (Activation)	(None, 29, 29, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_1 (Dropout)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 14, 14, 128)	73,856
activation_4 (Activation)	(None, 14, 14, 128)	0
conv2d_5 (Conv2D)	(None, 12, 12, 128)	147,584
activation_5 (Activation)	(None, 12, 12, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_2 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2,359,808
activation_6 (Activation)	(None, 512)	0
dropout_3 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5,130

Total params: 2,679,626 (10.22 MB)

Trainable params: 2,679,626 (10.22 MB)

Non-trainable params: 0 (0.00 B)

```
In [7]: STEP_SIZE_TRAIN=train_generator.n//train_generator.batch_size
# Measure the execution time
start_time = time.time()

model.fit(train_generator,validation_data=validation_gen,epochs=200)

end_time = time.time()
elapsed_time = end_time - start_time
```

Epoch 25/25		32s 1s/step - accuracy: 0.1023 - loss: 2.3116 - val_accuracy: 0.1750 - val_loss: 2.2917
Epoch	2/200	
25/25 Epoch		28s 1s/step - accuracy: 0.1221 - loss: 2.2753 - val_accuracy: 0.2100 - val_loss: 2.1145
25/25		28s 1s/step - accuracy: 0.1855 - loss: 2.2138 - val_accuracy: 0.1950 - val_loss: 2.1203
25/25	4/200	28s 1s/step - accuracy: 0.1924 - loss: 2.1035 - val_accuracy: 0.2450 - val_loss: 1.9382
Epoch 25/25		27s 1s/step - accuracy: 0.2253 - loss: 2.0658 - val_accuracy: 0.2900 - val_loss: 1.9680
Epoch	6/200	
25/25 Epoch		26s 1s/step - accuracy: 0.2518 - loss: 2.0428 - val_accuracy: 0.3600 - val_loss: 2.0611
25/25 Epoch	8/200	26s 1s/step - accuracy: 0.2844 - loss: 1.9486 - val_accuracy: 0.3550 - val_loss: 1.8568
25/25		26s 1s/step - accuracy: 0.3084 - loss: 1.8901 - val_accuracy: 0.2650 - val_loss: 2.1513
25/25	9/200	26s 1s/step - accuracy: 0.3282 - loss: 1.9096 - val_accuracy: 0.3850 - val_loss: 1.7727
Epoch 25/25	10/200	26s 1s/step - accuracy: 0.3618 - loss: 1.7941 - val_accuracy: 0.4300 - val_loss: 1.6640
	11/200	
25/25 Epoch	12/200	27s 1s/step - accuracy: 0.3581 - loss: 1.7205 - val_accuracy: 0.3650 - val_loss: 1.7061
25/25 Epoch	13/200	26s 1s/step - accuracy: 0.3839 - loss: 1.7241 - val_accuracy: 0.4250 - val_loss: 1.6044
25/25		25s 1s/step - accuracy: 0.4243 - loss: 1.6182 - val_accuracy: 0.3600 - val_loss: 1.7587
25/25	14/200	26s 1s/step - accuracy: 0.4233 - loss: 1.6799 - val_accuracy: 0.4700 - val_loss: 1.5579
Epoch 25/25	15/200	25s 1s/step - accuracy: 0.4338 - loss: 1.5815 - val_accuracy: 0.4800 - val_loss: 1.5056
Epoch	16/200	
25/25 Epoch	17/200	28s 1s/step - accuracy: 0.4718 - loss: 1.4877 - val_accuracy: 0.3650 - val_loss: 1.6719
25/25 Enoch	18/200	47s 1s/step - accuracy: 0.4637 - loss: 1.5132 - val_accuracy: 0.4750 - val_loss: 1.4645
25/25		28s 1s/step - accuracy: 0.4672 - loss: 1.4364 - val_accuracy: 0.5000 - val_loss: 1.4407
Epoch 25/25	19/200	27s 1s/step - accuracy: 0.4664 - loss: 1.4338 - val_accuracy: 0.4600 - val_loss: 1.4912
Epoch 25/25	20/200	26s 1s/step - accuracy: 0.4846 - loss: 1.3805 - val_accuracy: 0.4950 - val_loss: 1.4114
Epoch	21/200	
25/25 Epoch	22/200	28s 1s/step - accuracy: 0.5301 - loss: 1.3195 - val_accuracy: 0.5650 - val_loss: 1.3299
25/25 Enoch	23/200	39s 1s/step - accuracy: 0.5377 - loss: 1.3187 - val_accuracy: 0.5200 - val_loss: 1.4057
25/25		26s 1s/step - accuracy: 0.5209 - loss: 1.3272 - val_accuracy: 0.5250 - val_loss: 1.3737
Epoch 25/25	24/200	45s 1s/step - accuracy: 0.5138 - loss: 1.3195 - val_accuracy: 0.4800 - val_loss: 1.3970
Epoch 25/25	25/200	27s 1s/step - accuracy: 0.5613 - loss: 1.2026 - val accuracy: 0.4850 - val loss: 1.2982
Epoch	26/200	
25/25 Epoch	27/200	27s 1s/step - accuracy: 0.5462 - loss: 1.2349 - val_accuracy: 0.5150 - val_loss: 1.4277
25/25 Enoch	28/200	28s 1s/step - accuracy: 0.4924 - loss: 1.3835 - val_accuracy: 0.5600 - val_loss: 1.2649
25/25		27s 1s/step - accuracy: 0.5781 - loss: 1.2120 - val_accuracy: 0.5400 - val_loss: 1.3177
25/25	29/200	28s 1s/step - accuracy: 0.5310 - loss: 1.2935 - val_accuracy: 0.5350 - val_loss: 1.2755
Epoch 25/25	30/200	27s 1s/step - accuracy: 0.5842 - loss: 1.1832 - val_accuracy: 0.5650 - val_loss: 1.2301
Epoch	31/200	
25/25 Epoch	32/200	27s 1s/step - accuracy: 0.6170 - loss: 1.1483 - val_accuracy: 0.5300 - val_loss: 1.2360
25/25 Epoch	33/200	27s 1s/step - accuracy: 0.5694 - loss: 1.1414 - val_accuracy: 0.5300 - val_loss: 1.1925
25/25		28s 1s/step - accuracy: 0.5970 - loss: 1.1031 - val_accuracy: 0.5650 - val_loss: 1.1584
25/25	34/200	28s 1s/step - accuracy: 0.6325 - loss: 1.0118 - val_accuracy: 0.6000 - val_loss: 1.1658
Epoch 25/25	35/200	28s 1s/step - accuracy: 0.6461 - loss: 1.0089 - val_accuracy: 0.5200 - val_loss: 1.3458
Epoch	36/200	
25/25 Epoch	37/200	27s 1s/step - accuracy: 0.6280 - loss: 1.0219 - val_accuracy: 0.5500 - val_loss: 1.2555
25/25 Epoch	38/200	29s 1s/step - accuracy: 0.6563 - loss: 1.0223 - val_accuracy: 0.5600 - val_loss: 1.2584
25/25		29s 1s/step - accuracy: 0.6219 - loss: 1.0022 - val_accuracy: 0.5750 - val_loss: 1.1565
Epoch 25/25	39/200	28s 1s/step - accuracy: 0.6673 - loss: 0.9150 - val_accuracy: 0.6000 - val_loss: 1.1377
Epoch 25/25	40/200	27s 1s/step - accuracy: 0.6545 - loss: 0.9413 - val_accuracy: 0.5750 - val_loss: 1.1518
Epoch	41/200	
25/25		28s 1s/step - accuracy: 0.6994 - loss: 0.8533 - val_accuracy: 0.5050 - val_loss: 1.4275

	42/200	42- 1-/step	750
25/25 Epoch	43/200	• 42s 1s/step - accuracy: 0.6991 - loss: 0.8246 - val_accuracy: 0.5700 - val_loss: 1.1	/56
25/25 Epoch	44/200	27s 1s/step - accuracy: 0.7177 - loss: 0.7888 - val_accuracy: 0.6200 - val_loss: 1.20	∂86
25/25		29s 1s/step - accuracy: 0.7449 - loss: 0.7547 - val_accuracy: 0.6150 - val_loss: 1.1	535
25/25	45/200 ———————	42s 1s/step - accuracy: 0.7052 - loss: 0.7935 - val_accuracy: 0.5850 - val_loss: 1.30	ð92
Epoch 25/25	46/200	39s 1s/step - accuracy: 0.7307 - loss: 0.7758 - val_accuracy: 0.5550 - val_loss: 1.1	893
Epoch	47/200		
25/25 Epoch	48/200	• 41s 1s/step - accuracy: 0.7160 - loss: 0.8295 - val_accuracy: 0.5800 - val_loss: 1.2	159
25/25 Enoch	49/200	29s 1s/step - accuracy: 0.7674 - loss: 0.6261 - val_accuracy: 0.6150 - val_loss: 1.23	276
25/25		27s 1s/step - accuracy: 0.7672 - loss: 0.6559 - val_accuracy: 0.6100 - val_loss: 1.2	374
25/25	50/200	29s 1s/step - accuracy: 0.7832 - loss: 0.6168 - val_accuracy: 0.5850 - val_loss: 1.3	392
Epoch 25/25	51/200	30s 1s/step - accuracy: 0.7379 - loss: 0.7095 - val_accuracy: 0.5750 - val_loss: 1.4	176
Epoch	52/200		
25/25 Epoch	53/200	30s 1s/step - accuracy: 0.7792 - loss: 0.5799 - val_accuracy: 0.5300 - val_loss: 1.40	130
25/25 Epoch	54/200	30s 1s/step - accuracy: 0.7901 - loss: 0.5600 - val_accuracy: 0.6350 - val_loss: 1.29	566
25/25		28s 1s/step - accuracy: 0.7973 - loss: 0.5726 - val_accuracy: 0.5500 - val_loss: 1.3	714
25/25	55/200	27s 1s/step - accuracy: 0.7668 - loss: 0.7013 - val_accuracy: 0.6250 - val_loss: 1.34	406
Epoch 25/25	56/200	• 27s 1s/step - accuracy: 0.7882 - loss: 0.5294 - val accuracy: 0.6150 - val loss: 1.34	486
Epoch	57/200	- 30s 1s/step - accuracy: 0.8414 - loss: 0.4948 - val_accuracy: 0.5900 - val_loss: 1.40	
-	58/200		
25/25 Epoch	59/200	· 27s 1s/step - accuracy: 0.8212 - loss: 0.4944 - val_accuracy: 0.6150 - val_loss: 1.47	117
25/25 Enach	60/200	43s 1s/step - accuracy: 0.8479 - loss: 0.4028 - val_accuracy: 0.6050 - val_loss: 1.44	449
25/25		29s 1s/step - accuracy: 0.8089 - loss: 0.5510 - val_accuracy: 0.5700 - val_loss: 1.5	338
Epoch 25/25	61/200	• 28s 1s/step - accuracy: 0.8237 - loss: 0.4647 - val_accuracy: 0.5950 - val_loss: 1.4	833
Epoch 25/25	62/200	41s 1s/step - accuracy: 0.8668 - loss: 0.3522 - val accuracy: 0.5650 - val loss: 1.4	801
Epoch	63/200		
25/25 Epoch	64/200	28s 1s/step - accuracy: 0.8646 - loss: 0.4137 - val_accuracy: 0.6200 - val_loss: 1.59	929
25/25 Enoch	65/200	29s 1s/step - accuracy: 0.8611 - loss: 0.4100 - val_accuracy: 0.6150 - val_loss: 1.5	140
25/25		30s 1s/step - accuracy: 0.8943 - loss: 0.2995 - val_accuracy: 0.5300 - val_loss: 2.02	294
25/25	66/200	28s 1s/step - accuracy: 0.8659 - loss: 0.3771 - val_accuracy: 0.6100 - val_loss: 1.34	474
Epoch 25/25	67/200	28s 1s/step - accuracy: 0.8714 - loss: 0.3636 - val_accuracy: 0.6450 - val_loss: 1.4	715
Epoch 25/25	68/200	- 26s 1s/step - accuracy: 0.8731 - loss: 0.3432 - val_accuracy: 0.6350 - val_loss: 1.54	
Epoch	69/200		
25/25 Epoch	70/200	28s 1s/step - accuracy: 0.9113 - loss: 0.2766 - val_accuracy: 0.6000 - val_loss: 1.63	362
25/25 Enoch	71/200	30s 1s/step - accuracy: 0.9115 - loss: 0.2632 - val_accuracy: 0.5900 - val_loss: 1.79	935
25/25		30s 1s/step - accuracy: 0.8804 - loss: 0.3234 - val_accuracy: 0.5850 - val_loss: 1.75	517
25/25	72/200	29s 1s/step - accuracy: 0.9097 - loss: 0.2619 - val_accuracy: 0.6050 - val_loss: 1.76	557
Epoch 25/25	73/200	31s 1s/step - accuracy: 0.8978 - loss: 0.2883 - val_accuracy: 0.5900 - val_loss: 1.98	864
	74/200		
	75/200		
25/25 Epoch	76/200	• 28s 1s/step - accuracy: 0.9287 - loss: 0.2225 - val_accuracy: 0.6150 - val_loss: 1.83	275
25/25 Enoch	77/200	28s 1s/step - accuracy: 0.9313 - loss: 0.2132 - val_accuracy: 0.5600 - val_loss: 1.96	565
25/25		30s 1s/step - accuracy: 0.9132 - loss: 0.2315 - val_accuracy: 0.6250 - val_loss: 1.79	940
Epoch 25/25	78/200	30s 1s/step - accuracy: 0.9298 - loss: 0.2025 - val_accuracy: 0.6100 - val_loss: 1.7	548
Epoch 25/25	79/200	28s 1s/step - accuracy: 0.8969 - loss: 0.2864 - val_accuracy: 0.5700 - val_loss: 1.93	362
Epoch	80/200		
-	81/200	28s 1s/step - accuracy: 0.9037 - loss: 0.2708 - val_accuracy: 0.6050 - val_loss: 1.78	
25/25 Epoch	82/200	30s 1s/step - accuracy: 0.9479 - loss: 0.1578 - val_accuracy: 0.6050 - val_loss: 1.93	154
25/25		30s 1s/step - accuracy: 0.9278 - loss: 0.2297 - val_accuracy: 0.6100 - val_loss: 2.00	ð85

Epoch 25/25	83/200	• 28s 1s/step - accuracy: 0.9603 - loss: 0.1253 - val_accuracy: 0.6050 - val_loss: 1.99	044
	84/200	205 15/Step - accuracy. 0.9003 - 1055. 0.1233 - Val_accuracy. 0.0030 - Val_1055. 1.9	744
25/25 Epoch	85/200	28s 1s/step - accuracy: 0.9349 - loss: 0.2179 - val_accuracy: 0.5950 - val_loss: 1.76	598
25/25		29s 1s/step - accuracy: 0.9260 - loss: 0.2480 - val_accuracy: 0.6000 - val_loss: 1.90	ð84
25/25	86/200	29s 1s/step - accuracy: 0.9385 - loss: 0.1810 - val_accuracy: 0.6050 - val_loss: 1.98	800
Epoch 25/25	87/200	4 0s 1s/step - accuracy: 0.9315 - loss: 0.2115 - val_accuracy: 0.6000 - val_loss: 1.99	982
	88/200		
25/25 Epoch	89/200	· 29s 1s/step - accuracy: 0.9286 - loss: 0.1841 - val_accuracy: 0.6050 - val_loss: 1.82	210
25/25 Epoch	90/200	• 42s 1s/step - accuracy: 0.9449 - loss: 0.1455 - val_accuracy: 0.6050 - val_loss: 2.14	485
25/25		30s 1s/step - accuracy: 0.9580 - loss: 0.1347 - val_accuracy: 0.6050 - val_loss: 2.00	524
25/25	91/200	29s 1s/step - accuracy: 0.9658 - loss: 0.1287 - val_accuracy: 0.6050 - val_loss: 2.00	523
Epoch 25/25	92/200	41s 1s/step - accuracy: 0.9296 - loss: 0.2185 - val_accuracy: 0.5750 - val_loss: 2.18	836
Epoch	93/200		
25/25 Epoch	94/200	• 31s 1s/step - accuracy: 0.9534 - loss: 0.1375 - val_accuracy: 0.5700 - val_loss: 2.30	
25/25 Epoch	95/200	28s 1s/step - accuracy: 0.9321 - loss: 0.1736 - val_accuracy: 0.5900 - val_loss: 2.09	9 75
25/25		28s 1s/step - accuracy: 0.9544 - loss: 0.1090 - val_accuracy: 0.6050 - val_loss: 2.18	837
25/25	96/200	29s 1s/step - accuracy: 0.9666 - loss: 0.1233 - val_accuracy: 0.5250 - val_loss: 2.79	924
Epoch 25/25	97/200	29s 1s/step - accuracy: 0.9305 - loss: 0.2006 - val_accuracy: 0.5950 - val_loss: 2.16	621
Epoch 25/25	98/200	29s 1s/step - accuracy: 0.9653 - loss: 0.1326 - val_accuracy: 0.5600 - val_loss: 2.36	662
Epoch	99/200		
25/25 Epoch	100/200	4 0s 1s/step - accuracy: 0.9604 - loss: 0.1206 - val_accuracy: 0.6150 - val_loss: 2.28	331
25/25 Enoch	101/200	29s 1s/step - accuracy: 0.9483 - loss: 0.1839 - val_accuracy: 0.6150 - val_loss: 1.8	584
25/25		30s 1s/step - accuracy: 0.9725 - loss: 0.0949 - val_accuracy: 0.5950 - val_loss: 2.1	361
Epoch 25/25	102/200	28s 1s/step - accuracy: 0.9591 - loss: 0.1132 - val_accuracy: 0.6500 - val_loss: 2.1	777
Epoch 25/25	103/200	28s 1s/step - accuracy: 0.9600 - loss: 0.1332 - val_accuracy: 0.6000 - val_loss: 3.0	784
Epoch	104/200		
25/25 Epoch	105/200	• 29s 1s/step - accuracy: 0.9355 - loss: 0.1872 - val_accuracy: 0.5750 - val_loss: 2.49	3 /8
25/25 Enoch	106/200	30s 1s/step - accuracy: 0.9633 - loss: 0.1094 - val_accuracy: 0.6200 - val_loss: 2.23	200
25/25		28s 1s/step - accuracy: 0.9577 - loss: 0.1378 - val_accuracy: 0.5850 - val_loss: 2.54	441
25/25	107/200	41s 1s/step - accuracy: 0.9277 - loss: 0.2046 - val_accuracy: 0.6250 - val_loss: 2.16	515
Epoch 25/25	108/200	28s 1s/step - accuracy: 0.9599 - loss: 0.0948 - val_accuracy: 0.5750 - val_loss: 2.40	605
Epoch 25/25	109/200	- 43s 1s/step - accuracy: 0.9670 - loss: 0.0811 - val_accuracy: 0.5950 - val_loss: 2.54	
Epoch	110/200		
25/25 Epoch	111/200	28s 1s/step - accuracy: 0.9646 - loss: 0.1045 - val_accuracy: 0.5600 - val_loss: 3.16	549
25/25 Enoch	112/200	29s 1s/step - accuracy: 0.9542 - loss: 0.1843 - val_accuracy: 0.5850 - val_loss: 2.36	599
25/25		31s 1s/step - accuracy: 0.9595 - loss: 0.0938 - val_accuracy: 0.6200 - val_loss: 2.54	488
25/25	113/200	29s 1s/step - accuracy: 0.9756 - loss: 0.0709 - val_accuracy: 0.6450 - val_loss: 2.73	340
Epoch 25/25	114/200	28s 1s/step - accuracy: 0.9678 - loss: 0.0969 - val_accuracy: 0.5950 - val_loss: 2.56	608
	115/200	- 28s 1s/step - accuracy: 0.9566 - loss: 0.1611 - val_accuracy: 0.6350 - val_loss: 2.2	
	116/200		
25/25 Epoch	117/200	30s 1s/step - accuracy: 0.9844 - loss: 0.0740 - val_accuracy: 0.5950 - val_loss: 2.5	361
25/25 Enoch	118/200	30s 1s/step - accuracy: 0.9755 - loss: 0.0631 - val_accuracy: 0.6150 - val_loss: 2.50	∂37
25/25		28s 1s/step - accuracy: 0.9773 - loss: 0.0840 - val_accuracy: 0.5650 - val_loss: 2.72	249
Epoch 25/25	119/200	28s 1s/step - accuracy: 0.9873 - loss: 0.0544 - val_accuracy: 0.6150 - val_loss: 2.78	348
Epoch 25/25	120/200	42s 1s/step - accuracy: 0.9768 - loss: 0.0695 - val_accuracy: 0.6100 - val_loss: 2.99	545
Epoch	121/200		
	122/200	29s 1s/step - accuracy: 0.9731 - loss: 0.0767 - val_accuracy: 0.6150 - val_loss: 2.5	
25/25 Epoch	123/200	· 28s 1s/step - accuracy: 0.9797 - loss: 0.0810 - val_accuracy: 0.6100 - val_loss: 2.52	275
25/25		28s 1s/step - accuracy: 0.9807 - loss: 0.0617 - val_accuracy: 0.5100 - val_loss: 3.4	139

```
Epoch 124/200
                           30s 1s/step - accuracy: 0.9686 - loss: 0.1716 - val_accuracy: 0.5950 - val_loss: 2.5259
25/25
Epoch 125/200
25/25
                          29s 1s/step - accuracy: 0.9579 - loss: 0.1311 - val_accuracy: 0.6000 - val_loss: 2.3598
Epoch 126/200
25/25
                          28s 1s/step - accuracy: 0.9614 - loss: 0.1006 - val_accuracy: 0.6250 - val_loss: 2.3710
Epoch 127/200
                          42s 1s/step - accuracy: 0.9831 - loss: 0.0494 - val_accuracy: 0.5900 - val_loss: 2.4309
25/25
Epoch 128/200
25/25
                          30s 1s/step - accuracy: 0.9865 - loss: 0.0413 - val_accuracy: 0.6150 - val_loss: 2.5843
Epoch 129/200
25/25
                          29s 1s/step - accuracy: 0.9723 - loss: 0.0938 - val_accuracy: 0.6200 - val_loss: 2.5107
Epoch 130/200
25/25
                           40s 1s/step - accuracy: 0.9890 - loss: 0.0352 - val_accuracy: 0.6250 - val_loss: 2.6177
Epoch 131/200
25/25
                          42s 1s/step - accuracy: 0.9732 - loss: 0.1001 - val_accuracy: 0.6350 - val_loss: 2.6362
Epoch 132/200
                          28s 1s/step - accuracy: 0.9685 - loss: 0.0753 - val_accuracy: 0.6450 - val_loss: 2.7245
25/25
Epoch 133/200
25/25
                          30s 1s/step - accuracy: 0.9689 - loss: 0.1208 - val_accuracy: 0.6350 - val_loss: 2.4855
Epoch 134/200
25/25
                          29s 1s/step - accuracy: 0.9775 - loss: 0.0830 - val_accuracy: 0.5800 - val_loss: 2.6929
Epoch 135/200
25/25
                          27s 1s/step - accuracy: 0.9776 - loss: 0.0684 - val_accuracy: 0.6250 - val_loss: 2.4147
Epoch 136/200
25/25
                          29s 1s/step - accuracy: 0.9900 - loss: 0.0449 - val_accuracy: 0.6200 - val_loss: 2.4069
Epoch 137/200
25/25
                          28s 1s/step - accuracy: 0.9814 - loss: 0.0603 - val_accuracy: 0.6400 - val_loss: 2.3076
Epoch 138/200
                          29s 1s/step - accuracy: 0.9831 - loss: 0.0480 - val_accuracy: 0.6300 - val_loss: 2.4884
25/25
Epoch 139/200
25/25
                          - 29s 1s/step - accuracy: 0.9798 - loss: 0.0811 - val_accuracy: 0.6100 - val_loss: 2.8976
Epoch 140/200
25/25
                          30s 1s/step - accuracy: 0.9540 - loss: 0.1233 - val_accuracy: 0.6000 - val_loss: 2.5558
Epoch 141/200
25/25
                          28s 1s/step - accuracy: 0.9726 - loss: 0.0607 - val_accuracy: 0.6150 - val_loss: 2.5155
Epoch 142/200
25/25
                          26s 1s/step - accuracy: 0.9861 - loss: 0.0612 - val_accuracy: 0.6150 - val_loss: 2.5438
Epoch 143/200
25/25
                          27s 1s/step - accuracy: 0.9839 - loss: 0.0469 - val_accuracy: 0.6150 - val_loss: 2.7714
Epoch 144/200
25/25
                          28s 1s/step - accuracy: 0.9804 - loss: 0.0644 - val_accuracy: 0.6050 - val_loss: 2.6876
Epoch 145/200
25/25
                          28s 1s/step - accuracy: 0.9745 - loss: 0.0825 - val_accuracy: 0.6150 - val_loss: 2.5074
Epoch 146/200
25/25
                           28s 1s/step - accuracy: 0.9850 - loss: 0.0500 - val_accuracy: 0.5700 - val_loss: 2.7695
Epoch 147/200
25/25
                          13s 534ms/step - accuracy: 0.9874 - loss: 0.0435 - val_accuracy: 0.6100 - val_loss: 2.9878
Epoch 148/200
                          13s 516ms/step - accuracy: 0.9890 - loss: 0.0413 - val_accuracy: 0.5550 - val_loss: 2.9865
25/25
Epoch 149/200
25/25
                          - 13s 506ms/step - accuracy: 0.9630 - loss: 0.1362 - val_accuracy: 0.5950 - val_loss: 2.6534
Epoch 150/200
25/25
                          13s 536ms/step - accuracy: 0.9646 - loss: 0.1046 - val_accuracy: 0.6100 - val_loss: 2.6097
Epoch 151/200
25/25
                          13s 516ms/step - accuracy: 0.9851 - loss: 0.0577 - val_accuracy: 0.5700 - val_loss: 3.1224
Epoch 152/200
25/25
                          12s 472ms/step - accuracy: 0.9864 - loss: 0.0592 - val_accuracy: 0.6250 - val_loss: 3.0793
Epoch 153/200
25/25
                          12s 493ms/step - accuracy: 0.9766 - loss: 0.1038 - val_accuracy: 0.5900 - val_loss: 3.0936
Epoch 154/200
25/25
                          14s 545ms/step - accuracy: 0.9854 - loss: 0.0362 - val_accuracy: 0.6250 - val_loss: 2.8501
Epoch 155/200
25/25
                          13s 512ms/step - accuracy: 0.9807 - loss: 0.0572 - val accuracy: 0.6350 - val loss: 2.8597
Epoch 156/200
25/25
                          14s 553ms/step - accuracy: 0.9818 - loss: 0.0439 - val_accuracy: 0.6250 - val_loss: 2.8805
Epoch 157/200
25/25
                          13s 523ms/step - accuracy: 0.9693 - loss: 0.1114 - val_accuracy: 0.6150 - val_loss: 2.8996
Epoch 158/200
25/25
                          35s 1s/step - accuracy: 0.9876 - loss: 0.0445 - val_accuracy: 0.6100 - val_loss: 2.5862
Epoch 159/200
25/25
                          49s 2s/step - accuracy: 0.9832 - loss: 0.0491 - val accuracy: 0.6150 - val loss: 2.7456
Epoch 160/200
25/25
                          - 45s 2s/step - accuracy: 0.9890 - loss: 0.0416 - val_accuracy: 0.6350 - val_loss: 2.6034
Epoch 161/200
25/25
                          45s 2s/step - accuracy: 0.9840 - loss: 0.0574 - val_accuracy: 0.6200 - val_loss: 2.5495
Epoch 162/200
25/25
                           36s 1s/step - accuracy: 0.9752 - loss: 0.0676 - val_accuracy: 0.6400 - val_loss: 2.4306
Epoch 163/200
25/25
                          31s 1s/step - accuracy: 0.9842 - loss: 0.0456 - val_accuracy: 0.5650 - val_loss: 3.2392
Epoch 164/200
25/25
                           44s 2s/step - accuracy: 0.9629 - loss: 0.1236 - val_accuracy: 0.6050 - val_loss: 2.5910
```

```
Epoch 165/200
                                  39s 2s/step - accuracy: 0.9845 - loss: 0.0498 - val_accuracy: 0.5600 - val_loss: 3.4614
       25/25
       Epoch 166/200
       25/25
                                 • 41s 2s/step - accuracy: 0.9775 - loss: 0.0778 - val_accuracy: 0.5850 - val_loss: 3.4735
       Epoch 167/200
      25/25
                                 39s 2s/step - accuracy: 0.9761 - loss: 0.1006 - val_accuracy: 0.6250 - val_loss: 2.8417
       Epoch 168/200
                                 - 36s 1s/step - accuracy: 0.9890 - loss: 0.0297 - val_accuracy: 0.6100 - val_loss: 3.5556
      25/25
       Epoch 169/200
      25/25
                                 - 42s 2s/step - accuracy: 0.9769 - loss: 0.0796 - val_accuracy: 0.6250 - val_loss: 2.8833
      Epoch 170/200
      25/25
                                 - 76s 2s/step - accuracy: 0.9927 - loss: 0.0417 - val_accuracy: 0.5800 - val_loss: 3.4683
       Epoch 171/200
      25/25
                                 38s 2s/step - accuracy: 0.9753 - loss: 0.1325 - val_accuracy: 0.5600 - val_loss: 3.0028
       Epoch 172/200
       25/25
                                 41s 2s/step - accuracy: 0.9977 - loss: 0.0172 - val_accuracy: 0.5900 - val_loss: 3.1076
       Epoch 173/200
                                 36s 1s/step - accuracy: 0.9821 - loss: 0.0533 - val_accuracy: 0.6150 - val_loss: 2.7431
      25/25
      Epoch 174/200
       25/25
                                 - 32s 1s/step - accuracy: 0.9824 - loss: 0.0689 - val_accuracy: 0.6100 - val_loss: 2.9436
      Epoch 175/200
       25/25
                                 45s 1s/step - accuracy: 0.9848 - loss: 0.0335 - val_accuracy: 0.6350 - val_loss: 3.0803
       Epoch 176/200
      25/25
                                 42s 2s/step - accuracy: 0.9724 - loss: 0.0594 - val_accuracy: 0.6200 - val_loss: 2.8165
       Epoch 177/200
                                 • 38s 2s/step - accuracy: 0.9929 - loss: 0.0385 - val_accuracy: 0.5950 - val_loss: 2.7957
       25/25
       Epoch 178/200
      25/25
                                 • 39s 1s/step - accuracy: 0.9721 - loss: 0.0697 - val_accuracy: 0.5900 - val_loss: 3.0092
      Epoch 179/200
                                 37s 1s/step - accuracy: 0.9793 - loss: 0.0498 - val_accuracy: 0.6150 - val_loss: 3.1001
       25/25
       Epoch 180/200
      25/25
                                 - 40s 1s/step - accuracy: 0.9918 - loss: 0.0250 - val_accuracy: 0.6150 - val_loss: 3.1547
       Epoch 181/200
      25/25
                                 • 46s 2s/step - accuracy: 0.9845 - loss: 0.0484 - val_accuracy: 0.6300 - val_loss: 3.0083
       Epoch 182/200
      25/25
                                 • 41s 2s/step - accuracy: 0.9851 - loss: 0.0400 - val_accuracy: 0.6050 - val_loss: 3.2633
       Epoch 183/200
       25/25
                                 41s 2s/step - accuracy: 0.9809 - loss: 0.0622 - val_accuracy: 0.5850 - val_loss: 2.9806
       Epoch 184/200
      25/25
                                 - 78s 2s/step - accuracy: 0.9896 - loss: 0.0433 - val_accuracy: 0.5200 - val_loss: 4.4966
       Epoch 185/200
                                 - 36s 1s/step - accuracy: 0.9712 - loss: 0.0948 - val_accuracy: 0.6250 - val_loss: 3.3453
       25/25
       Epoch 186/200
      25/25
                                 • 48s 2s/step - accuracy: 0.9821 - loss: 0.0439 - val_accuracy: 0.6100 - val_loss: 3.2421
       Epoch 187/200
       25/25 -
                                 43s 2s/step - accuracy: 0.9888 - loss: 0.0370 - val_accuracy: 0.6150 - val_loss: 3.3000
       Epoch 188/200
      25/25
                                 - 52s 534ms/step - accuracy: 0.9968 - loss: 0.0204 - val_accuracy: 0.6100 - val_loss: 3.2120
       Epoch 189/200
      25/25
                                 - 14s 568ms/step - accuracy: 0.9670 - loss: 0.1086 - val_accuracy: 0.6300 - val_loss: 2.9584
      Epoch 190/200
      25/25
                                - 14s 542ms/step - accuracy: 0.9895 - loss: 0.0179 - val_accuracy: 0.6400 - val_loss: 3.2522
      Epoch 191/200
       25/25
                                 14s 539ms/step - accuracy: 0.9852 - loss: 0.0418 - val_accuracy: 0.6300 - val_loss: 2.9038
       Epoch 192/200
      25/25
                                 - 13s 522ms/step - accuracy: 0.9963 - loss: 0.0141 - val_accuracy: 0.6100 - val_loss: 3.7893
       Epoch 193/200
       25/25 -
                                 - 14s 549ms/step - accuracy: 0.9896 - loss: 0.0594 - val_accuracy: 0.6050 - val_loss: 3.6565
       Epoch 194/200
      25/25
                                 · 13s 531ms/step - accuracy: 0.9762 - loss: 0.0803 - val_accuracy: 0.6000 - val_loss: 3.0845
      Epoch 195/200
       25/25
                                 14s 553ms/step - accuracy: 0.9910 - loss: 0.0324 - val_accuracy: 0.5900 - val_loss: 3.1809
      Epoch 196/200
      25/25
                                 - 13s 531ms/step - accuracy: 0.9763 - loss: 0.0641 - val_accuracy: 0.6000 - val_loss: 3.3820
       Epoch 197/200
      25/25
                                 - 12s 490ms/step - accuracy: 0.9811 - loss: 0.0780 - val_accuracy: 0.6200 - val_loss: 3.3695
       Epoch 198/200
       25/25
                                 · 39s 2s/step - accuracy: 0.9879 - loss: 0.0313 - val_accuracy: 0.6050 - val_loss: 3.4536
       Epoch 199/200
       25/25
                                 37s 1s/step - accuracy: 0.9894 - loss: 0.0385 - val_accuracy: 0.6350 - val_loss: 3.3571
       Epoch 200/200
       25/25
                                 42s 2s/step - accuracy: 0.9831 - loss: 0.0570 - val_accuracy: 0.6000 - val_loss: 3.3900
In [8]: print(f"Execution time: {elapsed_time:.2f} seconds")
       Execution time: 6102.52 seconds
```

```
In [9]: def append_core_data(score_path, num_cores, elapsed_time):
    # Check if the file already exists
    file_exists = os.path.exists(score_path)

# Open the file in append mode
```

with open(score_path, mode='a', newline='') as file:

```
writer = csv.writer(file)

# If the file is new, write the header
if not file_exists:
    writer.writerow(["Number of Cores", "Elapsed Time"])

# Write the new data
writer.writerow([num_cores, elapsed_time])
```

In [10]: score_path = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DEEP LEARNING WITH HPSC\core_data.txt"
 append_core_data(score_path, number_of_worker, elapsed_time)

Import Req Lib

```
In [1]: %matplotlib inline
        import shutil
        import random
        import numpy as np
        from warnings import filterwarnings
        filterwarnings('ignore')
        from tensorflow.keras import layers, regularizers, optimizers
        from tensorflow.keras import models
        from tensorflow.keras.models import Sequential, Model
        from tensorflow.keras.layers import LeakyReLU,Dense, Activation, Flatten, Dropout, BatchNormalization,Conv2D, MaxPooling2D
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import tensorflow as tf
        import os
        import time
        import csv
        from matplotlib import figure
```

Define 2 worker

```
In [2]: # Set the number of threads
number_of_worker = 2
os.environ['OMP_NUM_THREADS'] = '2' # OpenMP threads for parallelism
os.environ['TF_NUM_INTEROP_THREADS'] = '2' # Threads for inter-operation parallelism
os.environ['TF_NUM_INTRAOP_THREADS'] = '2' # Threads for intra-operation parallelism

# Confirm TensorFlow is using the specified number of threads
tf.config.threading.set_inter_op_parallelism_threads(number_of_worker)
tf.config.threading.set_intra_op_parallelism_threads(number_of_worker)
```

Train Val data Split

```
In [3]: source_dir = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DATA\Convert_Audio_File_to_jpg_file"
        target_dir = r'genres_train_val_split_data'
        split_ratio = 0.8
        def Train_Test_Split(source_dir,target_dir,split_ratio):
            # Define source and target directories
            train_dir = os.path.join(target_dir, 'train')
            val_dir = os.path.join(target_dir, 'val')
            # Create target directories if they don't exist
            os.makedirs(train_dir, exist_ok=True)
            os.makedirs(val_dir, exist_ok=True)
            # Get the list of class directories
            classes = [d for d in os.listdir(source_dir) if os.path.isdir(os.path.join(source_dir, d))]
            for class name in classes:
                # Create class directories in train and val folders
                os.makedirs(os.path.join(train_dir, class_name), exist_ok=True)
                os.makedirs(os.path.join(val_dir, class_name), exist_ok=True)
                # Get list of images in the class directory
                class_dir = os.path.join(source_dir, class_name)
                images = [f for f in os.listdir(class_dir) if os.path.isfile(os.path.join(class_dir, f))]
                # Shuffle the images
                random.shuffle(images)
                # Compute the split point
                split_point = int(len(images) * split_ratio)
                # Split the images into training and validation sets
                train images = images[:split point]
                val_images = images[split_point:]
```

In [4]: Train_Test_Split(source_dir,target_dir,split_ratio)

Data split completed successfully!

Load the Data

```
In [5]: WIDTH = 64
        HEIGHT = 64
        BATCH_SIZE = 32
        TRAIN_DIR=r'genres_train_val_split_data/train'
        val_dir = r'genres_train_val_split_data/val'
        # data prep
        train_datagen = ImageDataGenerator(
            rescale=1./255.,validation_split=0.25)
        train_generator = train_datagen.flow_from_directory(
            TRAIN_DIR,
            target size=(HEIGHT, WIDTH),
                batch_size=BATCH_SIZE,
                class_mode='categorical')
        validation_gen = train_datagen.flow_from_directory(
            val_dir,target_size = (HEIGHT,WIDTH),
            batch_size = BATCH_SIZE,
            class_mode = 'categorical'
```

Found 800 images belonging to 10 classes. Found 200 images belonging to 10 classes.

Model Architecture

```
In [6]: model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                         input_shape=(64,64,3)))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(128, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(128, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(10, activation='softmax'))
        model.compile(optimizers.RMSprop(learning_rate=0.0005, decay=1e-6),loss="categorical_crossentropy",metrics=["accuracy"])
        model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 64, 64, 32)	896
activation (Activation)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 62, 62, 64)	18,496
activation_1 (Activation)	(None, 62, 62, 64)	0
max_pooling2d (MaxPooling2D)	(None, 31, 31, 64)	0
dropout (Dropout)	(None, 31, 31, 64)	0
conv2d_2 (Conv2D)	(None, 31, 31, 64)	36,928
activation_2 (Activation)	(None, 31, 31, 64)	0
conv2d_3 (Conv2D)	(None, 29, 29, 64)	36,928
activation_3 (Activation)	(None, 29, 29, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_1 (Dropout)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 14, 14, 128)	73,856
activation_4 (Activation)	(None, 14, 14, 128)	0
conv2d_5 (Conv2D)	(None, 12, 12, 128)	147,584
activation_5 (Activation)	(None, 12, 12, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_2 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2,359,808
activation_6 (Activation)	(None, 512)	0
dropout_3 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5,130

Total params: 2,679,626 (10.22 MB)

Trainable params: 2,679,626 (10.22 MB)

Non-trainable params: 0 (0.00 B)

```
In [7]: STEP_SIZE_TRAIN=train_generator.n//train_generator.batch_size
# Measure the execution time
start_time = time.time()

model.fit(train_generator,validation_data=validation_gen,epochs=200)

end_time = time.time()
elapsed_time = end_time - start_time
```

Epoch 25/25		205	675ms/sten -	accuracy:	0 0658 - 10	nee	- val_accuracy:	0 1000 -	val loss:	2 3021
Epoch	2/200			-					_	
25/25 Epoch	3/200			-			val_accuracy:		_	
25/25 Epoch		16s	658ms/step -	accuracy:	0.1090 - lo	oss: 2.2960	val_accuracy:	0.2100 -	val_loss:	2.2949
25/25 Epoch		16s	652ms/step -	accuracy:	0.1633 - 10	oss: 2.2699	<pre>- val_accuracy:</pre>	0.1900 -	val_loss:	2.1670
25/25		15s	607ms/step -	accuracy:	0.2279 - 10	oss: 2.1263	- val_accuracy:	0.2100 -	val_loss:	2.0336
Epoch 25/25		15s	598ms/step -	accuracy:	0.2639 - 10	oss: 2.0327	- val_accuracy:	0.2800 -	val_loss:	2.0367
Epoch 25/25		15s	595ms/step -	accuracy:	0.2955 - lo	oss: 1.9982	- val_accuracy:	0.3250 -	val loss:	1.9601
Epoch 25/25	8/200		-	-			- val_accuracy:		_	
Epoch	9/200			-					_	
25/25 Epoch	10/200	15s	589ms/step -	accuracy:	0.2926 - 10	oss: 1.92/6	- val_accuracy:	0.3500 -	val_loss:	1.9102
25/25 Epoch	11/200	15s	602ms/step -	accuracy:	0.2897 - lo	oss: 1.9524	<pre>- val_accuracy:</pre>	0.3400 -	val_loss:	1.8158
25/25 Enoch	12/200	16s	650ms/step -	accuracy:	0.3268 - 10	oss: 1.8249	<pre>- val_accuracy:</pre>	0.3350 -	val_loss:	1.7215
25/25		17s	687ms/step -	accuracy:	0.3834 - 10	oss: 1.7091	- val_accuracy:	0.3750 -	val_loss:	1.6773
25/25	13/200	18s	700ms/step -	accuracy:	0.3712 - lo	ss: 1.7670	- val_accuracy:	0.2500 -	val_loss:	2.0661
Epoch 25/25	14/200	18s	699ms/step -	accuracy:	0.3958 - lo	oss: 1.7096	- val_accuracy:	0.3350 -	val_loss:	1.6078
Epoch 25/25	15/200	· 18s	714ms/step -	accuracy:	0.4427 - lo	oss: 1.5865	- val_accuracy:	0.3950 -	val loss:	1.5535
	16/200			-			- val_accuracy:		_	
Epoch	17/200			-					_	
25/25 Epoch	18/200	· 19s	741ms/step -	accuracy:	0.4032 - lo	oss: 1.6410	- val_accuracy:	0.4050 -	val_loss:	1.5667
25/25 Epoch	19/200	18s	733ms/step -	accuracy:	0.4755 - lo	oss: 1.4892	val_accuracy:	0.4050 -	val_loss:	1.5534
25/25		18s	721ms/step -	accuracy:	0.4703 - lo	oss: 1.4501	<pre>- val_accuracy:</pre>	0.4250 -	val_loss:	1.5390
25/25		18s	715ms/step -	accuracy:	0.4298 - 10	oss: 1.5520	- val_accuracy:	0.3700 -	val_loss:	1.6879
25/25	21/200	18s	727ms/step -	accuracy:	0.4590 - lo	oss: 1.4803	- val_accuracy:	0.4050 -	val_loss:	1.5830
Epoch 25/25	22/200	- 18s	713ms/step -	accuracy:	0.4832 - lo	oss: 1.4678	<pre>- val_accuracy:</pre>	0.4850 -	val_loss:	1.3881
Epoch 25/25	23/200	185	721ms/sten -	accuracy:	0.5379 - lo	nss: 1.2982	- val_accuracy:	0.4600 -	val loss:	1.4611
	24/200						- val_accuracy:		_	
Epoch	25/200			-					_	
25/25 Epoch	26/200	185	724ms/step -	accuracy:	0.5432 - 10	oss: 1.2727	- val_accuracy:	0.5100 -	val_loss:	1.3474
25/25 Epoch	27/200	18s	719ms/step -	accuracy:	0.5513 - lo	oss: 1.2137	<pre>- val_accuracy:</pre>	0.5150 -	val_loss:	1.3525
25/25 Enoch	28/200	18s	722ms/step -	accuracy:	0.5891 - 10	oss: 1.1408	<pre>- val_accuracy:</pre>	0.5200 -	val_loss:	1.3607
25/25		18s	713ms/step -	accuracy:	0.5416 - lo	oss: 1.1971	<pre>- val_accuracy:</pre>	0.4800 -	val_loss:	1.4404
25/25	29/200	18s	738ms/step -	accuracy:	0.5836 - 10	oss: 1.1399	- val_accuracy:	0.5000 -	val_loss:	1.3671
Epoch 25/25	30/200	18s	735ms/step -	accuracy:	0.5671 - lo	ss: 1.1366	- val_accuracy:	0.5200 -	val_loss:	1.3093
Epoch 25/25	31/200	19s	751ms/step -	accuracy:	0.6065 - lo	oss: 1.0926	- val_accuracy:	0.5100 -	val loss:	1.2397
	32/200			-			- val_accuracy:		_	
Epoch	33/200			-					_	
25/25 Epoch	34/200			-			val_accuracy:		_	
25/25 Epoch	35/200	17s	690ms/step -	accuracy:	0.5982 - lo	oss: 1.1189	- val_accuracy:	0.5650 -	val_loss:	1.2984
25/25 Epoch	36/200	18s	715ms/step -	accuracy:	0.6693 - 10	oss: 0.9085	val_accuracy:	0.5400 -	val_loss:	1.2657
25/25		18s	735ms/step -	accuracy:	0.6650 - lo	oss: 0.9373	val_accuracy:	0.5600 -	val_loss:	1.1906
25/25		18s	730ms/step -	accuracy:	0.6666 - 10	oss: 0.9179	- val_accuracy:	0.5850 -	val_loss:	1.1938
Epoch 25/25	38/200	18s	735ms/step -	accuracy:	0.6960 - lo	oss: 0.8488	- val_accuracy:	0.6050 -	val_loss:	1.1426
Epoch 25/25	39/200	18s	713ms/step -	accuracy:	0.6970 - lo	oss: 0.8214	- val_accuracy:	0.5650 -	val_loss:	1.2680
Epoch 25/25	40/200	18s	712ms/sten -	accuracy:	0.6785 - 10	oss: 0.8729	- val_accuracy:	0.5750 -	val loss:	1.1458
	41/200								_	
23/23		102	,1/m3/3ceh -	accuracy:	0.0055 - 10	,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	val_accuracy:	J. JUUU -	^aT_TO22;	1.1/03

Enoch	42/200										
25/25	42/200	- 18s	707ms/step -	accuracy:	0.6824 -	loss:	0.8254 -	val_accuracy:	0.5600 -	val loss:	1.1993
Epoch	43/200		•	,				_ ,		_	
25/25		18s	734ms/step -	accuracy:	0.7287 -	loss:	0.8137 -	val_accuracy:	0.6000 -	val_loss:	1.1503
25/25	44/200	- 18s	725ms/step -	accuracv:	0.7484 -	loss:	0.6776 -	val_accuracy:	0.5950 -	val loss:	1.1720
	45/200			,				_ ,		_	
25/25		18s	730ms/step -	accuracy:	0.7591 -	loss:	0.7487 -	val_accuracy:	0.5900 -	val_loss:	1.2234
25/25	46/200	- 18s	708ms/step -	accuracy:	0.7634 -	loss:	0.6756 -	val_accuracy:	0.5950 -	val loss:	1.1382
	47/200		•					_ ,		_	
25/25		18s	721ms/step -	accuracy:	0.7567 -	loss:	0.6095 -	val_accuracy:	0.6150 -	val_loss:	1.2420
25/25	48/200	18s	710ms/step -	accuracy:	0.7541 -	loss:	0.6675 -	val_accuracy:	0.6150 -	val_loss:	1.2987
-	49/200										
25/25 Enoch	50/200	- 18s	735ms/step -	accuracy:	0.7979 -	loss:	0.5888 -	val_accuracy:	0.6400 -	val_loss:	1.2228
25/25		18s	731ms/step -	accuracy:	0.7945 -	loss:	0.5587 -	val_accuracy:	0.5650 -	val_loss:	1.4477
-	51/200		/ /								
25/25 Enoch	52/200	- 18s	728ms/step -	accuracy:	0.8001 -	loss:	0.5551 -	val_accuracy:	0.6050 -	val_loss:	1.4301
25/25		18s	728ms/step -	accuracy:	0.7869 -	loss:	0.6716 -	val_accuracy:	0.5750 -	val_loss:	1.3105
	53/200	10-	710		0.7010	1	0 5700		0 6300		1 2621
25/25 Epoch	54/200	185	/18ms/step -	accuracy:	0.7919 -	1055:	0.5/82 -	val_accuracy:	0.6300 -	vai_ioss:	1.2621
25/25		18s	719ms/step -	accuracy:	0.8451 -	loss:	0.4805 -	val_accuracy:	0.6350 -	<pre>val_loss:</pre>	1.2906
Epoch 25/25	55/200	10c	720ms /ston	26611112641	0 9170	locci	0 1026	val_accuracy:	0 6200	val locc:	1 /216
	56/200	103	/20113/3CEP -	accuracy.	0.01/9 -	1055.	0.4920 -	vai_accuracy.	0.0200 -	va1_1055.	1.4210
25/25		18s	725ms/step -	accuracy:	0.8320 -	loss:	0.4691 -	val_accuracy:	0.6300 -	<pre>val_loss:</pre>	1.4858
Epoch 25/25	57/200	205	718ms/sten -	accuracy:	a 8a99 -	loss	0 4724 -	val_accuracy:	0 6100 -	val loss:	1 5430
	58/200	203	710m3/3ccp	accar acy.	0.0033	1033.	0.4724	var_acca, acy.	0.0100	vu1_1033.	1.5-50
25/25		18s	728ms/step -	accuracy:	0.8482 -	loss:	0.4111 -	val_accuracy:	0.5950 -	val_loss:	1.5319
25/25	59/200 	- 18s	708ms/step -	accuracv:	0.8596 -	loss:	0.3476 -	val_accuracy:	0.6100 -	val loss:	1.6917
	60/200		•	-				_ ,		_	
25/25		18s	709ms/step -	accuracy:	0.8134 -	loss:	0.4908 -	val_accuracy:	0.6200 -	val_loss:	1.5236
25/25	61/200	18s	715ms/step -	accuracy:	0.8909 -	loss:	0.3588 -	val_accuracy:	0.6200 -	val_loss:	1.4370
	62/200										
25/25 Enoch	63/200	- 18s	713ms/step -	accuracy:	0.8414 -	loss:	0.4690 -	val_accuracy:	0.6300 -	val_loss:	1.7034
25/25		21s	716ms/step -	accuracy:	0.8408 -	loss:	0.5160 -	val_accuracy:	0.5900 -	val_loss:	1.6704
-	64/200	10-	722/-+		0.0054	1	0.2450		0 6050		1 (405
25/25 Epoch	65/200	182	/zzms/step -	accuracy:	0.8854 -	1055:	0.3459 -	val_accuracy:	0.6050 -	Va1_1055:	1.6405
25/25		18s	723ms/step -	accuracy:	0.9229 -	loss:	0.2495 -	val_accuracy:	0.6100 -	val_loss:	1.6316
Epoch 25/25	66/200	200	720ms/ston -	accuracy:	0 8007 -	locci	0 2830 -	val accuracy:	0 6150 -	val loss:	1 5/127
	67/200	203	72011373ccp	accuracy.	0.0557	1033.	0.2030	var_accar acy:	0.0130	va1_1033.	1.5457
25/25		18s	713ms/step -	accuracy:	0.9092 -	loss:	0.2726 -	val_accuracy:	0.6000 -	val_loss:	1.7793
25/25	68/200	185	715ms/step -	accuracy:	0.9390 -	loss:	0.1926 -	val_accuracy:	0.5800 -	val loss:	2.0801
	69/200		•					_ ,		_	
25/25 Enach	70/200	- 18s	723ms/step -	accuracy:	0.9058 -	loss:	0.3264 -	val_accuracy:	0.6200 -	val_loss:	1.8446
25/25		18s	733ms/step -	accuracy:	0.8686 -	loss:	0.4121 -	val_accuracy:	0.6000 -	val_loss:	1.9129
-	71/200		/ /								
25/25 Epoch	72/200	185	/30ms/step -	accuracy:	u.8806 -	TOSS:	v.36/1 -	val_accuracy:	u.6050 -	var_loss:	1.6417
25/25		18s	705ms/step -	accuracy:	0.9386 -	loss:	0.1723 -	val_accuracy:	0.5900 -	val_loss:	1.7384
	73/200	10-	71 5 / 2		0 0272	1	0 2002		0 5050	unl lass.	2 1407
25/25 Epoch	74/200	182	/isms/step -	accuracy:	0.92/3 -	1055:	0.2003 -	val_accuracy:	0.5850 -	Va1_1055:	2.148/
25/25		18s	714ms/step -	accuracy:	0.9156 -	loss:	0.2440 -	val_accuracy:	0.6000 -	val_loss:	1.9785
Epoch 25/25	75/200	- 21c	729ms/stan -	accuracy:	0 9193 -	loss	0 2718 -	val_accuracy:	0 5850 -	val loss.	2 0705
	76/200	213	725111373CCP	accuracy.	0.5155	1033.	0.2710	var_accar acy:	0.3030	va1_1033.	2.0703
25/25		18s	724ms/step -	accuracy:	0.9459 -	loss:	0.1662 -	<pre>val_accuracy:</pre>	0.6100 -	val_loss:	2.0769
Epoch 25/25	77/200	- 18s	726ms/sten -	accuracv:	0.9097 -	loss:	0.2411 -	val_accuracy:	0.6250 -	val loss:	2.1611
	78/200		•	-				_ ,		_	
25/25		18s	733ms/step -	accuracy:	0.9575 -	loss:	0.1431 -	val_accuracy:	0.5850 -	val_loss:	2.3373
25/25	79/200 ——————————	18s	722ms/step -	accuracy:	0.9352 -	loss:	0.2195 -	val_accuracy:	0.6050 -	val_loss:	2.0010
Epoch	80/200		•	-				_ ,		_	
25/25 Enoch	81/200	18s	707ms/step -	accuracy:	0.9505 -	loss:	0.1423 -	val_accuracy:	0.5950 -	val_loss:	1.8676
25/25		18s	708ms/step -	accuracy:	0.9476 -	loss:	0.1579 -	val_accuracy:	0.5950 -	val_loss:	2.0369
-	82/200		722 / /		0.000	1.	0 1000		0 6455		4 004-
25/25		- 18s	/23ms/step -	accuracy:	0.9688 -	loss:	0.1092 -	val_accuracy:	Ø.6150 -	val_loss:	1.9845

Enach	92/200										
25/25	83/200	- 18s	731ms/step	- accuracy:	0.9479 -	loss:	0.1606	- val_accuracy:	0.5800	- val loss:	2.6820
Epoch	84/200							_ ,		_	
25/25		18s	726ms/step	- accuracy:	0.9545 -	loss:	0.1623	- val_accuracy:	0.6100	- val_loss:	1.9176
25/25	85/200 	18s	706ms/step	- accuracv:	0.9567 -	loss:	0.1448	- val_accuracy:	0.6100	- val loss:	2.0825
	86/200			,				_ ,		_	
25/25		18s	710ms/step	- accuracy:	0.9462 -	loss:	0.1407	- val_accuracy:	0.6050	- val_loss:	2.2098
25/25	87/200 	- 18s	721ms/step	- accuracv:	0.9378 -	loss:	0.1713	- val_accuracy:	0.5900	- val loss:	2.2017
	88/200		-,					,			
25/25		18s	716ms/step	- accuracy:	0.9610 -	loss:	0.1053	- val_accuracy:	0.6200	- val_loss:	2.2934
25/25	89/200 	18s	728ms/step	- accuracy:	0.9620 -	loss:	0.1244	- val_accuracy:	0.6000	- val_loss:	2.2978
-	90/200										
25/25 Enoch	91/200	- 18s	718ms/step	- accuracy:	0.9526 -	loss:	0.1334	- val_accuracy:	0.5850	- val_loss:	2.2852
25/25		18s	730ms/step	- accuracy:	0.9629 -	loss:	0.1154	- val_accuracy:	0.6000	- val_loss:	2.0158
-	92/200		/ /								
25/25 Enoch	93/200	- 18s	708ms/step	- accuracy:	0.9730 -	loss:	0.0986	- val_accuracy:	0.6100	- val_loss:	2.1760
25/25		18s	716ms/step	- accuracy:	0.9814 -	loss:	0.0740	- val_accuracy:	0.5800	- val_loss:	2.4375
	94/200	10-	702 / - +		0.0403	1	0 1256		0 6300		2 1506
25/25 Epoch	95/200	185	/03ms/step	- accuracy:	0.9482 -	1055:	0.1256	- val_accuracy:	0.6200	- va1_10ss:	2.1586
25/25		18s	715ms/step	- accuracy:	0.9786 -	loss:	0.0733	- val_accuracy:	0.6100	- val_loss:	2.1589
	96/200	100	721ms/ston	2661102611	0.0690	10001	0 0007	val accumacy:	0 5000	val loss.	2 4064
25/25 Epoch	97/200	105	/sims/scep	- accuracy.	0.9000 -	1055.	0.0907	- val_accuracy:	0.5900	- vai_1055.	2.4004
25/25		19s	754ms/step	- accuracy:	0.9696 -	loss:	0.0912	<pre>- val_accuracy:</pre>	0.5600	- val_loss:	2.4183
Epoch 25/25	98/200	19c	733mc/cton	- accuracy:	0 0788 -	locci	0 0655	- val_accuracy:	0 6050	- val loss:	2 2126
	99/200	103	733m3/3ccp	accur acy.	0.5700	1033.	0.0033	vai_accaracy.	0.0030	vai_1033.	2.2120
25/25		20s	722ms/step	- accuracy:	0.9470 -	loss:	0.1776	<pre>- val_accuracy:</pre>	0.6300	- val_loss:	2.1396
Epoch 25/25	100/200	185	709ms/sten	- accuracy:	0 9711 -	loss	0 0763	- val_accuracy:	0 6050	- val loss:	2 0356
	101/200		, oss, seep	uccui ucy i	0,107,22	2000.	0.0703	var_acca. acy v	0.0050		2,0330
25/25		18s	724ms/step	- accuracy:	0.9650 -	loss:	0.1307	<pre>- val_accuracy:</pre>	0.6050	- val_loss:	2.1148
25/25	102/200	- 18s	724ms/step	- accuracv:	0.9790 -	loss:	0.0682	- val_accuracy:	0.6000	- val loss:	2.6045
	103/200		-,					,			
25/25	104/200	18s	718ms/step	- accuracy:	0.9743 -	loss:	0.0784	- val_accuracy:	0.5750	- val_loss:	2.6396
25/25		18s	711ms/step	- accuracy:	0.9579 -	loss:	0.1212	- val_accuracy:	0.5750	- val_loss:	2.9462
-	105/200										
25/25 Enoch	106/200	- 18s	719ms/step	- accuracy:	0.9851 -	loss:	0.0671	- val_accuracy:	0.6000	- val_loss:	2.5684
25/25		18s	726ms/step	- accuracy:	0.9438 -	loss:	0.2108	- val_accuracy:	0.5700	- val_loss:	2.1696
-	107/200	10-	720 / - +		0.0563	1	0 1120		0 5050		2 4554
25/25 Epoch	108/200	185	/20ms/step	- accuracy:	0.9562 -	1055:	0.1139	- val_accuracy:	0.5850	- va1_10ss:	2.4554
25/25		18s	724ms/step	- accuracy:	0.9572 -	loss:	0.1481	<pre>- val_accuracy:</pre>	0.6150	- val_loss:	2.1096
Epoch 25/25	109/200	19c	713mc/cton	- accuracy:	0 0787 -	locci	0 0580	- val_accuracy:	0 5750	- val loss:	3 2846
	110/200	103	713m3/3ccp	accuracy.	0.5707	1033.	0.0303	var_accar acy.	0.3730	va1_1033.	3.2040
25/25		18s	714ms/step	- accuracy:	0.9564 -	loss:	0.1865	<pre>- val_accuracy:</pre>	0.6050	- val_loss:	2.1813
Epoch 25/25	111/200	185	717ms/step	- accuracy:	0.9763 -	loss:	0.0739	<pre>- val_accuracy:</pre>	0.5700	- val loss:	2.6695
	112/200		·	-						_	
25/25		18s	712ms/step	- accuracy:	0.9603 -	loss:	0.1191	- val_accuracy:	0.6050	- val_loss:	2.3246
25/25	113/200	18s	713ms/step	- accuracy:	0.9768 -	loss:	0.0602	- val_accuracy:	0.6150	- val_loss:	2.4997
	114/200					_					
25/25 Enoch	115/200	- 17s	696ms/step	- accuracy:	0.9840 -	loss:	0.0772	- val_accuracy:	0.6000	- val_loss:	2.7861
25/25		18s	716ms/step	- accuracy:	0.9773 -	loss:	0.1082	- val_accuracy:	0.6150	- val_loss:	2.3787
	116/200	10-	720		0.0026	1	0.0534		0 6000	val lass.	2 0112
25/25 Epoch	117/200	182	/20ms/step	- accuracy:	0.9836 -	1055:	0.0524	- val_accuracy:	0.6000	- val_1055:	2.9112
25/25		18s	721ms/step	- accuracy:	0.9603 -	loss:	0.1101	<pre>- val_accuracy:</pre>	0.5900	- val_loss:	2.7486
Epoch 25/25	118/200	19c	710ms/s+an	- accuracy.	0 9200	1055.	0 0725	- val_accuracy:	0 5050	- val locc	2 8456
	119/200	102	, 10m3/31ch	accui acy.	J. JOUJ -	1033.	0.0105	var_accuracy.	שניני.	va1_1055.	2.0400
25/25		17s	702ms/step	- accuracy:	0.9673 -	loss:	0.1098	<pre>- val_accuracy:</pre>	0.5950	- val_loss:	2.6416
Epoch 25/25	120/200	185	701ms/sten	- accuracy:	0.9882 -	loss	0.0370	- val_accuracy:	0.6100	- val loss:	2.6769
	121/200		15, 5 сер	ucy.		_,,,,,		accar acy.			
25/25		18s	712ms/step	- accuracy:	0.9638 -	loss:	0.1044	<pre>- val_accuracy:</pre>	0.6200	- val_loss:	2.9083
Epoch 25/25	122/200	18s	725ms/step	- accuracy:	0.9782 -	loss:	0.0979	- val_accuracy:	0.6200	- val loss:	2.5686
Epoch	123/200		·	-						_	
25/25		18s	729ms/step	- accuracy:	0.9812 -	loss:	0.0578	<pre>- val_accuracy:</pre>	0.6250	- val_loss:	2.3114

Epoch	124/200										
25/25 Enoch	125/200	18s	716ms/step -	accuracy:	0.9868 -	loss:	0.0367 -	<pre>- val_accuracy:</pre>	0.5950	val_loss:	2.8347
25/25		18s	702ms/step -	accuracy:	0.9719 -	loss:	0.0724 -	- val_accuracy:	0.6250	val_loss:	2.6391
Epoch 25/25	126/200	- 18s	714ms/step -	accuracy:	0.9764 -	loss:	0.0920 -	- val_accuracy:	0.5950	val_loss:	2.5164
Epoch 25/25	127/200	185	709ms/sten -	accuracy:	0.9781 -	loss:	0.0665 -	- val_accuracy:	0.6250	val loss:	2.5423
Epoch	128/200			-						_	
25/25 Epoch	129/200			-				- val_accuracy:		_	
25/25 Epoch	130/200	- 18s	718ms/step -	accuracy:	0.9756 -	loss:	0.0772 -	<pre>- val_accuracy:</pre>	0.5450	val_loss:	3.1431
25/25 Enoch	131/200	18s	738ms/step -	accuracy:	0.9635 -	loss:	0.0942 -	<pre>- val_accuracy:</pre>	0.6100	val_loss:	2.8132
25/25		18s	711ms/step -	accuracy:	0.9839 -	loss:	0.0396 -	- val_accuracy:	0.5900	val_loss:	2.7533
Epoch 25/25	132/200	18s	720ms/step -	accuracy:	0.9689 -	loss:	0.1105 -	- val_accuracy:	0.6100	val_loss:	2.5281
Epoch 25/25	133/200	- 18s	710ms/step -	accuracy:	0.9722 -	loss:	0.0740 -	<pre>- val_accuracy:</pre>	0.5850	· val loss:	3.1726
Epoch 25/25	134/200	- 18c	71/ms/sten -	accuracy:	0 9883 -	loss	0 0474 -	- val_accuracy:	0 6200 .	· val loss:	2 6395
Epoch	135/200			-						_	
25/25 Epoch	136/200			-				- val_accuracy:		_	
25/25 Epoch	137/200	- 18s	728ms/step -	accuracy:	0.9670 -	loss:	0.0725 -	<pre>- val_accuracy:</pre>	0.6050	· val_loss:	2.2946
25/25 Enoch	138/200	18s	715ms/step -	accuracy:	0.9915 -	loss:	0.0330 -	<pre>- val_accuracy:</pre>	0.6050	val_loss:	2.6592
25/25		18s	717ms/step -	accuracy:	0.9781 -	loss:	0.0669 -	<pre>- val_accuracy:</pre>	0.6100	val_loss:	2.7602
25/25	139/200	18s	708ms/step -	accuracy:	0.9688 -	loss:	0.1116 -	- val_accuracy:	0.6100	val_loss:	3.3575
Epoch 25/25	140/200	- 17s	691ms/step -	accuracy:	0.9746 -	loss:	0.1208 -	- val_accuracy:	0.6200	val_loss:	2.8369
Epoch 25/25	141/200	- 18s	702ms/step -	· accuracv:	0.9881 -	loss:	0.0348 -	- val_accuracy:	0.6050	· val loss:	2.4070
	142/200			-				- val_accuracy:		_	
Epoch	143/200			-						_	
25/25 Epoch	144/200			-				- val_accuracy:		_	
25/25 Epoch	145/200	- 18s	727ms/step -	accuracy:	0.9570 -	loss:	0.1407 -	- val_accuracy:	0.6050	· val_loss:	3.0398
25/25 Enoch	146/200	18s	722ms/step -	accuracy:	0.9857 -	loss:	0.0526 -	- val_accuracy:	0.6100	val_loss:	2.8026
25/25		18s	704ms/step -	accuracy:	0.9799 -	loss:	0.0990 -	- val_accuracy:	0.5900	val_loss:	2.9637
25/25		18s	711ms/step -	accuracy:	0.9758 -	loss:	0.0744 -	- val_accuracy:	0.6050	val_loss:	2.6072
25/25	148/200	18s	714ms/step -	accuracy:	0.9780 -	loss:	0.0572 -	- val_accuracy:	0.5950	val_loss:	2.8387
Epoch 25/25	149/200	- 18s	729ms/step -	accuracy:	0.9769 -	loss:	0.0605 -	<pre>- val_accuracy:</pre>	0.5850	val_loss:	2.7072
Epoch 25/25	150/200	185	722ms/sten -	accuracy:	0.9864 -	loss:	0.0305 -	- val_accuracy:	0.6000	· val loss:	3.7635
Epoch	151/200			-						_	
25/25 Epoch	152/200			-				- val_accuracy:		_	
25/25 Epoch	153/200	- 18s	722ms/step -	accuracy:	0.9956 -	loss:	0.0176 -	- val_accuracy:	0.6300	· val_loss:	2.9177
25/25 Epoch	154/200	18s	721ms/step -	accuracy:	0.9782 -	loss:	0.0799 -	<pre>- val_accuracy:</pre>	0.5900	val_loss:	2.9872
25/25		18s	728ms/step -	accuracy:	0.9934 -	loss:	0.0160 -	<pre>- val_accuracy:</pre>	0.5900	val_loss:	3.0310
25/25		19s	762ms/step -	accuracy:	0.9743 -	loss:	0.0986 -	- val_accuracy:	0.5950	val_loss:	2.8097
25/25	156/200	17 s	684ms/step -	accuracy:	0.9687 -	loss:	0.1124 -	- val_accuracy:	0.6150	val_loss:	2.6391
Epoch 25/25	157/200	18s	722ms/step -	accuracy:	0.9891 -	loss:	0.0306 -	- val_accuracy:	0.6150	· val_loss:	2.9222
Epoch 25/25	158/200	- 17s	685ms/step -	· accuracv:	0.9726 -	loss:	0.0740 -	- val_accuracy:	0.6000	· val loss:	2.8231
	159/200			-				- val_accuracy:		_	
Epoch	160/200			-						_	
-	161/200			-				- val_accuracy:		_	
25/25 Epoch	162/200	18s	708ms/step -	accuracy:	0.9797 -	loss:	0.0624 -	- val_accuracy:	0.5900	val_loss:	2.7295
25/25		18s	731ms/step -	accuracy:	0.9947 -	loss:	0.0261 -	- val_accuracy:	0.6050	val_loss:	2.9180
25/25		18s	736ms/step -	accuracy:	0.9889 -	loss:	0.0323 -	- val_accuracy:	0.6050	val_loss:	2.9152
25/25	164/200	- 19s	741ms/step -	accuracy:	0.9805 -	loss:	0.0453 -	- val_accuracy:	0.6100	val_loss:	2.8851

```
Epoch 165/200
                                 18s 726ms/step - accuracy: 0.9907 - loss: 0.0331 - val_accuracy: 0.6150 - val_loss: 3.0317
       25/25
       Epoch 166/200
       25/25
                                 - 17s 688ms/step - accuracy: 0.9721 - loss: 0.0709 - val_accuracy: 0.6100 - val_loss: 3.0954
       Epoch 167/200
      25/25
                                 18s 713ms/step - accuracy: 0.9757 - loss: 0.0530 - val_accuracy: 0.6050 - val_loss: 3.2853
       Epoch 168/200
                                 18s 706ms/step - accuracy: 0.9767 - loss: 0.0868 - val_accuracy: 0.6100 - val_loss: 2.8212
      25/25
       Epoch 169/200
      25/25
                                 - 18s 721ms/step - accuracy: 0.9894 - loss: 0.0469 - val_accuracy: 0.6000 - val_loss: 3.0674
      Epoch 170/200
      25/25
                                 - 18s 718ms/step - accuracy: 0.9883 - loss: 0.0524 - val_accuracy: 0.5900 - val_loss: 2.7765
       Epoch 171/200
                                 18s 714ms/step - accuracy: 0.9719 - loss: 0.0858 - val_accuracy: 0.5950 - val_loss: 2.9619
      25/25
       Epoch 172/200
       25/25
                                 · 18s 723ms/step - accuracy: 0.9967 - loss: 0.0159 - val_accuracy: 0.5950 - val_loss: 3.2934
       Epoch 173/200
                                 18s 702ms/step - accuracy: 0.9748 - loss: 0.0891 - val_accuracy: 0.5900 - val_loss: 2.9908
      25/25
      Epoch 174/200
       25/25
                                 - 17s 694ms/step - accuracy: 0.9945 - loss: 0.0353 - val_accuracy: 0.6050 - val_loss: 3.2003
      Epoch 175/200
       25/25
                                 18s 710ms/step - accuracy: 0.9945 - loss: 0.0238 - val_accuracy: 0.5550 - val_loss: 5.2057
       Epoch 176/200
      25/25
                                 18s 728ms/step - accuracy: 0.9718 - loss: 0.1247 - val_accuracy: 0.5800 - val_loss: 3.1216
       Epoch 177/200
                                 - 18s 718ms/step - accuracy: 0.9788 - loss: 0.0614 - val_accuracy: 0.6100 - val_loss: 3.3939
       25/25
       Epoch 178/200
      25/25
                                 • 18s 729ms/step - accuracy: 0.9917 - loss: 0.0176 - val_accuracy: 0.5800 - val_loss: 3.5900
      Epoch 179/200
                                 18s 720ms/step - accuracy: 0.9924 - loss: 0.0369 - val_accuracy: 0.6100 - val_loss: 3.5684
       25/25
       Epoch 180/200
      25/25
                                - 18s 716ms/step - accuracy: 0.9919 - loss: 0.0242 - val_accuracy: 0.6100 - val_loss: 3.8176
       Epoch 181/200
      25/25
                                 - 18s 703ms/step - accuracy: 0.9974 - loss: 0.0112 - val_accuracy: 0.6200 - val_loss: 3.5469
       Epoch 182/200
       25/25
                                 - 18s 722ms/step - accuracy: 0.9864 - loss: 0.0510 - val_accuracy: 0.6000 - val_loss: 3.8091
       Epoch 183/200
       25/25
                                 18s 723ms/step - accuracy: 0.9632 - loss: 0.1131 - val_accuracy: 0.5950 - val_loss: 3.4164
       Epoch 184/200
      25/25
                                 - 18s 732ms/step - accuracy: 0.9814 - loss: 0.1048 - val_accuracy: 0.5850 - val_loss: 2.7835
       Epoch 185/200
       25/25
                                 - 18s 718ms/step - accuracy: 0.9960 - loss: 0.0176 - val_accuracy: 0.5900 - val_loss: 3.4109
       Epoch 186/200
      25/25
                                 - 18s 710ms/step - accuracy: 0.9937 - loss: 0.0333 - val_accuracy: 0.5950 - val_loss: 3.6571
       Epoch 187/200
       25/25
                                 17s 681ms/step - accuracy: 0.9916 - loss: 0.0364 - val_accuracy: 0.6050 - val_loss: 3.6589
       Epoch 188/200
      25/25
                                 - 18s 714ms/step - accuracy: 0.9962 - loss: 0.0138 - val_accuracy: 0.6000 - val_loss: 3.2003
       Epoch 189/200
       25/25
                                 - 18s 723ms/step - accuracy: 0.9920 - loss: 0.0247 - val_accuracy: 0.5800 - val_loss: 3.8158
      Epoch 190/200
      25/25
                                 - 18s 718ms/step - accuracy: 0.9831 - loss: 0.0737 - val_accuracy: 0.6250 - val_loss: 3.4735
      Epoch 191/200
       25/25
                                 18s 719ms/step - accuracy: 0.9852 - loss: 0.0763 - val_accuracy: 0.6000 - val_loss: 3.5289
       Epoch 192/200
      25/25
                                 - 18s 706ms/step - accuracy: 0.9861 - loss: 0.0350 - val_accuracy: 0.6000 - val_loss: 3.7007
       Epoch 193/200
       25/25
                                 - 17s 697ms/step - accuracy: 0.9921 - loss: 0.0248 - val_accuracy: 0.6050 - val_loss: 3.6667
       Epoch 194/200
                                 18s 708ms/step - accuracy: 0.9772 - loss: 0.0782 - val_accuracy: 0.6200 - val_loss: 4.1763
      25/25
      Epoch 195/200
       25/25
                                  20s 703ms/step - accuracy: 0.9855 - loss: 0.0633 - val_accuracy: 0.6000 - val_loss: 3.3390
      Epoch 196/200
      25/25
                                 - 18s 710ms/step - accuracy: 0.9920 - loss: 0.0385 - val_accuracy: 0.5800 - val_loss: 5.2956
       Epoch 197/200
      25/25
                                 · 18s 703ms/step - accuracy: 0.9785 - loss: 0.0978 - val_accuracy: 0.5950 - val_loss: 4.0987
       Epoch 198/200
       25/25
                                 · 18s 729ms/step - accuracy: 0.9750 - loss: 0.0530 - val_accuracy: 0.6100 - val_loss: 4.0447
       Epoch 199/200
       25/25
                                 18s 705ms/step - accuracy: 0.9891 - loss: 0.0457 - val_accuracy: 0.5850 - val_loss: 3.7655
       Epoch 200/200
       25/25
                                 · 17s 695ms/step - accuracy: 0.9887 - loss: 0.0432 - val_accuracy: 0.5750 - val_loss: 4.1484
In [8]: print(f"Execution time: {elapsed_time:.2f} seconds")
```

```
In [9]: def append_core_data(score_path, num_cores, elapsed_time):
    # Check if the file already exists
    file_exists = os.path.exists(score_path)

# Open the file in append mode
```

with open(score_path, mode='a', newline='') as file:

Execution time: 3590.61 seconds

```
writer = csv.writer(file)

# If the file is new, write the header
if not file_exists:
    writer.writerow(["Number of Cores", "Elapsed Time"])

# Write the new data
writer.writerow([num_cores, elapsed_time])
```

In [10]: score_path = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DEEP LEARNING WITH HPSC\core_data.txt"
 append_core_data(score_path, number_of_worker, elapsed_time)

Import Req Lib

```
In [1]: %matplotlib inline
        import shutil
        import random
        import numpy as np
        from warnings import filterwarnings
        filterwarnings('ignore')
        from tensorflow.keras import layers, regularizers, optimizers
        from tensorflow.keras import models
        from tensorflow.keras.models import Sequential, Model
        from tensorflow.keras.layers import LeakyReLU,Dense, Activation, Flatten, Dropout, BatchNormalization,Conv2D, MaxPooling2D
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import tensorflow as tf
        import os
        import time
        import csv
        from matplotlib import figure
```

Define 3 worker

```
In [2]: # Set the number of threads
number_of_worker = 3
os.environ['OMP_NUM_THREADS'] = '3' # OpenMP threads for parallelism
os.environ['TF_NUM_INTEROP_THREADS'] = '3' # Threads for inter-operation parallelism
os.environ['TF_NUM_INTRAOP_THREADS'] = '3' # Threads for intra-operation parallelism
# Confirm TensorFlow is using the specified number of threads
tf.config.threading.set_inter_op_parallelism_threads(number_of_worker)
tf.config.threading.set_intra_op_parallelism_threads(number_of_worker)
```

Train Val data Split

```
In [3]: source_dir = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DATA\Convert_Audio_File_to_jpg_file"
        target_dir = r'genres_train_val_split_data'
        split_ratio = 0.8
        def Train_Test_Split(source_dir,target_dir,split_ratio):
            # Define source and target directories
            train_dir = os.path.join(target_dir, 'train')
            val_dir = os.path.join(target_dir, 'val')
            # Create target directories if they don't exist
            os.makedirs(train_dir, exist_ok=True)
            os.makedirs(val_dir, exist_ok=True)
            # Get the list of class directories
            classes = [d for d in os.listdir(source_dir) if os.path.isdir(os.path.join(source_dir, d))]
            for class name in classes:
                # Create class directories in train and val folders
                os.makedirs(os.path.join(train_dir, class_name), exist_ok=True)
                os.makedirs(os.path.join(val_dir, class_name), exist_ok=True)
                # Get list of images in the class directory
                class_dir = os.path.join(source_dir, class_name)
                images = [f for f in os.listdir(class_dir) if os.path.isfile(os.path.join(class_dir, f))]
                # Shuffle the images
                random.shuffle(images)
                # Compute the split point
                split_point = int(len(images) * split_ratio)
                # Split the images into training and validation sets
                train images = images[:split point]
                val_images = images[split_point:]
```

In [4]: Train_Test_Split(source_dir,target_dir,split_ratio)

Data split completed successfully!

Load the Data

```
In [5]: WIDTH = 64
        HEIGHT = 64
        BATCH_SIZE = 32
        TRAIN_DIR=r'genres_train_val_split_data/train'
        val_dir = r'genres_train_val_split_data/val'
        # data prep
        train_datagen = ImageDataGenerator(
            rescale=1./255.,validation_split=0.25)
        train_generator = train_datagen.flow_from_directory(
            TRAIN_DIR,
            target size=(HEIGHT, WIDTH),
                batch_size=BATCH_SIZE,
                class_mode='categorical')
        validation_gen = train_datagen.flow_from_directory(
            val_dir,target_size = (HEIGHT,WIDTH),
            batch_size = BATCH_SIZE,
            class_mode = 'categorical'
```

Found 800 images belonging to 10 classes. Found 200 images belonging to 10 classes.

Model Architecture

```
In [6]: model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                         input_shape=(64,64,3)))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(128, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(128, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(10, activation='softmax'))
        model.compile(optimizers.RMSprop(learning_rate=0.0005, decay=1e-6),loss="categorical_crossentropy",metrics=["accuracy"])
        model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 64, 64, 32)	896
activation (Activation)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 62, 62, 64)	18,496
activation_1 (Activation)	(None, 62, 62, 64)	0
max_pooling2d (MaxPooling2D)	(None, 31, 31, 64)	0
dropout (Dropout)	(None, 31, 31, 64)	0
conv2d_2 (Conv2D)	(None, 31, 31, 64)	36,928
activation_2 (Activation)	(None, 31, 31, 64)	0
conv2d_3 (Conv2D)	(None, 29, 29, 64)	36,928
activation_3 (Activation)	(None, 29, 29, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_1 (Dropout)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 14, 14, 128)	73,856
activation_4 (Activation)	(None, 14, 14, 128)	0
conv2d_5 (Conv2D)	(None, 12, 12, 128)	147,584
activation_5 (Activation)	(None, 12, 12, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_2 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2,359,808
activation_6 (Activation)	(None, 512)	0
dropout_3 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5,130

Total params: 2,679,626 (10.22 MB)

Trainable params: 2,679,626 (10.22 MB)

Non-trainable params: 0 (0.00 B)

```
In [7]: STEP_SIZE_TRAIN=train_generator.n//train_generator.batch_size
# Measure the execution time
start_time = time.time()

model.fit(train_generator,validation_data=validation_gen,epochs=200)

end_time = time.time()
elapsed_time = end_time - start_time
```

Epoch		17 a F24m	- / - t n		0 1172	1	2 2200	val accumant	0 1000	wal lass.	2 2000
25/25 Epoch		1/5 5241113	s/step -	accuracy:	0.11/2 -	1055:	2.3200 -	val_accuracy:	0.1000 -	Va1_1055:	2.2990
25/25 Epoch	3/200	13s 532ms	s/step -	accuracy:	0.1234 -	loss:	2.2973 -	val_accuracy:	0.2100 -	val_loss:	2.1400
25/25 Enoch	4/200	11s 446ms	s/step -	accuracy:	0.1766 -	loss:	2.2264 -	val_accuracy:	0.1950 -	val_loss:	2.1511
25/25		11s 450ms	s/step -	accuracy:	0.2260 -	loss:	2.0911 -	val_accuracy:	0.2800 -	val_loss:	1.9925
Epoch 25/25		11s 451ms	s/step -	accuracy:	0.2132 -	loss:	2.0567 -	val_accuracy:	0.3100 -	val_loss:	1.9550
Epoch 25/25		12s 498m	s/sten -	accuracy:	0.2552 -	loss:	1.9979 -	val accuracy:	0.2850 -	val loss:	1.9534
Epoch	7/200		•					_ ,		_	
25/25 Epoch	8/200		•	-				val_accuracy:		_	
25/25 Epoch	9/200	12s 482ms	s/step -	accuracy:	0.2902 -	loss:	1.9696 -	val_accuracy:	0.3850 -	val_loss:	1.8503
25/25 Epoch	10/200	12s 477ms	s/step -	accuracy:	0.3088 -	loss:	1.9267 -	val_accuracy:	0.3300 -	val_loss:	1.8287
25/25		13s 508ms	s/step -	accuracy:	0.3537 -	loss:	1.8062 -	val_accuracy:	0.3650 -	val_loss:	1.8139
25/25		14s 565ms	s/step -	accuracy:	0.3134 -	loss:	1.8423 -	val_accuracy:	0.3800 -	val_loss:	1.7176
Epoch 25/25	12/200	12s 460ms	s/step -	accuracy:	0.3454 -	loss:	1.7533 -	val_accuracy:	0.3850 -	val_loss:	1.6785
Epoch 25/25	13/200	12s 460ms	s/step -	accuracy:	0.4122 -	loss:	1.7147 -	val_accuracy:	0.3450 -	val loss:	1.7537
Epoch 25/25	14/200	12c //72mg	s/stan =	accuracy.	0 3678 -	loss.	1 6829 -	val_accuracy:	0 3700 -	val loss:	1 7117
Epoch	15/200		•							_	
25/25 Epoch	16/200		•					val_accuracy:		_	
25/25 Epoch	17/200	12s 464ms	s/step -	accuracy:	0.4147 -	loss:	1.6018 -	val_accuracy:	0.3400 -	val_loss:	1.8386
25/25 Enoch	18/200	12s 474ms	s/step -	accuracy:	0.4380 -	loss:	1.5571 -	val_accuracy:	0.4600 -	val_loss:	1.4757
25/25		12s 469ms	s/step -	accuracy:	0.4516 -	loss:	1.4895 -	val_accuracy:	0.3950 -	val_loss:	1.6467
25/25		11s 451ms	s/step -	accuracy:	0.4831 -	loss:	1.3935 -	val_accuracy:	0.3550 -	val_loss:	1.6572
Epoch 25/25	20/200	11s 450ms	s/step -	accuracy:	0.4701 -	loss:	1.4577 -	val_accuracy:	0.4850 -	val_loss:	1.4225
Epoch 25/25	21/200	12s 475ms	s/step -	accuracy:	0.5340 -	loss:	1.3667 -	val_accuracy:	0.4450 -	val loss:	1.4603
	22/200		•	-				val_accuracy:		_	
Epoch	23/200		•	-						_	
25/25 Epoch	24/200	12s 4/6ms	s/step -	accuracy:	0.5124 -	loss:	1.3269 -	val_accuracy:	0.4250 -	val_loss:	1.4842
25/25 Epoch	25/200	11s 454ms	s/step -	accuracy:	0.5223 -	loss:	1.3252 -	val_accuracy:	0.4550 -	val_loss:	1.4401
25/25 Epoch	26/200	11s 430ms	s/step -	accuracy:	0.5324 -	loss:	1.3002 -	val_accuracy:	0.4650 -	val_loss:	1.3797
25/25		12s 480ms	s/step -	accuracy:	0.5712 -	loss:	1.1731 -	val_accuracy:	0.4650 -	val_loss:	1.4238
25/25		11s 442ms	s/step -	accuracy:	0.5879 -	loss:	1.2535 -	val_accuracy:	0.5150 -	val_loss:	1.4551
25/25	28/200	12s 471ms	s/step -	accuracy:	0.5800 -	loss:	1.1953 -	val_accuracy:	0.4650 -	val_loss:	1.4764
Epoch 25/25	29/200 —————————	11s 439ms	s/step -	accuracy:	0.6032 -	loss:	1.0950 -	val_accuracy:	0.4900 -	val_loss:	1.3933
Epoch 25/25	30/200	11s 432m	s/sten -	accuracy:	0.6123 -	loss:	1.0926 -	val_accuracy:	0.5650 -	val loss:	1.3480
	31/200		•	-				<pre>val_accuracy:</pre>		_	
Epoch	32/200		•	-						_	
25/25 Epoch	33/200	11s 433ms	s/step -	accuracy:	0.6167 -	loss:	1.0415 -	val_accuracy:	0.5100 -	val_loss:	1.4623
25/25 Epoch	34/200	11s 447ms	s/step -	accuracy:	0.5989 -	loss:	1.0725 -	val_accuracy:	0.5750 -	val_loss:	1.2746
25/25		11s 435ms	s/step -	accuracy:	0.6767 -	loss:	0.9582 -	val_accuracy:	0.5500 -	val_loss:	1.2775
25/25		11s 452ms	s/step -	accuracy:	0.6763 -	loss:	0.9264 -	val_accuracy:	0.5600 -	val_loss:	1.3444
25/25	36/200	11s 438ms	s/step -	accuracy:	0.6904 -	loss:	0.8116 -	val_accuracy:	0.4150 -	val_loss:	1.8726
Epoch 25/25	37/200	11s 449ms	s/step -	accuracy:	0.6764 -	loss:	0.9323 -	val_accuracy:	0.5450 -	val_loss:	1.4784
Epoch 25/25	38/200	11s 447m	s/sten -	accuracv:	0.7223 -	loss:	0.8499 -	val_accuracy:	0.5700 -	val loss:	1.3681
Epoch	39/200		•	-				<pre>val_accuracy:</pre>		_	
-	40/200		•	-						_	
25/25 Epoch	41/200	11s 442ms	s/step -	accuracy:	U./209 -	TOSS:	⊌./767 -	val_accuracy:	U.5100 -	var_loss:	1.3763
25/25		11s 456ms	s/step -	accuracy:	0.7288 -	loss:	0.8238 -	val_accuracy:	0.5850 -	val_loss:	1.2973

	42/200	11c /29mc/cton 25cunacy, 0.7672 loca, 0.7206 wal accuracy, 0.4550 wal loca, 1.712
	43/200	11s 438ms/step - accuracy: 0.7672 - loss: 0.7296 - val_accuracy: 0.4550 - val_loss: 1.712
25/25 Epoch	44/200	11s 445ms/step - accuracy: 0.7301 - loss: 0.7281 - val_accuracy: 0.4950 - val_loss: 1.539
25/25		11s 458ms/step - accuracy: 0.7694 - loss: 0.6188 - val_accuracy: 0.5400 - val_loss: 1.492
25/25	45/200	11s 451ms/step - accuracy: 0.7854 - loss: 0.6459 - val_accuracy: 0.5050 - val_loss: 1.696
Epoch 25/25	46/200	13s 502ms/step - accuracy: 0.7587 - loss: 0.7253 - val_accuracy: 0.5800 - val_loss: 1.469
	47/200	
25/25 Epoch	48/200	12s 486ms/step - accuracy: 0.8226 - loss: 0.5347 - val_accuracy: 0.5500 - val_loss: 1.469
25/25 Epoch	49/200	12s 494ms/step - accuracy: 0.8037 - loss: 0.5382 - val_accuracy: 0.5500 - val_loss: 1.445
25/25		12s 464ms/step - accuracy: 0.8286 - loss: 0.4913 - val_accuracy: 0.5650 - val_loss: 1.396
25/25		10s 408ms/step - accuracy: 0.8572 - loss: 0.4218 - val_accuracy: 0.5700 - val_loss: 1.549
Epoch 25/25	51/200	11s 439ms/step - accuracy: 0.8326 - loss: 0.4634 - val_accuracy: 0.5250 - val_loss: 1.872
Epoch 25/25	52/200	12s 462ms/step - accuracy: 0.8203 - loss: 0.5689 - val_accuracy: 0.5500 - val_loss: 1.634
Epoch	53/200	
25/25 Epoch	54/200	13s 499ms/step - accuracy: 0.8558 - loss: 0.4413 - val_accuracy: 0.5250 - val_loss: 1.711
25/25 Enoch	55/200	13s 517ms/step - accuracy: 0.8577 - loss: 0.4649 - val_accuracy: 0.5750 - val_loss: 1.553
25/25		13s 510ms/step - accuracy: 0.8622 - loss: 0.3862 - val_accuracy: 0.5800 - val_loss: 1.442
25/25	56/200	13s 500ms/step - accuracy: 0.8415 - loss: 0.4130 - val_accuracy: 0.5600 - val_loss: 1.629
Epoch 25/25	57/200	12s 473ms/step - accuracy: 0.8331 - loss: 0.4591 - val_accuracy: 0.5550 - val_loss: 1.761
Epoch	58/200	
25/25 Epoch	59/200	13s 512ms/step - accuracy: 0.8447 - loss: 0.4063 - val_accuracy: 0.5600 - val_loss: 1.702
25/25 Epoch	60/200	13s 497ms/step - accuracy: 0.8745 - loss: 0.3720 - val_accuracy: 0.5500 - val_loss: 1.759
25/25 Enoch	61/200	12s 484ms/step - accuracy: 0.8637 - loss: 0.3368 - val_accuracy: 0.5600 - val_loss: 1.479
25/25		12s 458ms/step - accuracy: 0.9153 - loss: 0.2569 - val_accuracy: 0.5450 - val_loss: 1.611
Epoch 25/25	62/200	12s 471ms/step - accuracy: 0.9336 - loss: 0.2270 - val_accuracy: 0.5000 - val_loss: 2.413
Epoch 25/25	63/200	11s 418ms/step - accuracy: 0.9275 - loss: 0.2432 - val_accuracy: 0.5550 - val_loss: 2.104
Epoch	64/200	
25/25 Epoch	65/200	11s 426ms/step - accuracy: 0.9020 - loss: 0.2548 - val_accuracy: 0.5650 - val_loss: 1.960
25/25 Epoch	66/200	11s 445ms/step - accuracy: 0.9056 - loss: 0.2923 - val_accuracy: 0.6000 - val_loss: 1.782
25/25 Enoch	67/200	11s 419ms/step - accuracy: 0.9148 - loss: 0.2506 - val_accuracy: 0.5550 - val_loss: 1.818
25/25		11s 441ms/step - accuracy: 0.9171 - loss: 0.2525 - val_accuracy: 0.5900 - val_loss: 1.813
25/25	68/200	11s 447ms/step - accuracy: 0.9299 - loss: 0.2053 - val_accuracy: 0.5550 - val_loss: 2.369
Epoch 25/25	69/200	11s 448ms/step - accuracy: 0.9197 - loss: 0.2721 - val_accuracy: 0.5500 - val_loss: 2.192
Epoch	70/200	
-	71/200	12s 459ms/step - accuracy: 0.9512 - loss: 0.1595 - val_accuracy: 0.5800 - val_loss: 2.380
25/25 Epoch	72/200	12s 480ms/step - accuracy: 0.9184 - loss: 0.2301 - val_accuracy: 0.5500 - val_loss: 2.023
25/25 Enoch	73/200	13s 522ms/step - accuracy: 0.9068 - loss: 0.2479 - val_accuracy: 0.5700 - val_loss: 2.173
25/25		12s 479ms/step - accuracy: 0.9345 - loss: 0.1671 - val_accuracy: 0.5800 - val_loss: 1.801
25/25	74/200	12s 466ms/step - accuracy: 0.9508 - loss: 0.1823 - val_accuracy: 0.5800 - val_loss: 1.984
Epoch 25/25	75/200 	11s 450ms/step - accuracy: 0.9300 - loss: 0.2190 - val_accuracy: 0.5750 - val_loss: 2.536
Epoch	76/200	
25/25 Epoch	77/200	12s 464ms/step - accuracy: 0.9385 - loss: 0.1612 - val_accuracy: 0.5600 - val_loss: 2.364
25/25 Epoch	78/200	12s 486ms/step - accuracy: 0.9416 - loss: 0.1744 - val_accuracy: 0.6050 - val_loss: 2.223
25/25		12s 466ms/step - accuracy: 0.9580 - loss: 0.1393 - val_accuracy: 0.5850 - val_loss: 2.232
25/25		12s 466ms/step - accuracy: 0.9435 - loss: 0.1561 - val_accuracy: 0.6100 - val_loss: 2.264
Epoch 25/25	80/200	12s 475ms/step - accuracy: 0.9536 - loss: 0.1949 - val_accuracy: 0.5950 - val_loss: 2.230
Epoch 25/25	81/200	12s 466ms/step - accuracy: 0.9532 - loss: 0.1703 - val_accuracy: 0.5700 - val_loss: 2.264
Epoch	82/200	
25/25		11s 459ms/step - accuracy: 0.9617 - loss: 0.0973 - val_accuracy: 0.5700 - val_loss: 2.414

	00.4000										
Epoch 25/25	83/200	- 11s	450ms/sten -	accuracy:	0 9347 -	loss	a 1793 .	- val_accuracy:	0 5800 -	val loss.	2 2786
	84/200		43011137 3 6 6 7	accui acy i	0.3347	1033.	0.1755	var_accar acy.	0.3000	vu1_1055.	2.2700
25/25		- 11s	449ms/step -	accuracy:	0.9474 -	loss:	0.1033	<pre>- val_accuracy:</pre>	0.5750 -	<pre>val_loss:</pre>	2.3670
Epoch 25/25	85/200	- 11c	152mc/cton	2661102641	0 0549	locci	0 1271	- val_accuracy:	0 5600	val loss:	2 4042
	86/200	113	4551113/3CEP -	accui acy.	0.5546 -	1033.	0.13/1	- vai_accuracy.	0.3000	va1_1033.	2.4043
25/25		- 11 s	455ms/step -	accuracy:	0.9691 -	loss:	0.0769	- val_accuracy:	0.5900 -	val_loss:	2.4306
-	87/200	124	F00ms /ston		0.0630	1	0 1250		0 5050	unl lange	2 1000
25/25 Epoch	88/200	135	509ms/step -	accuracy:	0.9638 -	1055:	0.1350	- val_accuracy:	0.5850 -	va1_1055:	2.1009
25/25		- 13s	520ms/step -	accuracy:	0.9666 -	loss:	0.0947	- val_accuracy:	0.5700 -	val_loss:	2.6385
	89/200	12-	462ma /atam		0.0640	1	0 1116		0 5050	unl lange	2 4650
25/25 Epoch	90/200	125	403111S/Step -	accuracy.	0.9046 -	1055.	0.1110	- val_accuracy:	0.3630 -	Va1_1055.	2.4059
25/25		12s	472ms/step -	accuracy:	0.9513 -	loss:	0.1856	<pre>- val_accuracy:</pre>	0.6000 -	val_loss:	2.4104
Epoch 25/25	91/200	126	179ms /s+on	26611112671	0 0626	locci	0 1200	- val_accuracy:	0 6000	val loss:	2 7544
	92/200	123	470113/3CEP -	accui acy.	0.5050 -	1033.	0.1250	- vai_accuracy.	0.0000	vai_1033.	2.7344
25/25		12 s	466ms/step -	accuracy:	0.9492 -	loss:	0.1673	<pre>- val_accuracy:</pre>	0.5850 -	<pre>val_loss:</pre>	2.5110
Epoch 25/25	93/200	- 12c	158mc/cton -	accuracy.	0 9847 -	1000	0 0773 .	- val_accuracy:	0 5650 -	val loss.	2 4438
	94/200	123	430m3/3ccp	accuracy.	0.5047	1033.	0.0773	vai_accuracy.	0.3030	va1_1033.	2.4430
25/25		11 s	446ms/step -	accuracy:	0.9749 -	loss:	0.0914	<pre>- val_accuracy:</pre>	0.5900 -	val_loss:	2.8167
Epoch 25/25	95/200	125	463ms/sten -	accuracy:	0.9696 -	loss:	0.0948	<pre>- val_accuracy:</pre>	0.5450 -	val loss:	2.9110
	96/200		, s ccp	uccu. ucy i	0.000	2000.	0.02.0	var_acca. acy.	0.5.50	.01_1000.	
25/25		12s	485ms/step -	accuracy:	0.9595 -	loss:	0.1383	<pre>- val_accuracy:</pre>	0.5800 -	val_loss:	2.2549
25/25	97/200	- 20s	463ms/step -	accuracy:	0.9758 -	loss:	0.0668	- val_accuracy:	0.5550 -	val loss:	2.1897
	98/200		•	•				_ ,		_	
25/25		- 11s	458ms/step -	accuracy:	0.9547 -	loss:	0.1388	<pre>- val_accuracy:</pre>	0.5750 -	val_loss:	2.3984
25/25	99/200	- 11s	452ms/step -	accuracy:	0.9875 -	loss:	0.0453	- val_accuracy:	0.5850 -	val loss:	2.7185
	100/200			-						_	
25/25 Enach	101/200	12s	470ms/step -	accuracy:	0.9522 -	loss:	0.1185	- val_accuracy:	0.5650 -	val_loss:	2.8550
25/25		- 12s	466ms/step -	accuracy:	0.9707 -	loss:	0.0789	- val_accuracy:	0.5600 -	val_loss:	2.9188
-	102/200							_			
25/25 Enoch	103/200	- 11s	453ms/step -	accuracy:	0.9429 -	loss:	0.1708	- val_accuracy:	0.5800 -	val_loss:	2.5333
25/25		12 s	460ms/step -	accuracy:	0.9541 -	loss:	0.1467	- val_accuracy:	0.5650 -	val_loss:	2.8877
-	104/200	40	466 / 1		0.0750	,	0.0700		0 5700		2 0475
25/25 Epoch	105/200	125	466ms/step -	accuracy:	0.9750 -	1055:	0.0700	<pre>- val_accuracy:</pre>	0.5/00 -	vai_ioss:	2.94/5
25/25		- 11 s	453ms/step -	accuracy:	0.9723 -	loss:	0.0933	- val_accuracy:	0.5550 -	val_loss:	2.5764
-	106/200	- 116	151ms /s+on	2661192611	0 0011	10001	0 0727	val accuracy.	0 5650	val locci	2 2400
25/25 Epoch	107/200	115	451ms/step -	accuracy.	0.9611 -	1055.	0.0/2/	- val_accuracy:	0.5050 -	Va1_1055.	2.3499
25/25		12s	466ms/step -	accuracy:	0.9667 -	loss:	0.0938	<pre>- val_accuracy:</pre>	0.5650 -	<pre>val_loss:</pre>	3.0872
Epoch 25/25	108/200	- 11c	136ms/stan -	accuracy.	0 9589 -	loss	0 1/1/1 .	- val_accuracy:	0 5950 -	val loss.	3 2135
	109/200	113	450m3/ 5ccp	accui acy.	0.5505	1033.	0.1-1	var_accaracy.	0.3330	vu1_1055.	3.2133
25/25		11 s	445ms/step -	accuracy:	0.9679 -	loss:	0.1062	<pre>- val_accuracy:</pre>	0.5700 -	val_loss:	2.9951
25/25	110/200	- 11s	451ms/step -	accuracv:	0.9856 -	loss:	0.0488	- val accuracy:	0.5850 -	val loss:	3.3977
	111/200			ĺ				_ ,		_	
25/25 Enach		- 11 s	436ms/step -	accuracy:	0.9446 -	loss:	0.1606	- val_accuracy:	0.5650 -	val_loss:	3.1326
25/25	112/200	- 12s	476ms/step -	accuracy:	0.9832 -	loss:	0.0424	- val_accuracy:	0.5650 -	val_loss:	2.7708
-	113/200			-						_	
25/25 Enoch	114/200	- 11s	451ms/step -	accuracy:	0.9500 -	loss:	0.1129	- val_accuracy:	0.5750 -	val_loss:	2.9067
25/25		12 s	475ms/step -	accuracy:	0.9897 -	loss:	0.0355	- val_accuracy:	0.5950 -	val_loss:	2.9419
-	115/200	- 116	116ms /ston	2661192611	0.760	10001	0 0644	val accuracy.	0 5600	val locci	2 7060
25/25 Epoch	116/200	115	440ms/step -	accuracy.	0.9709 -	1055.	0.0044	- val_accuracy:	0.5000 -	Va1_1055.	3.7900
25/25		- 11s	456ms/step -	accuracy:	0.9770 -	loss:	0.0871	<pre>- val_accuracy:</pre>	0.5700 -	val_loss:	3.1759
Epoch 25/25	117/200	12c	159mc/c+an	accuracy:	0 9854	1055.	0 0627	- val_accuracy:	0 5700	val locc	3 0052
	118/200	123		accui acy.	0.7074 -	1033.	0.002/	var_accuracy.	3.3/00 -	vu1_1055.	J. 00JZ
25/25		12 s	459ms/step -	accuracy:	0.9806 -	loss:	0.0613	<pre>- val_accuracy:</pre>	0.5750 -	<pre>val_loss:</pre>	3.2695
Epoch 25/25	119/200	- 11s	446ms/sten -	accuracy.	0.9679 -	loss	0.1012	- val_accuracy:	0.5650 -	val loss.	2.8748
	120/200	3	с э сср	accui acy i	3.50,5		3,1012	.uaccaracy.	3.3030	1033.	,
25/25		11 s	441ms/step -	accuracy:	0.9692 -	loss:	0.0690	<pre>- val_accuracy:</pre>	0.6050 -	val_loss:	2.5103
Epoch 25/25	121/200	- 11 s	440ms/step -	accuracy:	0.9825 -	loss:	0.0566	- val_accuracy:	0.6000 -	val loss:	2.9525
Epoch	122/200			-						_	
25/25 Enoch		- 21s	467ms/step -	accuracy:	0.9866 -	loss:	0.0416	- val_accuracy:	0.5750 -	val_loss:	3.5913
25/25	123/200	- 13s	506ms/step -	accuracy:	0.9780 -	loss:	0.1305	- val_accuracy:	0.5700 -	val_loss:	2.7642
			•	•				•		_	

Epoch	124/200										
25/25 Enoch	125/200	13s	500ms/step	- accuracy:	0.9718 -	loss:	0.0886 -	val_accuracy:	0.5750 -	val_loss:	3.1585
25/25		12s	458ms/step	accuracy:	0.9804 -	loss:	0.0619 -	val_accuracy:	0.5650 -	val_loss:	2.9625
25/25	126/200	12s	461ms/step	- accuracy:	0.9762 -	loss:	0.0649 -	· val_accuracy:	0.5750 -	val_loss:	3.0198
Epoch 25/25	127/200	- 11s	448ms/step	- accuracv:	0.9707 -	· loss:	0.1140 -	· val_accuracy:	0.5750 -	· val loss:	2.7649
	128/200			-				- val_accuracy:		_	
Epoch	129/200		•	-				_ ,		_	
25/25 Epoch	130/200	· 11s	441ms/step	- accuracy:	0.9855 -	loss:	0.0462 -	· val_accuracy:	0.5600 -	· val_loss:	3.7263
25/25 Epoch	131/200	11s	431ms/step	accuracy:	0.9819 -	loss:	0.0629 -	<pre>val_accuracy:</pre>	0.5700 -	val_loss:	3.2173
25/25		11s	431ms/step	- accuracy:	0.9690 -	loss:	0.0907 -	val_accuracy:	0.5850 -	val_loss:	3.3084
25/25		11s	444ms/step	- accuracy:	0.9616 -	loss:	0.0969 -	val_accuracy:	0.5950 -	val_loss:	2.6866
Epoch 25/25	133/200	12s	462ms/step	- accuracy:	0.9888 -	loss:	0.0369 -	· val_accuracy:	0.5950 -	val_loss:	2.8829
Epoch 25/25	134/200	- 11s	447ms/step	- accuracy:	0.9853 -	loss:	0.0488 -	· val_accuracy:	0.5750 -	val loss:	2.7760
Epoch 25/25	135/200		•	-				- val accuracy:		_	
Epoch	136/200		•					_ ,		_	
25/25 Epoch	137/200			-				· val_accuracy:		_	
25/25 Epoch	138/200	12s	484ms/step	- accuracy:	0.9857 -	loss:	0.0469 -	· val_accuracy:	0.5500 -	val_loss:	3.1712
25/25 Enoch	139/200	12s	481ms/step	- accuracy:	0.9869 -	loss:	0.0471 -	val_accuracy:	0.5950 -	val_loss:	2.9931
25/25		12s	495ms/step	- accuracy:	0.9815 -	loss:	0.0554 -	val_accuracy:	0.5850 -	val_loss:	2.8937
25/25		12s	484ms/step	- accuracy:	0.9905 -	loss:	0.0404 -	val_accuracy:	0.5650 -	val_loss:	3.5348
Epoch 25/25	141/200	12s	474ms/step	- accuracy:	0.9711 -	loss:	0.1107 -	· val_accuracy:	0.5450 -	val_loss:	2.9330
Epoch 25/25	142/200	- 11s	456ms/step	- accuracy:	0.9827 -	loss:	0.0638 -	<pre>- val_accuracy:</pre>	0.5800 -	val_loss:	2.7175
Epoch 25/25	143/200	115	453ms/sten	- accuracy:	0.9693 -	· loss:	0.1044 -	- val_accuracy:	0.5900 -	· val loss:	2.5806
Epoch	144/200		•	-				_ ,		_	
-	145/200		•	-				· val_accuracy:		_	
25/25 Epoch	146/200	12s	459ms/step	- accuracy:	0.9614 -	· loss:	0.1111 -	· val_accuracy:	0.5550 -	· val_loss:	3.4666
25/25 Epoch	147/200	11s	456ms/step	- accuracy:	0.9848 -	loss:	0.0526 -	· val_accuracy:	0.5650 -	val_loss:	3.5513
25/25 Enoch	148/200	12s	474ms/step	- accuracy:	0.9844 -	loss:	0.0788 -	val_accuracy:	0.5750 -	val_loss:	2.8310
25/25		11s	441ms/step	- accuracy:	0.9880 -	loss:	0.0294 -	val_accuracy:	0.6150 -	val_loss:	3.0327
25/25		12s	463ms/step	- accuracy:	0.9798 -	loss:	0.0653 -	· val_accuracy:	0.6100 -	val_loss:	2.5929
Epoch 25/25	150/200	12s	465ms/step	- accuracy:	0.9912 -	loss:	0.0427 -	· val_accuracy:	0.5750 -	val_loss:	3.6527
Epoch 25/25	151/200	12s	464ms/step	- accuracy:	0.9785 -	loss:	0.1017 -	· val_accuracy:	0.5600 -	val loss:	3.2230
Epoch 25/25	152/200			-				- val_accuracy:		_	
Epoch	153/200			-				_ ,		_	
-	154/200			-				· val_accuracy:		_	
25/25 Epoch	155/200	· 12s	486ms/step	- accuracy:	0.9900 -	· loss:	0.0370 -	· val_accuracy:	0.5850 -	· val_loss:	3.6133
25/25 Epoch	156/200	12s	475ms/step	- accuracy:	0.9752 -	loss:	0.0612 -	· val_accuracy:	0.5850 -	val_loss:	3.4285
25/25 Epoch	157/200	12s	476ms/step	accuracy:	0.9927 -	loss:	0.0228 -	· val_accuracy:	0.5850 -	val_loss:	3.9586
25/25		12s	487ms/step	- accuracy:	0.9683 -	loss:	0.1098 -	val_accuracy:	0.5600 -	val_loss:	3.3739
25/25		11s	453ms/step	accuracy:	0.9795 -	loss:	0.0823 -	val_accuracy:	0.5700 -	val_loss:	3.5236
25/25	159/200	12s	470ms/step	- accuracy:	0.9853 -	loss:	0.0560 -	· val_accuracy:	0.5550 -	val_loss:	3.9693
Epoch 25/25	160/200	13s	507ms/step	- accuracy:	0.9642 -	loss:	0.1503 -	· val_accuracy:	0.5900 -	val_loss:	3.8540
Epoch 25/25	161/200			-				- val_accuracy:		_	
	162/200			-				<pre>val_accuracy:</pre>		_	
Epoch	163/200			-				_ ,		_	
25/25 Epoch	164/200			-				· val_accuracy:		_	
25/25		12s	472ms/step	- accuracy:	0.9842 -	loss:	0.0422 -	<pre>val_accuracy:</pre>	0.5850 -	val_loss:	3.5753

```
Epoch 165/200
                                 13s 509ms/step - accuracy: 0.9946 - loss: 0.0276 - val_accuracy: 0.5800 - val_loss: 4.1496
       25/25
       Epoch 166/200
       25/25
                                 · 12s 475ms/step - accuracy: 0.9832 - loss: 0.0464 - val_accuracy: 0.5450 - val_loss: 3.8821
       Epoch 167/200
      25/25
                                 13s 500ms/step - accuracy: 0.9846 - loss: 0.0371 - val_accuracy: 0.5700 - val_loss: 4.1487
       Epoch 168/200
                                 12s 498ms/step - accuracy: 0.9689 - loss: 0.1025 - val_accuracy: 0.5750 - val_loss: 3.4159
      25/25
       Epoch 169/200
      25/25
                                 - 12s 481ms/step - accuracy: 0.9792 - loss: 0.0696 - val_accuracy: 0.5800 - val_loss: 3.8318
      Epoch 170/200
      25/25
                                 - 12s 474ms/step - accuracy: 0.9868 - loss: 0.0269 - val_accuracy: 0.5700 - val_loss: 3.6744
       Epoch 171/200
      25/25
                                 20s 456ms/step - accuracy: 0.9849 - loss: 0.0440 - val_accuracy: 0.5550 - val_loss: 3.8352
       Epoch 172/200
       25/25
                                 12s 463ms/step - accuracy: 0.9948 - loss: 0.0249 - val_accuracy: 0.5500 - val_loss: 4.1566
       Epoch 173/200
                                 13s 528ms/step - accuracy: 0.9784 - loss: 0.0598 - val_accuracy: 0.5800 - val_loss: 4.4257
      25/25
      Epoch 174/200
       25/25
                                 · 20s 489ms/step - accuracy: 0.9742 - loss: 0.0952 - val_accuracy: 0.5450 - val_loss: 4.0259
      Epoch 175/200
       25/25
                                 12s 472ms/step - accuracy: 0.9933 - loss: 0.0179 - val_accuracy: 0.5750 - val_loss: 3.7908
       Epoch 176/200
      25/25
                                 12s 475ms/step - accuracy: 0.9774 - loss: 0.1114 - val_accuracy: 0.5750 - val_loss: 3.6599
       Epoch 177/200
                                 · 12s 466ms/step - accuracy: 0.9853 - loss: 0.0346 - val_accuracy: 0.5600 - val_loss: 4.2077
       25/25
       Epoch 178/200
      25/25
                                 · 12s 475ms/step - accuracy: 0.9845 - loss: 0.0378 - val_accuracy: 0.5700 - val_loss: 4.1975
      Epoch 179/200
                                 12s 477ms/step - accuracy: 0.9792 - loss: 0.0861 - val_accuracy: 0.5750 - val_loss: 4.6767
       25/25
      Epoch 180/200
      25/25
                                - 12s 481ms/step - accuracy: 0.9858 - loss: 0.0325 - val_accuracy: 0.5700 - val_loss: 4.3123
       Epoch 181/200
      25/25
                                 - 12s 484ms/step - accuracy: 0.9795 - loss: 0.0905 - val_accuracy: 0.5850 - val_loss: 4.3368
       Epoch 182/200
      25/25
                                 - 11s 446ms/step - accuracy: 0.9910 - loss: 0.0388 - val_accuracy: 0.6150 - val_loss: 4.0462
       Epoch 183/200
       25/25
                                 11s 452ms/step - accuracy: 0.9808 - loss: 0.0731 - val_accuracy: 0.6000 - val_loss: 3.6067
       Epoch 184/200
      25/25
                                 · 12s 465ms/step - accuracy: 0.9906 - loss: 0.0207 - val_accuracy: 0.5900 - val_loss: 4.4340
       Epoch 185/200
       25/25
                                 - 11s 453ms/step - accuracy: 0.9866 - loss: 0.0524 - val_accuracy: 0.5950 - val_loss: 4.0456
       Epoch 186/200
      25/25
                                 11s 435ms/step - accuracy: 0.9843 - loss: 0.0646 - val_accuracy: 0.6100 - val_loss: 3.4074
       Epoch 187/200
       25/25 -
                                 11s 434ms/step - accuracy: 0.9754 - loss: 0.0781 - val_accuracy: 0.5750 - val_loss: 3.6424
       Epoch 188/200
      25/25
                                 - 11s 435ms/step - accuracy: 0.9921 - loss: 0.0286 - val_accuracy: 0.5900 - val_loss: 3.4255
       Epoch 189/200
                                 - 11s 458ms/step - accuracy: 0.9945 - loss: 0.0156 - val_accuracy: 0.6150 - val_loss: 3.5043
      25/25
      Epoch 190/200
      25/25
                                 - 11s 457ms/step - accuracy: 0.9765 - loss: 0.0669 - val_accuracy: 0.5900 - val_loss: 3.0471
      Epoch 191/200
       25/25
                                 11s 441ms/step - accuracy: 0.9822 - loss: 0.0497 - val_accuracy: 0.5900 - val_loss: 3.5785
       Epoch 192/200
      25/25
                                 11s 432ms/step - accuracy: 0.9866 - loss: 0.0376 - val_accuracy: 0.6000 - val_loss: 3.5430
       Epoch 193/200
       25/25
                                 - 11s 447ms/step - accuracy: 0.9905 - loss: 0.0331 - val_accuracy: 0.5700 - val_loss: 3.9939
       Epoch 194/200
                                 21s 459ms/step - accuracy: 0.9802 - loss: 0.0612 - val_accuracy: 0.5750 - val_loss: 4.2690
      25/25
      Epoch 195/200
       25/25
                                 12s 458ms/step - accuracy: 0.9828 - loss: 0.0629 - val_accuracy: 0.5700 - val_loss: 3.7037
      Epoch 196/200
      25/25
                                 - 11s 457ms/step - accuracy: 0.9917 - loss: 0.0292 - val_accuracy: 0.6000 - val_loss: 3.6349
       Epoch 197/200
      25/25
                                 11s 457ms/step - accuracy: 0.9881 - loss: 0.0378 - val_accuracy: 0.5800 - val_loss: 3.7407
       Epoch 198/200
       25/25
                                 12s 460ms/step - accuracy: 0.9869 - loss: 0.0346 - val_accuracy: 0.5800 - val_loss: 3.9399
       Epoch 199/200
       25/25
                                 12s 461ms/step - accuracy: 0.9959 - loss: 0.0230 - val_accuracy: 0.5700 - val_loss: 4.6127
       Epoch 200/200
       25/25
                                 12s 467ms/step - accuracy: 0.9954 - loss: 0.0106 - val_accuracy: 0.5800 - val_loss: 4.2340
In [8]: print(f"Execution time: {elapsed_time:.2f} seconds")
```

Execution time: 2385.14 seconds

```
In [9]: def append_core_data(score_path, num_cores, elapsed_time):
    # Check if the file already exists
    file_exists = os.path.exists(score_path)

# Open the file in append mode
with open(score_path, mode='a', newline='') as file:
```

```
writer = csv.writer(file)

# If the file is new, write the header
if not file_exists:
    writer.writerow(["Number of Cores", "Elapsed Time"])

# Write the new data
writer.writerow([num_cores, elapsed_time])
```

```
In [1]: %matplotlib inline
        import shutil
        import random
        import numpy as np
        from warnings import filterwarnings
        filterwarnings('ignore')
        from tensorflow.keras import layers, regularizers, optimizers
        from tensorflow.keras import models
        from tensorflow.keras.models import Sequential, Model
        from tensorflow.keras.layers import LeakyReLU,Dense, Activation, Flatten, Dropout, BatchNormalization,Conv2D, MaxPooling2D
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import tensorflow as tf
        import os
        import time
        import csv
        from matplotlib import figure
```

Define 4 worker

```
In [2]: # Set the number of threads
number_of_worker = 4
os.environ['OMP_NUM_THREADS'] = '4' # OpenMP threads for parallelism
os.environ['TF_NUM_INTEROP_THREADS'] = '4' # Threads for inter-operation parallelism
os.environ['TF_NUM_INTRAOP_THREADS'] = '4' # Threads for intra-operation parallelism
# Confirm TensorFlow is using the specified number of threads
tf.config.threading.set_inter_op_parallelism_threads(number_of_worker)
tf.config.threading.set_intra_op_parallelism_threads(number_of_worker)
```

```
In [3]: source_dir = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DATA\Convert_Audio_File_to_jpg_file"
        target_dir = r'genres_train_val_split_data'
        split_ratio = 0.8
        def Train_Test_Split(source_dir,target_dir,split_ratio):
            # Define source and target directories
            train_dir = os.path.join(target_dir, 'train')
            val_dir = os.path.join(target_dir, 'val')
            # Create target directories if they don't exist
            os.makedirs(train_dir, exist_ok=True)
            os.makedirs(val_dir, exist_ok=True)
            # Get the list of class directories
            classes = [d for d in os.listdir(source_dir) if os.path.isdir(os.path.join(source_dir, d))]
            for class name in classes:
                # Create class directories in train and val folders
                os.makedirs(os.path.join(train_dir, class_name), exist_ok=True)
                os.makedirs(os.path.join(val_dir, class_name), exist_ok=True)
                # Get list of images in the class directory
                class_dir = os.path.join(source_dir, class_name)
                images = [f for f in os.listdir(class_dir) if os.path.isfile(os.path.join(class_dir, f))]
                # Shuffle the images
                random.shuffle(images)
                # Compute the split point
                split_point = int(len(images) * split_ratio)
                # Split the images into training and validation sets
                train images = images[:split point]
                val_images = images[split_point:]
```

Data split completed successfully!

Load the Data

```
In [5]: WIDTH = 64
        HEIGHT = 64
        BATCH_SIZE = 32
        TRAIN_DIR=r'genres_train_val_split_data/train'
        val_dir = r'genres_train_val_split_data/val'
        # data prep
        train_datagen = ImageDataGenerator(
            rescale=1./255.,validation_split=0.25)
        train_generator = train_datagen.flow_from_directory(
            TRAIN_DIR,
            target size=(HEIGHT, WIDTH),
                batch_size=BATCH_SIZE,
                class_mode='categorical')
        validation_gen = train_datagen.flow_from_directory(
            val_dir,target_size = (HEIGHT,WIDTH),
            batch_size = BATCH_SIZE,
            class_mode = 'categorical'
```

Found 800 images belonging to 10 classes. Found 200 images belonging to 10 classes.

Model Architecture

```
In [6]: model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                         input_shape=(64,64,3)))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(128, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(128, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(10, activation='softmax'))
        model.compile(optimizers.RMSprop(learning_rate=0.0005, decay=1e-6),loss="categorical_crossentropy",metrics=["accuracy"])
        model.summary()
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 64, 64, 32)	896
activation (Activation)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 62, 62, 64)	18,496
activation_1 (Activation)	(None, 62, 62, 64)	0
max_pooling2d (MaxPooling2D)	(None, 31, 31, 64)	0
dropout (Dropout)	(None, 31, 31, 64)	0
conv2d_2 (Conv2D)	(None, 31, 31, 64)	36,928
activation_2 (Activation)	(None, 31, 31, 64)	0
conv2d_3 (Conv2D)	(None, 29, 29, 64)	36,928
activation_3 (Activation)	(None, 29, 29, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_1 (Dropout)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 14, 14, 128)	73,856
activation_4 (Activation)	(None, 14, 14, 128)	0
conv2d_5 (Conv2D)	(None, 12, 12, 128)	147,584
activation_5 (Activation)	(None, 12, 12, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_2 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2,359,808
activation_6 (Activation)	(None, 512)	0
dropout_3 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5,130

```
In [7]: STEP_SIZE_TRAIN=train_generator.n//train_generator.batch_size
# Measure the execution time
start_time = time.time()

model.fit(train_generator,validation_data=validation_gen,epochs=200)

end_time = time.time()
elapsed_time = end_time - start_time
```

```
Epoch 1/200
                           5s 173ms/step - accuracy: 0.0870 - loss: 2.3293 - val_accuracy: 0.1100 - val_loss: 2.3020
25/25
Epoch 2/200
25/25 -
                           5s 216ms/step - accuracy: 0.0947 - loss: 2.2986 - val_accuracy: 0.1950 - val_loss: 2.2405
Epoch 3/200
25/25
                           6s 225ms/step - accuracy: 0.1855 - loss: 2.2011 - val_accuracy: 0.2150 - val_loss: 2.0685
Epoch 4/200
                           5s 208ms/step - accuracy: 0.2282 - loss: 2.0942 - val_accuracy: 0.2550 - val_loss: 2.0424
25/25
Epoch 5/200
25/25
                          • 5s 215ms/step - accuracy: 0.2203 - loss: 2.0702 - val_accuracy: 0.2550 - val_loss: 2.0806
Epoch 6/200
25/25
                          5s 217ms/step - accuracy: 0.2572 - loss: 1.9645 - val_accuracy: 0.2900 - val_loss: 1.9782
Epoch 7/200
25/25
                           5s 209ms/step - accuracy: 0.3156 - loss: 1.9579 - val_accuracy: 0.3250 - val_loss: 1.8348
Epoch 8/200
25/25
                          5s 213ms/step - accuracy: 0.3074 - loss: 1.8452 - val_accuracy: 0.3250 - val_loss: 1.8351
Epoch 9/200
                           6s 226ms/step - accuracy: 0.3443 - loss: 1.8782 - val_accuracy: 0.2900 - val_loss: 1.8884
25/25
Epoch 10/200
25/25
                           5s 212ms/step - accuracy: 0.3463 - loss: 1.8369 - val_accuracy: 0.3250 - val_loss: 1.8362
Epoch 11/200
25/25
                           5s 217ms/step - accuracy: 0.3712 - loss: 1.7110 - val_accuracy: 0.3650 - val_loss: 1.7915
Epoch 12/200
25/25
                           5s 220ms/step - accuracy: 0.3908 - loss: 1.6327 - val_accuracy: 0.3700 - val_loss: 1.6816
Epoch 13/200
25/25
                           5s 211ms/step - accuracy: 0.4393 - loss: 1.5735 - val_accuracy: 0.3700 - val_loss: 1.5917
Epoch 14/200
25/25
                           5s 204ms/step - accuracy: 0.4326 - loss: 1.5463 - val_accuracy: 0.4000 - val_loss: 1.5241
Epoch 15/200
                           5s 217ms/step - accuracy: 0.4399 - loss: 1.5921 - val_accuracy: 0.4650 - val_loss: 1.4620
25/25
Epoch 16/200
25/25
                          - 5s 220ms/step - accuracy: 0.4507 - loss: 1.4386 - val_accuracy: 0.4450 - val_loss: 1.4631
Epoch 17/200
25/25
                           5s 215ms/step - accuracy: 0.4912 - loss: 1.3967 - val_accuracy: 0.4400 - val_loss: 1.4481
Epoch 18/200
25/25
                           5s 217ms/step - accuracy: 0.5001 - loss: 1.3790 - val_accuracy: 0.3750 - val_loss: 2.0581
Epoch 19/200
25/25
                           5s 205ms/step - accuracy: 0.5079 - loss: 1.4559 - val_accuracy: 0.5000 - val_loss: 1.3329
Epoch 20/200
25/25
                           5s 196ms/step - accuracy: 0.5149 - loss: 1.3141 - val_accuracy: 0.5050 - val_loss: 1.3870
Epoch 21/200
25/25
                           5s 213ms/step - accuracy: 0.5731 - loss: 1.1841 - val_accuracy: 0.4200 - val_loss: 1.6203
Epoch 22/200
25/25
                           5s 206ms/step - accuracy: 0.5305 - loss: 1.2805 - val_accuracy: 0.4950 - val_loss: 1.3256
Epoch 23/200
25/25
                           5s 203ms/step - accuracy: 0.5929 - loss: 1.1132 - val_accuracy: 0.5200 - val_loss: 1.3397
Epoch 24/200
25/25
                          5s 210ms/step - accuracy: 0.5806 - loss: 1.2093 - val_accuracy: 0.4350 - val_loss: 1.6354
Epoch 25/200
                           5s 213ms/step - accuracy: 0.5817 - loss: 1.1601 - val_accuracy: 0.4750 - val_loss: 1.3573
25/25
Epoch 26/200
25/25
                          5s 196ms/step - accuracy: 0.5897 - loss: 1.1286 - val_accuracy: 0.5000 - val_loss: 1.2507
Epoch 27/200
25/25
                           5s 199ms/step - accuracy: 0.6270 - loss: 1.0092 - val_accuracy: 0.4900 - val_loss: 1.2686
Epoch 28/200
25/25
                           5s 208ms/step - accuracy: 0.6185 - loss: 1.0675 - val_accuracy: 0.4900 - val_loss: 1.2638
Epoch 29/200
25/25 -
                           5s 196ms/step - accuracy: 0.6287 - loss: 1.0181 - val_accuracy: 0.5050 - val_loss: 1.2547
Epoch 30/200
25/25
                           5s 215ms/step - accuracy: 0.6260 - loss: 0.9962 - val_accuracy: 0.5350 - val_loss: 1.1956
Epoch 31/200
25/25
                           5s 207ms/step - accuracy: 0.6691 - loss: 0.9625 - val_accuracy: 0.5700 - val_loss: 1.2107
Epoch 32/200
25/25
                           5s 197ms/step - accuracy: 0.6510 - loss: 0.9998 - val accuracy: 0.5350 - val loss: 1.2706
Epoch 33/200
25/25
                           5s 199ms/step - accuracy: 0.6723 - loss: 0.9011 - val_accuracy: 0.5750 - val_loss: 1.2675
Epoch 34/200
25/25
                           5s 202ms/step - accuracy: 0.6595 - loss: 0.9651 - val_accuracy: 0.5950 - val_loss: 1.1681
Epoch 35/200
25/25
                           5s 199ms/step - accuracy: 0.6944 - loss: 0.9113 - val_accuracy: 0.5800 - val_loss: 1.2224
Epoch 36/200
25/25
                           5s 205ms/step - accuracy: 0.7236 - loss: 0.7811 - val accuracy: 0.5850 - val loss: 1.2078
Epoch 37/200
25/25
                          - 5s 202ms/step - accuracy: 0.7506 - loss: 0.7508 - val_accuracy: 0.5300 - val_loss: 1.3483
Epoch 38/200
25/25
                           5s 199ms/step - accuracy: 0.6970 - loss: 0.7597 - val_accuracy: 0.5600 - val_loss: 1.2773
Epoch 39/200
25/25
                           5s 209ms/step - accuracy: 0.7253 - loss: 0.7940 - val_accuracy: 0.5750 - val_loss: 1.3194
Epoch 40/200
25/25
                           5s 213ms/step - accuracy: 0.7113 - loss: 0.7393 - val_accuracy: 0.6100 - val_loss: 1.2596
Epoch 41/200
25/25
                           5s 191ms/step - accuracy: 0.7447 - loss: 0.7134 - val_accuracy: 0.5700 - val_loss: 1.3982
```

```
Epoch 42/200
                           5s 209ms/step - accuracy: 0.7806 - loss: 0.6490 - val_accuracy: 0.5650 - val_loss: 1.2638
25/25
Epoch 43/200
25/25
                           6s 223ms/step - accuracy: 0.7568 - loss: 0.6466 - val_accuracy: 0.5850 - val_loss: 1.2449
Epoch 44/200
25/25
                           5s 207ms/step - accuracy: 0.7851 - loss: 0.5774 - val_accuracy: 0.5750 - val_loss: 1.3505
Epoch 45/200
                           5s 213ms/step - accuracy: 0.7652 - loss: 0.6773 - val_accuracy: 0.6100 - val_loss: 1.4523
25/25
Epoch 46/200
25/25
                          • 5s 204ms/step - accuracy: 0.7777 - loss: 0.5810 - val_accuracy: 0.5950 - val_loss: 1.4568
Epoch 47/200
25/25
                          5s 205ms/step - accuracy: 0.7944 - loss: 0.5911 - val_accuracy: 0.6000 - val_loss: 1.4199
Epoch 48/200
                           5s 202ms/step - accuracy: 0.8184 - loss: 0.5705 - val_accuracy: 0.5950 - val_loss: 1.4033
25/25
Epoch 49/200
25/25
                          5s 208ms/step - accuracy: 0.8032 - loss: 0.4969 - val_accuracy: 0.5350 - val_loss: 1.5725
Epoch 50/200
                           5s 203ms/step - accuracy: 0.8524 - loss: 0.4568 - val_accuracy: 0.5650 - val_loss: 1.5162
25/25
Epoch 51/200
25/25
                           5s 210ms/step - accuracy: 0.8675 - loss: 0.4323 - val_accuracy: 0.5550 - val_loss: 1.6451
Epoch 52/200
25/25
                           5s 204ms/step - accuracy: 0.8801 - loss: 0.3655 - val_accuracy: 0.6250 - val_loss: 1.4862
Epoch 53/200
25/25
                           5s 199ms/step - accuracy: 0.8359 - loss: 0.4234 - val_accuracy: 0.6050 - val_loss: 1.5294
Epoch 54/200
25/25
                           5s 215ms/step - accuracy: 0.8761 - loss: 0.3410 - val_accuracy: 0.5500 - val_loss: 1.8546
Epoch 55/200
25/25
                           5s 210ms/step - accuracy: 0.8498 - loss: 0.4090 - val_accuracy: 0.6050 - val_loss: 1.5737
Epoch 56/200
                           5s 216ms/step - accuracy: 0.8667 - loss: 0.3876 - val_accuracy: 0.5550 - val_loss: 2.0229
25/25
Epoch 57/200
25/25
                          - 5s 201ms/step - accuracy: 0.8925 - loss: 0.3106 - val_accuracy: 0.6200 - val_loss: 1.6577
Epoch 58/200
25/25
                           5s 207ms/step - accuracy: 0.8861 - loss: 0.3055 - val_accuracy: 0.6050 - val_loss: 1.7928
Epoch 59/200
25/25
                           5s 199ms/step - accuracy: 0.9239 - loss: 0.2821 - val_accuracy: 0.6300 - val_loss: 1.6734
Epoch 60/200
25/25
                           5s 211ms/step - accuracy: 0.9099 - loss: 0.2721 - val_accuracy: 0.6400 - val_loss: 1.8624
Epoch 61/200
25/25
                           5s 209ms/step - accuracy: 0.8851 - loss: 0.2915 - val_accuracy: 0.6300 - val_loss: 1.7194
Epoch 62/200
25/25
                           5s 205ms/step - accuracy: 0.9221 - loss: 0.2401 - val_accuracy: 0.5750 - val_loss: 2.4121
Epoch 63/200
25/25
                           5s 211ms/step - accuracy: 0.8771 - loss: 0.4255 - val_accuracy: 0.6400 - val_loss: 1.6150
Epoch 64/200
25/25
                           5s 208ms/step - accuracy: 0.9291 - loss: 0.2064 - val_accuracy: 0.6100 - val_loss: 2.1086
Epoch 65/200
25/25
                          5s 203ms/step - accuracy: 0.8845 - loss: 0.2733 - val accuracy: 0.5550 - val loss: 1.9794
Epoch 66/200
                           5s 208ms/step - accuracy: 0.9221 - loss: 0.2269 - val_accuracy: 0.6400 - val_loss: 1.6637
25/25
Epoch 67/200
25/25
                          • 5s 212ms/step - accuracy: 0.9380 - loss: 0.2449 - val_accuracy: 0.5950 - val_loss: 1.9453
Epoch 68/200
25/25
                           5s 196ms/step - accuracy: 0.9244 - loss: 0.1791 - val_accuracy: 0.6150 - val_loss: 1.8998
Epoch 69/200
25/25
                           5s 218ms/step - accuracy: 0.9393 - loss: 0.1946 - val_accuracy: 0.6400 - val_loss: 2.0027
Epoch 70/200
25/25 -
                           5s 212ms/step - accuracy: 0.9358 - loss: 0.1815 - val_accuracy: 0.6250 - val_loss: 2.1165
Epoch 71/200
25/25
                           5s 197ms/step - accuracy: 0.8922 - loss: 0.2702 - val_accuracy: 0.6250 - val_loss: 1.6935
Epoch 72/200
25/25
                           5s 211ms/step - accuracy: 0.9485 - loss: 0.1672 - val_accuracy: 0.6050 - val_loss: 1.8513
Epoch 73/200
25/25
                           5s 202ms/step - accuracy: 0.9588 - loss: 0.1375 - val accuracy: 0.6150 - val loss: 2.1926
Epoch 74/200
25/25
                           5s 204ms/step - accuracy: 0.9385 - loss: 0.1757 - val_accuracy: 0.6100 - val_loss: 2.0779
Epoch 75/200
                           5s 203ms/step - accuracy: 0.9415 - loss: 0.1850 - val_accuracy: 0.6000 - val_loss: 2.5533
25/25
Epoch 76/200
25/25
                           5s 202ms/step - accuracy: 0.9361 - loss: 0.1930 - val_accuracy: 0.6200 - val_loss: 2.0268
Epoch 77/200
25/25
                           5s 202ms/step - accuracy: 0.9443 - loss: 0.1485 - val accuracy: 0.6150 - val loss: 1.9755
Epoch 78/200
25/25
                          - 5s 203ms/step - accuracy: 0.9376 - loss: 0.1604 - val_accuracy: 0.6000 - val_loss: 2.1409
Epoch 79/200
25/25
                           5s 203ms/step - accuracy: 0.9639 - loss: 0.1241 - val_accuracy: 0.6350 - val_loss: 1.8937
Epoch 80/200
25/25
                           5s 199ms/step - accuracy: 0.9295 - loss: 0.1915 - val_accuracy: 0.6250 - val_loss: 2.3712
Epoch 81/200
25/25
                           5s 209ms/step - accuracy: 0.9673 - loss: 0.1002 - val_accuracy: 0.6200 - val_loss: 2.1481
Epoch 82/200
25/25
                           5s 215ms/step - accuracy: 0.9725 - loss: 0.1002 - val_accuracy: 0.6100 - val_loss: 2.2447
```

```
Epoch 83/200
                           5s 197ms/step - accuracy: 0.9341 - loss: 0.1945 - val_accuracy: 0.6350 - val_loss: 2.1486
25/25
Epoch 84/200
25/25
                          5s 210ms/step - accuracy: 0.9491 - loss: 0.1159 - val_accuracy: 0.6250 - val_loss: 2.4833
Epoch 85/200
25/25
                           5s 202ms/step - accuracy: 0.9479 - loss: 0.1705 - val_accuracy: 0.6550 - val_loss: 2.0087
Epoch 86/200
                          5s 207ms/step - accuracy: 0.9279 - loss: 0.2786 - val_accuracy: 0.6250 - val_loss: 1.9962
25/25
Epoch 87/200
25/25
                          5s 201ms/step - accuracy: 0.9769 - loss: 0.0652 - val_accuracy: 0.6300 - val_loss: 2.4628
Epoch 88/200
25/25
                          5s 199ms/step - accuracy: 0.9497 - loss: 0.1454 - val_accuracy: 0.6500 - val_loss: 2.3110
Epoch 89/200
                           5s 199ms/step - accuracy: 0.9618 - loss: 0.1068 - val_accuracy: 0.6050 - val_loss: 2.2710
25/25
Epoch 90/200
25/25
                          5s 205ms/step - accuracy: 0.9721 - loss: 0.0832 - val_accuracy: 0.5950 - val_loss: 2.7386
Epoch 91/200
                           5s 204ms/step - accuracy: 0.9349 - loss: 0.1676 - val_accuracy: 0.6250 - val_loss: 2.4609
25/25
Epoch 92/200
25/25
                          5s 205ms/step - accuracy: 0.9738 - loss: 0.0941 - val_accuracy: 0.6350 - val_loss: 2.3282
Epoch 93/200
                          5s 207ms/step - accuracy: 0.9720 - loss: 0.0798 - val_accuracy: 0.6100 - val_loss: 2.5202
25/25
Epoch 94/200
25/25
                          5s 213ms/step - accuracy: 0.9720 - loss: 0.0775 - val_accuracy: 0.6000 - val_loss: 2.6692
Epoch 95/200
25/25
                          5s 204ms/step - accuracy: 0.9602 - loss: 0.1247 - val_accuracy: 0.6050 - val_loss: 2.3386
Epoch 96/200
25/25
                           5s 211ms/step - accuracy: 0.9693 - loss: 0.1137 - val_accuracy: 0.5850 - val_loss: 2.4407
Epoch 97/200
                          5s 213ms/step - accuracy: 0.9746 - loss: 0.0910 - val_accuracy: 0.6400 - val_loss: 2.2343
25/25
Epoch 98/200
25/25
                          - 5s 201ms/step - accuracy: 0.9675 - loss: 0.0999 - val_accuracy: 0.6050 - val_loss: 2.4878
Epoch 99/200
25/25
                          5s 209ms/step - accuracy: 0.9729 - loss: 0.1077 - val_accuracy: 0.6300 - val_loss: 2.3810
Epoch 100/200
25/25
                          5s 207ms/step - accuracy: 0.9738 - loss: 0.0803 - val_accuracy: 0.6200 - val_loss: 2.4213
Epoch 101/200
25/25
                          5s 208ms/step - accuracy: 0.9728 - loss: 0.0843 - val_accuracy: 0.6150 - val_loss: 2.6465
Epoch 102/200
25/25
                          5s 206ms/step - accuracy: 0.9731 - loss: 0.0868 - val_accuracy: 0.6150 - val_loss: 2.2760
Epoch 103/200
25/25
                          5s 208ms/step - accuracy: 0.9727 - loss: 0.0999 - val_accuracy: 0.6450 - val_loss: 2.4515
Epoch 104/200
25/25
                          5s 202ms/step - accuracy: 0.9519 - loss: 0.1494 - val_accuracy: 0.6300 - val_loss: 2.3754
Epoch 105/200
25/25
                           5s 209ms/step - accuracy: 0.9746 - loss: 0.0898 - val_accuracy: 0.6300 - val_loss: 2.7129
Epoch 106/200
25/25
                          5s 196ms/step - accuracy: 0.9821 - loss: 0.0691 - val_accuracy: 0.6300 - val_loss: 2.8243
Epoch 107/200
                          5s 213ms/step - accuracy: 0.9801 - loss: 0.0752 - val_accuracy: 0.6200 - val_loss: 2.7709
25/25
Epoch 108/200
25/25
                          • 5s 216ms/step - accuracy: 0.9664 - loss: 0.1235 - val_accuracy: 0.6100 - val_loss: 2.7197
Epoch 109/200
25/25
                          5s 206ms/step - accuracy: 0.9812 - loss: 0.0641 - val_accuracy: 0.6400 - val_loss: 2.7450
Epoch 110/200
25/25
                          5s 200ms/step - accuracy: 0.9763 - loss: 0.0770 - val_accuracy: 0.6200 - val_loss: 2.6184
Epoch 111/200
25/25
                           5s 207ms/step - accuracy: 0.9752 - loss: 0.0604 - val_accuracy: 0.6100 - val_loss: 2.6203
Epoch 112/200
25/25
                          5s 204ms/step - accuracy: 0.9920 - loss: 0.0344 - val_accuracy: 0.6200 - val_loss: 2.7276
Epoch 113/200
25/25
                          5s 213ms/step - accuracy: 0.9461 - loss: 0.1494 - val_accuracy: 0.6650 - val_loss: 2.4975
Epoch 114/200
25/25
                          5s 212ms/step - accuracy: 0.9848 - loss: 0.0547 - val accuracy: 0.6350 - val loss: 2.7020
Epoch 115/200
25/25
                          5s 210ms/step - accuracy: 0.9735 - loss: 0.0877 - val_accuracy: 0.6250 - val_loss: 2.7517
Epoch 116/200
                          5s 211ms/step - accuracy: 0.9790 - loss: 0.0701 - val_accuracy: 0.6350 - val_loss: 2.5372
25/25
Epoch 117/200
25/25
                          5s 208ms/step - accuracy: 0.9793 - loss: 0.0651 - val_accuracy: 0.6350 - val_loss: 3.0050
Epoch 118/200
25/25
                          5s 207ms/step - accuracy: 0.9703 - loss: 0.0804 - val accuracy: 0.6500 - val loss: 2.2230
Epoch 119/200
25/25
                          - 5s 196ms/step - accuracy: 0.9836 - loss: 0.0570 - val_accuracy: 0.6300 - val_loss: 2.7662
Epoch 120/200
25/25
                          5s 216ms/step - accuracy: 0.9821 - loss: 0.0468 - val_accuracy: 0.6450 - val_loss: 2.6975
Epoch 121/200
25/25
                           5s 207ms/step - accuracy: 0.9702 - loss: 0.0829 - val_accuracy: 0.6350 - val_loss: 2.5993
Epoch 122/200
25/25
                          5s 207ms/step - accuracy: 0.9644 - loss: 0.1291 - val_accuracy: 0.6450 - val_loss: 2.7768
Epoch 123/200
25/25
                           5s 210ms/step - accuracy: 0.9766 - loss: 0.0689 - val_accuracy: 0.6200 - val_loss: 2.8439
```

```
Epoch 124/200
                           5s 210ms/step - accuracy: 0.9886 - loss: 0.0279 - val_accuracy: 0.6100 - val_loss: 3.0244
25/25
Epoch 125/200
25/25
                          5s 200ms/step - accuracy: 0.9813 - loss: 0.0663 - val_accuracy: 0.6500 - val_loss: 3.1208
Epoch 126/200
25/25
                          5s 215ms/step - accuracy: 0.9855 - loss: 0.0411 - val_accuracy: 0.6500 - val_loss: 2.7819
Epoch 127/200
                          5s 208ms/step - accuracy: 0.9888 - loss: 0.0385 - val_accuracy: 0.6600 - val_loss: 3.1580
25/25
Epoch 128/200
25/25
                          • 5s 203ms/step - accuracy: 0.9817 - loss: 0.0764 - val_accuracy: 0.6650 - val_loss: 2.7910
Epoch 129/200
25/25
                          5s 207ms/step - accuracy: 0.9767 - loss: 0.0670 - val_accuracy: 0.6450 - val_loss: 3.1073
Epoch 130/200
25/25
                           5s 204ms/step - accuracy: 0.9796 - loss: 0.0653 - val_accuracy: 0.6350 - val_loss: 2.9026
Epoch 131/200
25/25
                          5s 208ms/step - accuracy: 0.9897 - loss: 0.0606 - val_accuracy: 0.5650 - val_loss: 4.0656
Epoch 132/200
                          5s 213ms/step - accuracy: 0.9431 - loss: 0.2323 - val_accuracy: 0.6550 - val_loss: 2.6973
25/25
Epoch 133/200
25/25
                          6s 220ms/step - accuracy: 0.9872 - loss: 0.0522 - val_accuracy: 0.6400 - val_loss: 2.6923
Epoch 134/200
25/25
                          5s 201ms/step - accuracy: 0.9980 - loss: 0.0147 - val_accuracy: 0.6550 - val_loss: 2.8682
Epoch 135/200
25/25
                          5s 207ms/step - accuracy: 0.9789 - loss: 0.0659 - val_accuracy: 0.6550 - val_loss: 2.9029
Epoch 136/200
25/25
                          5s 213ms/step - accuracy: 0.9781 - loss: 0.0519 - val_accuracy: 0.6450 - val_loss: 3.3952
Epoch 137/200
25/25
                          5s 204ms/step - accuracy: 0.9841 - loss: 0.0550 - val_accuracy: 0.6950 - val_loss: 2.4342
Epoch 138/200
                          5s 213ms/step - accuracy: 0.9835 - loss: 0.0416 - val_accuracy: 0.5950 - val_loss: 3.0843
25/25
Epoch 139/200
25/25
                          - 5s 212ms/step - accuracy: 0.9796 - loss: 0.0624 - val_accuracy: 0.6200 - val_loss: 3.3547
Epoch 140/200
25/25
                          5s 201ms/step - accuracy: 0.9814 - loss: 0.0590 - val_accuracy: 0.6500 - val_loss: 2.9060
Epoch 141/200
25/25
                          5s 211ms/step - accuracy: 0.9790 - loss: 0.0728 - val_accuracy: 0.6400 - val_loss: 3.7736
Epoch 142/200
25/25
                          5s 218ms/step - accuracy: 0.9811 - loss: 0.0637 - val_accuracy: 0.6450 - val_loss: 3.2781
Epoch 143/200
25/25
                          5s 206ms/step - accuracy: 0.9746 - loss: 0.0844 - val_accuracy: 0.6050 - val_loss: 3.2517
Epoch 144/200
25/25
                          5s 201ms/step - accuracy: 0.9946 - loss: 0.0231 - val_accuracy: 0.6300 - val_loss: 3.2410
Epoch 145/200
25/25
                          5s 213ms/step - accuracy: 0.9646 - loss: 0.1369 - val_accuracy: 0.6450 - val_loss: 3.3306
Epoch 146/200
25/25
                           5s 196ms/step - accuracy: 0.9848 - loss: 0.0446 - val_accuracy: 0.6400 - val_loss: 2.9006
Epoch 147/200
25/25
                          5s 212ms/step - accuracy: 0.9890 - loss: 0.0328 - val_accuracy: 0.6350 - val_loss: 3.0408
Epoch 148/200
                          5s 206ms/step - accuracy: 0.9745 - loss: 0.0681 - val_accuracy: 0.6250 - val_loss: 2.8711
25/25
Epoch 149/200
25/25
                          • 5s 210ms/step - accuracy: 0.9827 - loss: 0.0773 - val_accuracy: 0.6700 - val_loss: 3.0453
Epoch 150/200
25/25
                          5s 203ms/step - accuracy: 0.9862 - loss: 0.0550 - val_accuracy: 0.6600 - val_loss: 2.8548
Epoch 151/200
25/25
                          5s 207ms/step - accuracy: 0.9790 - loss: 0.0703 - val accuracy: 0.6650 - val loss: 2.9448
Epoch 152/200
25/25
                           5s 206ms/step - accuracy: 0.9811 - loss: 0.0715 - val_accuracy: 0.6550 - val_loss: 3.0744
Epoch 153/200
25/25
                          5s 210ms/step - accuracy: 0.9831 - loss: 0.0430 - val_accuracy: 0.6500 - val_loss: 3.1276
Epoch 154/200
25/25
                          5s 208ms/step - accuracy: 0.9720 - loss: 0.0975 - val_accuracy: 0.6400 - val_loss: 2.9376
Epoch 155/200
25/25
                          5s 206ms/step - accuracy: 0.9923 - loss: 0.0249 - val accuracy: 0.6200 - val loss: 3.2989
Epoch 156/200
25/25
                          5s 207ms/step - accuracy: 0.9571 - loss: 0.0950 - val_accuracy: 0.6300 - val_loss: 3.2689
Epoch 157/200
                          5s 202ms/step - accuracy: 0.9922 - loss: 0.0247 - val_accuracy: 0.6300 - val_loss: 3.2124
25/25
Epoch 158/200
25/25
                          5s 199ms/step - accuracy: 0.9918 - loss: 0.0198 - val_accuracy: 0.6200 - val_loss: 3.5101
Epoch 159/200
25/25
                          5s 208ms/step - accuracy: 0.9756 - loss: 0.0976 - val accuracy: 0.6450 - val loss: 3.0890
Epoch 160/200
25/25
                          - 5s 200ms/step - accuracy: 0.9899 - loss: 0.0320 - val_accuracy: 0.5250 - val_loss: 4.3071
Epoch 161/200
25/25
                          5s 200ms/step - accuracy: 0.9500 - loss: 0.1923 - val_accuracy: 0.6150 - val_loss: 2.8679
Epoch 162/200
25/25
                           5s 206ms/step - accuracy: 0.9766 - loss: 0.0628 - val_accuracy: 0.6350 - val_loss: 2.5324
Epoch 163/200
25/25
                          5s 205ms/step - accuracy: 0.9859 - loss: 0.0392 - val_accuracy: 0.6400 - val_loss: 2.7800
Epoch 164/200
25/25
                           5s 205ms/step - accuracy: 0.9729 - loss: 0.0914 - val_accuracy: 0.6250 - val_loss: 3.1907
```

```
Epoch 165/200
                                  5s 206ms/step - accuracy: 0.9707 - loss: 0.0861 - val_accuracy: 0.6700 - val_loss: 3.0659
       25/25
       Epoch 166/200
       25/25
                                 - 5s 201ms/step - accuracy: 0.9804 - loss: 0.0650 - val_accuracy: 0.6550 - val_loss: 2.8892
       Epoch 167/200
      25/25
                                 5s 204ms/step - accuracy: 0.9777 - loss: 0.0900 - val_accuracy: 0.6450 - val_loss: 3.1240
       Epoch 168/200
      25/25
                                 • 5s 207ms/step - accuracy: 0.9789 - loss: 0.0641 - val_accuracy: 0.6600 - val_loss: 2.8889
       Epoch 169/200
      25/25
                                 - 5s 199ms/step - accuracy: 0.9611 - loss: 0.1326 - val_accuracy: 0.6800 - val_loss: 2.6419
      Epoch 170/200
      25/25
                                 - 5s 211ms/step - accuracy: 0.9922 - loss: 0.0444 - val_accuracy: 0.6250 - val_loss: 3.5239
       Epoch 171/200
      25/25
                                  6s 224ms/step - accuracy: 0.9777 - loss: 0.0627 - val_accuracy: 0.6350 - val_loss: 3.0843
       Epoch 172/200
       25/25
                                 • 5s 209ms/step - accuracy: 0.9845 - loss: 0.0833 - val_accuracy: 0.6400 - val_loss: 3.2503
       Epoch 173/200
                                 5s 207ms/step - accuracy: 0.9847 - loss: 0.0406 - val_accuracy: 0.6300 - val_loss: 3.2927
      25/25
      Epoch 174/200
       25/25
                                 - 5s 205ms/step - accuracy: 0.9957 - loss: 0.0156 - val_accuracy: 0.6350 - val_loss: 3.5053
      Epoch 175/200
       25/25
                                 5s 203ms/step - accuracy: 0.9836 - loss: 0.0420 - val_accuracy: 0.5950 - val_loss: 3.7164
       Epoch 176/200
      25/25
                                 5s 204ms/step - accuracy: 0.9834 - loss: 0.0705 - val_accuracy: 0.6450 - val_loss: 3.0541
       Epoch 177/200
                                 - 5s 204ms/step - accuracy: 0.9733 - loss: 0.0997 - val_accuracy: 0.6600 - val_loss: 3.0176
       25/25
       Epoch 178/200
      25/25
                                 - 5s 195ms/step - accuracy: 0.9859 - loss: 0.0259 - val_accuracy: 0.6150 - val_loss: 3.4758
      Epoch 179/200
                                 • 5s 199ms/step - accuracy: 0.9645 - loss: 0.1496 - val_accuracy: 0.6200 - val_loss: 3.6725
       25/25
       Epoch 180/200
      25/25
                                - 5s 203ms/step - accuracy: 0.9876 - loss: 0.0493 - val_accuracy: 0.6200 - val_loss: 3.2856
       Epoch 181/200
      25/25
                                 - 5s 195ms/step - accuracy: 0.9886 - loss: 0.0565 - val_accuracy: 0.6250 - val_loss: 3.9449
       Epoch 182/200
      25/25 •
                                 - 5s 202ms/step - accuracy: 0.9770 - loss: 0.0616 - val_accuracy: 0.6200 - val_loss: 3.2310
       Epoch 183/200
       25/25
                                 5s 213ms/step - accuracy: 0.9864 - loss: 0.0475 - val_accuracy: 0.6450 - val_loss: 3.0371
       Epoch 184/200
      25/25
                                 6s 228ms/step - accuracy: 0.9862 - loss: 0.0486 - val_accuracy: 0.6350 - val_loss: 3.6485
       Epoch 185/200
                                 - 5s 207ms/step - accuracy: 0.9795 - loss: 0.0680 - val_accuracy: 0.6100 - val_loss: 3.3821
       25/25
       Epoch 186/200
      25/25
                                 - 5s 197ms/step - accuracy: 0.9834 - loss: 0.0764 - val_accuracy: 0.6450 - val_loss: 3.2714
       Epoch 187/200
       25/25 -
                                 5s 201ms/step - accuracy: 0.9888 - loss: 0.0797 - val_accuracy: 0.6800 - val_loss: 3.0651
       Epoch 188/200
      25/25
                                 - 5s 198ms/step - accuracy: 0.9895 - loss: 0.0452 - val_accuracy: 0.6000 - val_loss: 3.8176
       Epoch 189/200
      25/25
                                 - 5s 212ms/step - accuracy: 0.9764 - loss: 0.0732 - val_accuracy: 0.6300 - val_loss: 3.2673
      Epoch 190/200
      25/25
                                 - 5s 208ms/step - accuracy: 0.9867 - loss: 0.0555 - val_accuracy: 0.6400 - val_loss: 3.1765
      Epoch 191/200
       25/25
                                 5s 206ms/step - accuracy: 0.9903 - loss: 0.0231 - val_accuracy: 0.6750 - val_loss: 3.3168
       Epoch 192/200
      25/25
                                 • 5s 204ms/step - accuracy: 0.9844 - loss: 0.0556 - val_accuracy: 0.6400 - val_loss: 2.8865
       Epoch 193/200
       25/25
                                 - 5s 206ms/step - accuracy: 0.9779 - loss: 0.0447 - val_accuracy: 0.6750 - val_loss: 3.1290
       Epoch 194/200
      25/25
                                 5s 210ms/step - accuracy: 0.9906 - loss: 0.0313 - val_accuracy: 0.5900 - val_loss: 3.4275
      Epoch 195/200
       25/25
                                 5s 206ms/step - accuracy: 0.9902 - loss: 0.0367 - val_accuracy: 0.6250 - val_loss: 3.2306
      Epoch 196/200
      25/25
                                 • 5s 215ms/step - accuracy: 0.9877 - loss: 0.0718 - val_accuracy: 0.5900 - val_loss: 3.6389
       Epoch 197/200
      25/25
                                 • 5s 211ms/step - accuracy: 0.9846 - loss: 0.0554 - val_accuracy: 0.6650 - val_loss: 3.2400
       Epoch 198/200
       25/25
                                 5s 209ms/step - accuracy: 0.9880 - loss: 0.0415 - val_accuracy: 0.6450 - val_loss: 3.2804
       Epoch 199/200
       25/25
                                 5s 197ms/step - accuracy: 0.9821 - loss: 0.0445 - val_accuracy: 0.6550 - val_loss: 3.3964
       Epoch 200/200
       25/25
                                 5s 212ms/step - accuracy: 0.9776 - loss: 0.0669 - val accuracy: 0.6150 - val loss: 3.7908
In [8]: print(f"Execution time: {elapsed_time:.2f} seconds")
       Execution time: 1036.82 seconds
```

In [9]: def append_core_data(score_path, num_cores, elapsed_time):
 # Check if the file already exists

Check if the file already exists
file_exists = os.path.exists(score_path)

Open the file in append mode
with open(score_path, mode='a', newline='') as file:

```
writer = csv.writer(file)

# If the file is new, write the header
if not file_exists:
    writer.writerow(["Number of Cores", "Elapsed Time"])

# Write the new data
writer.writerow([num_cores, elapsed_time])
```

```
In [1]: %matplotlib inline
        import shutil
        import random
        import numpy as np
        from warnings import filterwarnings
        filterwarnings('ignore')
        from tensorflow.keras import layers, regularizers, optimizers
        from tensorflow.keras import models
        from tensorflow.keras.models import Sequential, Model
        from tensorflow.keras.layers import LeakyReLU,Dense, Activation, Flatten, Dropout, BatchNormalization,Conv2D, MaxPooling2D
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import tensorflow as tf
        import os
        import time
        import csv
        from matplotlib import figure
```

Define 5 worker

```
In [2]: # Set the number of threads
number_of_worker = 5
os.environ['OMP_NUM_THREADS'] = '5' # OpenMP threads for parallelism
os.environ['TF_NUM_INTEROP_THREADS'] = '5' # Threads for inter-operation parallelism
os.environ['TF_NUM_INTRAOP_THREADS'] = '5' # Threads for intra-operation parallelism
# Confirm TensorFlow is using the specified number of threads
tf.config.threading.set_inter_op_parallelism_threads(number_of_worker)
tf.config.threading.set_intra_op_parallelism_threads(number_of_worker)
```

```
In [3]: source_dir = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DATA\Convert_Audio_File_to_jpg_file"
        target_dir = r'genres_train_val_split_data'
        split_ratio = 0.8
        def Train_Test_Split(source_dir,target_dir,split_ratio):
            # Define source and target directories
            train_dir = os.path.join(target_dir, 'train')
            val_dir = os.path.join(target_dir, 'val')
            # Create target directories if they don't exist
            os.makedirs(train_dir, exist_ok=True)
            os.makedirs(val_dir, exist_ok=True)
            # Get the list of class directories
            classes = [d for d in os.listdir(source_dir) if os.path.isdir(os.path.join(source_dir, d))]
            for class name in classes:
                # Create class directories in train and val folders
                os.makedirs(os.path.join(train_dir, class_name), exist_ok=True)
                os.makedirs(os.path.join(val_dir, class_name), exist_ok=True)
                # Get list of images in the class directory
                class_dir = os.path.join(source_dir, class_name)
                images = [f for f in os.listdir(class_dir) if os.path.isfile(os.path.join(class_dir, f))]
                # Shuffle the images
                random.shuffle(images)
                # Compute the split point
                split_point = int(len(images) * split_ratio)
                # Split the images into training and validation sets
                train images = images[:split point]
                val_images = images[split_point:]
```

Data split completed successfully!

Load the Data

```
In [5]: WIDTH = 64
        HEIGHT = 64
        BATCH_SIZE = 32
        TRAIN_DIR=r'genres_train_val_split_data/train'
        val_dir = r'genres_train_val_split_data/val'
        # data prep
        train_datagen = ImageDataGenerator(
            rescale=1./255.,validation_split=0.25)
        train_generator = train_datagen.flow_from_directory(
            TRAIN_DIR,
            target size=(HEIGHT, WIDTH),
                batch_size=BATCH_SIZE,
                class_mode='categorical')
        validation_gen = train_datagen.flow_from_directory(
            val_dir,target_size = (HEIGHT,WIDTH),
            batch_size = BATCH_SIZE,
            class_mode = 'categorical'
```

Found 800 images belonging to 10 classes. Found 200 images belonging to 10 classes.

Model Architecture

```
In [6]: model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                         input_shape=(64,64,3)))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(128, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(128, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(10, activation='softmax'))
        model.compile(optimizers.RMSprop(learning_rate=0.0005, decay=1e-6),loss="categorical_crossentropy",metrics=["accuracy"])
        model.summary()
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 64, 64, 32)	896
activation (Activation)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 62, 62, 64)	18,496
activation_1 (Activation)	(None, 62, 62, 64)	0
max_pooling2d (MaxPooling2D)	(None, 31, 31, 64)	0
dropout (Dropout)	(None, 31, 31, 64)	0
conv2d_2 (Conv2D)	(None, 31, 31, 64)	36,928
activation_2 (Activation)	(None, 31, 31, 64)	0
conv2d_3 (Conv2D)	(None, 29, 29, 64)	36,928
activation_3 (Activation)	(None, 29, 29, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_1 (Dropout)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 14, 14, 128)	73,856
activation_4 (Activation)	(None, 14, 14, 128)	0
conv2d_5 (Conv2D)	(None, 12, 12, 128)	147,584
activation_5 (Activation)	(None, 12, 12, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_2 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2,359,808
activation_6 (Activation)	(None, 512)	0
dropout_3 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5,130

```
In [7]: STEP_SIZE_TRAIN=train_generator.n//train_generator.batch_size
# Measure the execution time
start_time = time.time()

model.fit(train_generator,validation_data=validation_gen,epochs=200)

end_time = time.time()
elapsed_time = end_time - start_time
```

```
Epoch 1/200
                           5s 153ms/step - accuracy: 0.0947 - loss: 2.3395 - val_accuracy: 0.1800 - val_loss: 2.2968
25/25
Epoch 2/200
25/25 -
                           5s 189ms/step - accuracy: 0.1267 - loss: 2.2874 - val_accuracy: 0.2250 - val_loss: 2.2684
Epoch 3/200
25/25
                           5s 194ms/step - accuracy: 0.1525 - loss: 2.2579 - val_accuracy: 0.1450 - val_loss: 2.2890
Epoch 4/200
                           5s 191ms/step - accuracy: 0.1746 - loss: 2.2096 - val_accuracy: 0.1300 - val_loss: 2.3036
25/25
Epoch 5/200
25/25
                          • 5s 192ms/step - accuracy: 0.2015 - loss: 2.1897 - val_accuracy: 0.2700 - val_loss: 1.9919
Epoch 6/200
25/25
                          5s 193ms/step - accuracy: 0.2600 - loss: 2.0449 - val_accuracy: 0.3150 - val_loss: 1.9102
Epoch 7/200
25/25
                           5s 193ms/step - accuracy: 0.2192 - loss: 2.1385 - val_accuracy: 0.3200 - val_loss: 1.9702
Epoch 8/200
25/25
                           5s 193ms/step - accuracy: 0.2778 - loss: 1.9687 - val_accuracy: 0.2950 - val_loss: 2.0048
Epoch 9/200
                           5s 194ms/step - accuracy: 0.2959 - loss: 1.9138 - val_accuracy: 0.3400 - val_loss: 1.8138
25/25
Epoch 10/200
25/25
                           5s 195ms/step - accuracy: 0.3270 - loss: 1.8565 - val_accuracy: 0.3700 - val_loss: 1.8105
Epoch 11/200
25/25
                           5s 195ms/step - accuracy: 0.3472 - loss: 1.7873 - val_accuracy: 0.2900 - val_loss: 1.9588
Epoch 12/200
25/25
                           5s 197ms/step - accuracy: 0.3980 - loss: 1.6698 - val_accuracy: 0.4450 - val_loss: 1.6245
Epoch 13/200
25/25
                           5s 194ms/step - accuracy: 0.4249 - loss: 1.6145 - val_accuracy: 0.4050 - val_loss: 1.6579
Epoch 14/200
25/25
                           5s 195ms/step - accuracy: 0.3869 - loss: 1.6297 - val_accuracy: 0.4300 - val_loss: 1.5762
Epoch 15/200
                           5s 197ms/step - accuracy: 0.3926 - loss: 1.6391 - val_accuracy: 0.4050 - val_loss: 1.6126
25/25
Epoch 16/200
25/25
                          - 5s 186ms/step - accuracy: 0.4444 - loss: 1.5492 - val_accuracy: 0.4450 - val_loss: 1.4956
Epoch 17/200
25/25
                           5s 197ms/step - accuracy: 0.3875 - loss: 1.6314 - val_accuracy: 0.4400 - val_loss: 1.5098
Epoch 18/200
25/25
                           5s 192ms/step - accuracy: 0.4552 - loss: 1.4967 - val_accuracy: 0.4350 - val_loss: 1.5309
Epoch 19/200
25/25
                           5s 191ms/step - accuracy: 0.4820 - loss: 1.3829 - val_accuracy: 0.4600 - val_loss: 1.4706
Epoch 20/200
25/25
                           5s 194ms/step - accuracy: 0.4636 - loss: 1.4114 - val_accuracy: 0.4750 - val_loss: 1.4325
Epoch 21/200
25/25
                           5s 192ms/step - accuracy: 0.5140 - loss: 1.3883 - val_accuracy: 0.4900 - val_loss: 1.4102
Epoch 22/200
25/25
                           5s 185ms/step - accuracy: 0.5524 - loss: 1.2626 - val_accuracy: 0.4800 - val_loss: 1.4580
Epoch 23/200
25/25
                           5s 195ms/step - accuracy: 0.5119 - loss: 1.3195 - val_accuracy: 0.5000 - val_loss: 1.3673
Epoch 24/200
25/25
                          5s 186ms/step - accuracy: 0.5571 - loss: 1.2439 - val_accuracy: 0.4900 - val_loss: 1.3707
Epoch 25/200
                           5s 190ms/step - accuracy: 0.5772 - loss: 1.1844 - val_accuracy: 0.5100 - val_loss: 1.3961
25/25
Epoch 26/200
25/25
                          5s 187ms/step - accuracy: 0.5738 - loss: 1.1956 - val_accuracy: 0.5150 - val_loss: 1.3461
Epoch 27/200
25/25
                           5s 188ms/step - accuracy: 0.5962 - loss: 1.1263 - val_accuracy: 0.5350 - val_loss: 1.3342
Epoch 28/200
25/25
                           5s 191ms/step - accuracy: 0.5756 - loss: 1.1709 - val_accuracy: 0.5850 - val_loss: 1.2785
Epoch 29/200
25/25 -
                           5s 194ms/step - accuracy: 0.6157 - loss: 1.0428 - val_accuracy: 0.5050 - val_loss: 1.3166
Epoch 30/200
25/25
                           5s 188ms/step - accuracy: 0.5870 - loss: 1.1603 - val_accuracy: 0.5200 - val_loss: 1.3354
Epoch 31/200
25/25
                           5s 190ms/step - accuracy: 0.5897 - loss: 1.0747 - val_accuracy: 0.5750 - val_loss: 1.3224
Epoch 32/200
25/25
                           5s 187ms/step - accuracy: 0.6535 - loss: 0.9584 - val accuracy: 0.5150 - val loss: 1.3154
Epoch 33/200
25/25
                           5s 190ms/step - accuracy: 0.6701 - loss: 0.9027 - val_accuracy: 0.5900 - val_loss: 1.2023
Epoch 34/200
                           5s 187ms/step - accuracy: 0.6354 - loss: 0.9593 - val_accuracy: 0.6050 - val_loss: 1.1918
25/25
Epoch 35/200
25/25
                           5s 192ms/step - accuracy: 0.6474 - loss: 0.9642 - val_accuracy: 0.5800 - val_loss: 1.2895
Epoch 36/200
25/25
                           5s 200ms/step - accuracy: 0.7099 - loss: 0.7981 - val accuracy: 0.5900 - val loss: 1.2998
Epoch 37/200
25/25
                          - 5s 182ms/step - accuracy: 0.6821 - loss: 0.8373 - val_accuracy: 0.5450 - val_loss: 1.3157
Epoch 38/200
25/25
                           5s 198ms/step - accuracy: 0.7395 - loss: 0.7698 - val_accuracy: 0.5550 - val_loss: 1.3241
Epoch 39/200
25/25
                           5s 194ms/step - accuracy: 0.7437 - loss: 0.7579 - val_accuracy: 0.5700 - val_loss: 1.3051
Epoch 40/200
25/25
                           5s 200ms/step - accuracy: 0.7123 - loss: 0.7496 - val_accuracy: 0.6300 - val_loss: 1.2596
Epoch 41/200
25/25
                           5s 193ms/step - accuracy: 0.7576 - loss: 0.6787 - val_accuracy: 0.5650 - val_loss: 1.4545
```

```
Epoch 42/200
                           5s 204ms/step - accuracy: 0.7592 - loss: 0.7017 - val_accuracy: 0.5650 - val_loss: 1.4580
25/25
Epoch 43/200
25/25
                          5s 196ms/step - accuracy: 0.7450 - loss: 0.6825 - val_accuracy: 0.5950 - val_loss: 1.3154
Epoch 44/200
25/25
                           5s 200ms/step - accuracy: 0.8020 - loss: 0.5688 - val_accuracy: 0.5600 - val_loss: 1.3848
Epoch 45/200
                          5s 197ms/step - accuracy: 0.8079 - loss: 0.5317 - val_accuracy: 0.5700 - val_loss: 1.4590
25/25
Epoch 46/200
25/25
                          5s 193ms/step - accuracy: 0.7713 - loss: 0.6108 - val_accuracy: 0.5650 - val_loss: 1.3786
Epoch 47/200
25/25
                          5s 192ms/step - accuracy: 0.7668 - loss: 0.6010 - val_accuracy: 0.5900 - val_loss: 1.4130
Epoch 48/200
25/25
                           5s 197ms/step - accuracy: 0.8221 - loss: 0.4975 - val_accuracy: 0.5700 - val_loss: 1.4629
Epoch 49/200
25/25
                          5s 197ms/step - accuracy: 0.7979 - loss: 0.5379 - val_accuracy: 0.5650 - val_loss: 1.4927
Epoch 50/200
                           5s 191ms/step - accuracy: 0.8335 - loss: 0.4722 - val_accuracy: 0.6000 - val_loss: 1.5870
25/25
Epoch 51/200
25/25
                          5s 191ms/step - accuracy: 0.8321 - loss: 0.4891 - val_accuracy: 0.5650 - val_loss: 1.8292
Epoch 52/200
25/25
                          5s 194ms/step - accuracy: 0.8426 - loss: 0.4643 - val_accuracy: 0.6250 - val_loss: 1.5597
Epoch 53/200
25/25
                          5s 191ms/step - accuracy: 0.8810 - loss: 0.3997 - val_accuracy: 0.5900 - val_loss: 1.5873
Epoch 54/200
25/25
                          5s 188ms/step - accuracy: 0.8325 - loss: 0.4602 - val_accuracy: 0.6000 - val_loss: 1.7301
Epoch 55/200
25/25
                           5s 196ms/step - accuracy: 0.8486 - loss: 0.4165 - val_accuracy: 0.5950 - val_loss: 1.6603
Epoch 56/200
                          5s 194ms/step - accuracy: 0.8677 - loss: 0.3930 - val_accuracy: 0.6050 - val_loss: 1.5699
25/25
Epoch 57/200
25/25
                          - 5s 189ms/step - accuracy: 0.8873 - loss: 0.3440 - val_accuracy: 0.6050 - val_loss: 1.5525
Epoch 58/200
25/25
                          5s 199ms/step - accuracy: 0.8883 - loss: 0.3339 - val_accuracy: 0.5850 - val_loss: 1.8909
Epoch 59/200
25/25
                          5s 185ms/step - accuracy: 0.8920 - loss: 0.2942 - val_accuracy: 0.5700 - val_loss: 1.7802
Epoch 60/200
25/25
                          5s 200ms/step - accuracy: 0.9011 - loss: 0.3193 - val_accuracy: 0.5950 - val_loss: 2.0626
Epoch 61/200
25/25
                          5s 190ms/step - accuracy: 0.8948 - loss: 0.3518 - val_accuracy: 0.6150 - val_loss: 1.9260
Epoch 62/200
25/25
                          5s 191ms/step - accuracy: 0.8986 - loss: 0.2625 - val_accuracy: 0.5950 - val_loss: 1.8665
Epoch 63/200
25/25
                          5s 190ms/step - accuracy: 0.9059 - loss: 0.2910 - val_accuracy: 0.6100 - val_loss: 1.7793
Epoch 64/200
25/25
                           5s 191ms/step - accuracy: 0.9151 - loss: 0.2510 - val_accuracy: 0.6300 - val_loss: 1.8918
Epoch 65/200
25/25
                          5s 187ms/step - accuracy: 0.9331 - loss: 0.2084 - val accuracy: 0.5800 - val loss: 1.9621
Epoch 66/200
                          5s 194ms/step - accuracy: 0.9136 - loss: 0.2767 - val_accuracy: 0.6150 - val_loss: 1.8002
25/25
Epoch 67/200
25/25
                          5s 195ms/step - accuracy: 0.9290 - loss: 0.2592 - val_accuracy: 0.6050 - val_loss: 2.0399
Epoch 68/200
25/25
                          5s 195ms/step - accuracy: 0.8988 - loss: 0.2698 - val_accuracy: 0.5950 - val_loss: 2.1498
Epoch 69/200
25/25
                          5s 190ms/step - accuracy: 0.9527 - loss: 0.1518 - val_accuracy: 0.6100 - val_loss: 2.1151
Epoch 70/200
25/25 -
                           5s 188ms/step - accuracy: 0.9299 - loss: 0.1859 - val_accuracy: 0.6050 - val_loss: 2.1399
Epoch 71/200
25/25
                           5s 186ms/step - accuracy: 0.9354 - loss: 0.1745 - val_accuracy: 0.5850 - val_loss: 2.1911
Epoch 72/200
25/25
                           5s 196ms/step - accuracy: 0.9467 - loss: 0.1526 - val_accuracy: 0.5800 - val_loss: 2.0856
Epoch 73/200
25/25
                          5s 191ms/step - accuracy: 0.9430 - loss: 0.1523 - val accuracy: 0.5550 - val loss: 2.4861
Epoch 74/200
25/25
                           5s 188ms/step - accuracy: 0.9512 - loss: 0.1692 - val_accuracy: 0.5850 - val_loss: 2.1487
Epoch 75/200
                           5s 194ms/step - accuracy: 0.9661 - loss: 0.1040 - val_accuracy: 0.6400 - val_loss: 2.2880
25/25
Epoch 76/200
25/25
                          5s 192ms/step - accuracy: 0.9370 - loss: 0.1945 - val_accuracy: 0.6150 - val_loss: 2.5149
Epoch 77/200
25/25
                          5s 188ms/step - accuracy: 0.9633 - loss: 0.2527 - val accuracy: 0.6050 - val loss: 2.1542
Epoch 78/200
25/25
                          - 5s 194ms/step - accuracy: 0.9281 - loss: 0.2165 - val_accuracy: 0.5850 - val_loss: 2.1198
Epoch 79/200
25/25
                          5s 190ms/step - accuracy: 0.9671 - loss: 0.1049 - val_accuracy: 0.5950 - val_loss: 2.1969
Epoch 80/200
25/25
                           5s 192ms/step - accuracy: 0.9430 - loss: 0.1505 - val_accuracy: 0.5700 - val_loss: 2.4307
Epoch 81/200
25/25
                          5s 185ms/step - accuracy: 0.9560 - loss: 0.1293 - val_accuracy: 0.5650 - val_loss: 2.3232
Epoch 82/200
25/25
                           5s 186ms/step - accuracy: 0.9472 - loss: 0.1393 - val_accuracy: 0.5800 - val_loss: 2.6178
```

```
Epoch 83/200
                           5s 199ms/step - accuracy: 0.9441 - loss: 0.1490 - val_accuracy: 0.5950 - val_loss: 2.2539
25/25
Epoch 84/200
25/25
                          5s 189ms/step - accuracy: 0.9621 - loss: 0.1124 - val_accuracy: 0.5850 - val_loss: 2.4427
Epoch 85/200
25/25
                           5s 195ms/step - accuracy: 0.9460 - loss: 0.1516 - val_accuracy: 0.6250 - val_loss: 2.1418
Epoch 86/200
25/25
                          5s 200ms/step - accuracy: 0.9625 - loss: 0.1522 - val_accuracy: 0.6050 - val_loss: 2.3322
Epoch 87/200
25/25
                          5s 189ms/step - accuracy: 0.9580 - loss: 0.1109 - val_accuracy: 0.6100 - val_loss: 2.1526
Epoch 88/200
25/25
                          5s 195ms/step - accuracy: 0.9683 - loss: 0.1126 - val_accuracy: 0.6200 - val_loss: 2.5570
Epoch 89/200
25/25
                           5s 189ms/step - accuracy: 0.9610 - loss: 0.1214 - val_accuracy: 0.5750 - val_loss: 2.4782
Epoch 90/200
25/25
                          5s 196ms/step - accuracy: 0.9599 - loss: 0.1022 - val_accuracy: 0.5550 - val_loss: 2.3816
Epoch 91/200
                           5s 196ms/step - accuracy: 0.9712 - loss: 0.1005 - val_accuracy: 0.5700 - val_loss: 2.6433
25/25
Epoch 92/200
25/25
                          5s 186ms/step - accuracy: 0.9472 - loss: 0.1302 - val_accuracy: 0.5750 - val_loss: 3.0499
Epoch 93/200
25/25
                          5s 193ms/step - accuracy: 0.9653 - loss: 0.1075 - val_accuracy: 0.5850 - val_loss: 2.9033
Epoch 94/200
25/25
                          5s 194ms/step - accuracy: 0.9574 - loss: 0.1204 - val_accuracy: 0.5850 - val_loss: 2.7043
Epoch 95/200
                          5s 190ms/step - accuracy: 0.9684 - loss: 0.1105 - val_accuracy: 0.6000 - val_loss: 2.6137
25/25
Epoch 96/200
25/25
                           5s 185ms/step - accuracy: 0.9573 - loss: 0.1077 - val_accuracy: 0.6000 - val_loss: 2.3739
Epoch 97/200
                          5s 193ms/step - accuracy: 0.9490 - loss: 0.1130 - val_accuracy: 0.6100 - val_loss: 2.3640
25/25
Epoch 98/200
25/25
                          - 5s 194ms/step - accuracy: 0.9790 - loss: 0.0706 - val_accuracy: 0.5750 - val_loss: 3.0800
Epoch 99/200
25/25
                          5s 194ms/step - accuracy: 0.9857 - loss: 0.0605 - val_accuracy: 0.6050 - val_loss: 2.4089
Epoch 100/200
25/25
                          5s 194ms/step - accuracy: 0.9794 - loss: 0.0811 - val_accuracy: 0.5750 - val_loss: 2.9083
Epoch 101/200
25/25
                          5s 194ms/step - accuracy: 0.9759 - loss: 0.0625 - val_accuracy: 0.6000 - val_loss: 3.0784
Epoch 102/200
25/25
                          5s 190ms/step - accuracy: 0.9781 - loss: 0.0984 - val_accuracy: 0.6300 - val_loss: 2.7971
Epoch 103/200
25/25
                          5s 190ms/step - accuracy: 0.9554 - loss: 0.1220 - val_accuracy: 0.6150 - val_loss: 2.3903
Epoch 104/200
25/25
                          5s 187ms/step - accuracy: 0.9743 - loss: 0.0762 - val_accuracy: 0.5700 - val_loss: 2.5217
Epoch 105/200
25/25
                           5s 193ms/step - accuracy: 0.9893 - loss: 0.0418 - val_accuracy: 0.5900 - val_loss: 2.6067
Epoch 106/200
25/25
                          5s 192ms/step - accuracy: 0.9804 - loss: 0.0651 - val_accuracy: 0.6000 - val_loss: 2.2410
Epoch 107/200
                          5s 187ms/step - accuracy: 0.9545 - loss: 0.0943 - val_accuracy: 0.5700 - val_loss: 2.6078
25/25
Epoch 108/200
25/25
                          • 5s 190ms/step - accuracy: 0.9713 - loss: 0.0938 - val_accuracy: 0.5700 - val_loss: 2.6523
Epoch 109/200
25/25
                          5s 194ms/step - accuracy: 0.9663 - loss: 0.1164 - val_accuracy: 0.6000 - val_loss: 2.6464
Epoch 110/200
25/25
                          5s 194ms/step - accuracy: 0.9702 - loss: 0.0860 - val accuracy: 0.5700 - val loss: 2.7814
Epoch 111/200
25/25
                           5s 196ms/step - accuracy: 0.9639 - loss: 0.1292 - val_accuracy: 0.6100 - val_loss: 2.4844
Epoch 112/200
25/25
                          5s 188ms/step - accuracy: 0.9848 - loss: 0.0520 - val_accuracy: 0.6100 - val_loss: 3.0841
Epoch 113/200
25/25
                          5s 196ms/step - accuracy: 0.9740 - loss: 0.0715 - val_accuracy: 0.5800 - val_loss: 2.8770
Epoch 114/200
25/25
                          5s 199ms/step - accuracy: 0.9860 - loss: 0.0509 - val accuracy: 0.6000 - val loss: 3.2372
Epoch 115/200
25/25
                           5s 192ms/step - accuracy: 0.9700 - loss: 0.0779 - val_accuracy: 0.5750 - val_loss: 3.0240
Epoch 116/200
                          5s 194ms/step - accuracy: 0.9712 - loss: 0.1071 - val_accuracy: 0.5850 - val_loss: 2.9968
25/25
Epoch 117/200
25/25
                          5s 191ms/step - accuracy: 0.9741 - loss: 0.0612 - val_accuracy: 0.6000 - val_loss: 3.3724
Epoch 118/200
25/25
                          5s 190ms/step - accuracy: 0.9812 - loss: 0.0701 - val accuracy: 0.6150 - val loss: 2.8990
Epoch 119/200
25/25
                          - 5s 190ms/step - accuracy: 0.9770 - loss: 0.0852 - val_accuracy: 0.5900 - val_loss: 2.3540
Epoch 120/200
25/25
                          5s 188ms/step - accuracy: 0.9712 - loss: 0.0952 - val_accuracy: 0.6200 - val_loss: 2.4252
Epoch 121/200
25/25
                           5s 189ms/step - accuracy: 0.9906 - loss: 0.0291 - val_accuracy: 0.5500 - val_loss: 3.8702
Epoch 122/200
25/25
                          5s 191ms/step - accuracy: 0.9757 - loss: 0.0829 - val_accuracy: 0.6100 - val_loss: 3.0957
Epoch 123/200
25/25
                           5s 193ms/step - accuracy: 0.9720 - loss: 0.0825 - val_accuracy: 0.5850 - val_loss: 3.0245
```

```
Epoch 124/200
                           5s 197ms/step - accuracy: 0.9676 - loss: 0.1079 - val_accuracy: 0.5900 - val_loss: 3.5802
25/25
Epoch 125/200
25/25
                          5s 191ms/step - accuracy: 0.9640 - loss: 0.1216 - val_accuracy: 0.5550 - val_loss: 2.6271
Epoch 126/200
25/25
                          5s 200ms/step - accuracy: 0.9681 - loss: 0.1053 - val_accuracy: 0.5950 - val_loss: 3.2023
Epoch 127/200
                          5s 195ms/step - accuracy: 0.9804 - loss: 0.0468 - val_accuracy: 0.5600 - val_loss: 2.7726
25/25
Epoch 128/200
25/25
                          • 5s 191ms/step - accuracy: 0.9893 - loss: 0.0385 - val_accuracy: 0.5950 - val_loss: 3.2134
Epoch 129/200
25/25
                          5s 192ms/step - accuracy: 0.9809 - loss: 0.0590 - val_accuracy: 0.5950 - val_loss: 2.9174
Epoch 130/200
25/25
                           5s 200ms/step - accuracy: 0.9780 - loss: 0.0603 - val_accuracy: 0.5900 - val_loss: 2.7604
Epoch 131/200
25/25
                          5s 194ms/step - accuracy: 0.9894 - loss: 0.0479 - val_accuracy: 0.5800 - val_loss: 3.1874
Epoch 132/200
                          5s 196ms/step - accuracy: 0.9859 - loss: 0.0606 - val_accuracy: 0.5750 - val_loss: 3.1136
25/25
Epoch 133/200
25/25
                          5s 194ms/step - accuracy: 0.9753 - loss: 0.0876 - val_accuracy: 0.6050 - val_loss: 2.9528
Epoch 134/200
25/25
                          5s 194ms/step - accuracy: 0.9896 - loss: 0.0464 - val_accuracy: 0.5900 - val_loss: 3.2710
Epoch 135/200
25/25
                          5s 192ms/step - accuracy: 0.9720 - loss: 0.1045 - val_accuracy: 0.5950 - val_loss: 2.7128
Epoch 136/200
25/25
                          5s 191ms/step - accuracy: 0.9933 - loss: 0.0282 - val_accuracy: 0.6050 - val_loss: 3.0566
Epoch 137/200
25/25
                          5s 195ms/step - accuracy: 0.9833 - loss: 0.0745 - val_accuracy: 0.5650 - val_loss: 3.8128
Epoch 138/200
                          5s 191ms/step - accuracy: 0.9689 - loss: 0.0789 - val_accuracy: 0.6200 - val_loss: 2.9367
25/25
Epoch 139/200
25/25
                          - 5s 195ms/step - accuracy: 0.9794 - loss: 0.0616 - val_accuracy: 0.5650 - val_loss: 2.7952
Epoch 140/200
25/25
                          5s 201ms/step - accuracy: 0.9961 - loss: 0.0180 - val_accuracy: 0.5850 - val_loss: 3.2651
Epoch 141/200
25/25
                          5s 195ms/step - accuracy: 0.9910 - loss: 0.0361 - val_accuracy: 0.5300 - val_loss: 3.2572
Epoch 142/200
25/25
                          5s 191ms/step - accuracy: 0.9853 - loss: 0.0725 - val_accuracy: 0.5800 - val_loss: 2.8074
Epoch 143/200
25/25
                          5s 189ms/step - accuracy: 0.9817 - loss: 0.0475 - val_accuracy: 0.5700 - val_loss: 3.0449
Epoch 144/200
25/25
                          5s 196ms/step - accuracy: 0.9891 - loss: 0.0474 - val_accuracy: 0.5900 - val_loss: 3.5814
Epoch 145/200
25/25
                          5s 186ms/step - accuracy: 0.9811 - loss: 0.0532 - val_accuracy: 0.5900 - val_loss: 3.5755
Epoch 146/200
25/25
                           5s 187ms/step - accuracy: 0.9628 - loss: 0.1155 - val_accuracy: 0.5650 - val_loss: 3.1388
Epoch 147/200
25/25
                          5s 198ms/step - accuracy: 0.9839 - loss: 0.0366 - val_accuracy: 0.6000 - val_loss: 3.2828
Epoch 148/200
                          5s 187ms/step - accuracy: 0.9905 - loss: 0.0346 - val_accuracy: 0.5800 - val_loss: 2.7950
25/25
Epoch 149/200
25/25
                          • 5s 192ms/step - accuracy: 0.9843 - loss: 0.0378 - val_accuracy: 0.5900 - val_loss: 3.3712
Epoch 150/200
25/25
                          5s 195ms/step - accuracy: 0.9826 - loss: 0.0814 - val_accuracy: 0.5600 - val_loss: 3.1906
Epoch 151/200
25/25
                          5s 196ms/step - accuracy: 0.9868 - loss: 0.0411 - val_accuracy: 0.6100 - val_loss: 3.3522
Epoch 152/200
25/25
                           5s 196ms/step - accuracy: 0.9826 - loss: 0.0764 - val_accuracy: 0.5850 - val_loss: 3.1451
Epoch 153/200
25/25
                           5s 192ms/step - accuracy: 0.9711 - loss: 0.0864 - val_accuracy: 0.5700 - val_loss: 3.3927
Epoch 154/200
25/25
                          5s 193ms/step - accuracy: 0.9854 - loss: 0.0434 - val_accuracy: 0.5650 - val_loss: 3.5499
Epoch 155/200
25/25
                          5s 202ms/step - accuracy: 0.9890 - loss: 0.0358 - val accuracy: 0.5900 - val loss: 3.3332
Epoch 156/200
25/25
                          5s 186ms/step - accuracy: 0.9874 - loss: 0.0431 - val_accuracy: 0.5850 - val_loss: 3.6944
Epoch 157/200
25/25
                          5s 192ms/step - accuracy: 0.9905 - loss: 0.0310 - val_accuracy: 0.5700 - val_loss: 3.6082
Epoch 158/200
25/25
                          5s 192ms/step - accuracy: 0.9814 - loss: 0.0555 - val_accuracy: 0.5950 - val_loss: 3.4962
Epoch 159/200
25/25
                          5s 189ms/step - accuracy: 0.9585 - loss: 0.0979 - val accuracy: 0.6000 - val loss: 3.2705
Epoch 160/200
25/25
                          - 5s 192ms/step - accuracy: 0.9963 - loss: 0.0190 - val_accuracy: 0.5800 - val_loss: 3.3270
Epoch 161/200
25/25
                          5s 190ms/step - accuracy: 0.9895 - loss: 0.0377 - val_accuracy: 0.5650 - val_loss: 2.9575
Epoch 162/200
25/25
                           5s 194ms/step - accuracy: 0.9772 - loss: 0.0976 - val_accuracy: 0.5950 - val_loss: 3.2801
Epoch 163/200
25/25
                          5s 189ms/step - accuracy: 0.9748 - loss: 0.0613 - val_accuracy: 0.5600 - val_loss: 2.9793
Epoch 164/200
25/25
                           5s 193ms/step - accuracy: 0.9830 - loss: 0.0523 - val_accuracy: 0.6050 - val_loss: 3.0790
```

```
Epoch 165/200
                                  5s 192ms/step - accuracy: 0.9915 - loss: 0.0286 - val_accuracy: 0.6050 - val_loss: 3.7859
       25/25
       Epoch 166/200
       25/25
                                 - 5s 186ms/step - accuracy: 0.9891 - loss: 0.0436 - val_accuracy: 0.5500 - val_loss: 3.0983
       Epoch 167/200
      25/25
                                 5s 207ms/step - accuracy: 0.9789 - loss: 0.0917 - val_accuracy: 0.5500 - val_loss: 3.3727
       Epoch 168/200
      25/25
                                 • 5s 197ms/step - accuracy: 0.9814 - loss: 0.0524 - val_accuracy: 0.5650 - val_loss: 3.5163
       Epoch 169/200
      25/25
                                 - 5s 191ms/step - accuracy: 0.9922 - loss: 0.0449 - val_accuracy: 0.5700 - val_loss: 3.3930
      Epoch 170/200
      25/25
                                 - 5s 188ms/step - accuracy: 0.9858 - loss: 0.0606 - val_accuracy: 0.5400 - val_loss: 3.3500
       Epoch 171/200
      25/25
                                 5s 187ms/step - accuracy: 0.9942 - loss: 0.0179 - val_accuracy: 0.5700 - val_loss: 3.6854
       Epoch 172/200
       25/25
                                 • 5s 202ms/step - accuracy: 0.9836 - loss: 0.0725 - val_accuracy: 0.6050 - val_loss: 3.5931
       Epoch 173/200
                                 5s 194ms/step - accuracy: 0.9918 - loss: 0.0542 - val_accuracy: 0.5700 - val_loss: 3.5743
      25/25
      Epoch 174/200
       25/25
                                 - 5s 194ms/step - accuracy: 0.9908 - loss: 0.0405 - val_accuracy: 0.5850 - val_loss: 3.6747
      Epoch 175/200
       25/25
                                 5s 193ms/step - accuracy: 0.9892 - loss: 0.0261 - val_accuracy: 0.6050 - val_loss: 3.7958
       Epoch 176/200
      25/25
                                 5s 189ms/step - accuracy: 0.9804 - loss: 0.0539 - val_accuracy: 0.5300 - val_loss: 3.9636
       Epoch 177/200
                                 - 5s 195ms/step - accuracy: 0.9904 - loss: 0.0310 - val_accuracy: 0.6100 - val_loss: 3.5858
       25/25
       Epoch 178/200
      25/25
                                 - 5s 192ms/step - accuracy: 0.9821 - loss: 0.0586 - val_accuracy: 0.6000 - val_loss: 3.3208
      Epoch 179/200
       25/25
                                 • 5s 190ms/step - accuracy: 0.9860 - loss: 0.0452 - val_accuracy: 0.5850 - val_loss: 3.4362
       Epoch 180/200
      25/25
                                - 5s 193ms/step - accuracy: 0.9893 - loss: 0.0315 - val_accuracy: 0.6050 - val_loss: 3.4990
       Epoch 181/200
      25/25
                                 - 5s 192ms/step - accuracy: 0.9871 - loss: 0.0347 - val_accuracy: 0.5950 - val_loss: 3.3163
       Epoch 182/200
      25/25 •
                                 - 5s 188ms/step - accuracy: 0.9882 - loss: 0.0513 - val_accuracy: 0.5800 - val_loss: 3.4056
       Epoch 183/200
       25/25
                                 5s 185ms/step - accuracy: 0.9828 - loss: 0.0370 - val_accuracy: 0.5450 - val_loss: 3.7495
       Epoch 184/200
      25/25
                                 • 5s 189ms/step - accuracy: 0.9824 - loss: 0.0682 - val_accuracy: 0.6000 - val_loss: 3.6786
       Epoch 185/200
                                 - 5s 192ms/step - accuracy: 0.9805 - loss: 0.0477 - val_accuracy: 0.5550 - val_loss: 3.7180
       25/25
       Epoch 186/200
      25/25
                                 - 5s 199ms/step - accuracy: 0.9865 - loss: 0.0487 - val_accuracy: 0.5350 - val_loss: 3.4560
       Epoch 187/200
       25/25 -
                                  5s 200ms/step - accuracy: 0.9870 - loss: 0.0400 - val_accuracy: 0.5700 - val_loss: 3.4295
       Epoch 188/200
      25/25
                                 - 5s 194ms/step - accuracy: 0.9863 - loss: 0.0447 - val_accuracy: 0.5650 - val_loss: 3.5240
       Epoch 189/200
      25/25
                                 - 5s 195ms/step - accuracy: 0.9796 - loss: 0.0755 - val_accuracy: 0.5950 - val_loss: 3.6435
      Epoch 190/200
      25/25
                                 - 5s 193ms/step - accuracy: 0.9626 - loss: 0.1290 - val_accuracy: 0.5600 - val_loss: 3.2255
      Epoch 191/200
       25/25
                                 5s 189ms/step - accuracy: 0.9832 - loss: 0.0353 - val_accuracy: 0.5250 - val_loss: 3.6364
       Epoch 192/200
      25/25
                                 5s 196ms/step - accuracy: 0.9832 - loss: 0.0420 - val_accuracy: 0.5600 - val_loss: 3.4615
       Epoch 193/200
       25/25
                                 - 5s 192ms/step - accuracy: 0.9907 - loss: 0.0301 - val_accuracy: 0.5600 - val_loss: 4.0156
       Epoch 194/200
      25/25
                                 5s 194ms/step - accuracy: 0.9744 - loss: 0.0831 - val_accuracy: 0.5800 - val_loss: 3.4824
      Epoch 195/200
       25/25
                                 - 5s 193ms/step - accuracy: 0.9933 - loss: 0.0319 - val_accuracy: 0.5500 - val_loss: 4.6713
      Epoch 196/200
      25/25
                                 • 5s 196ms/step - accuracy: 0.9767 - loss: 0.1003 - val_accuracy: 0.5400 - val_loss: 5.9740
       Epoch 197/200
      25/25
                                 • 5s 193ms/step - accuracy: 0.9628 - loss: 0.2369 - val_accuracy: 0.5600 - val_loss: 3.9092
       Epoch 198/200
                                 5s 192ms/step - accuracy: 0.9807 - loss: 0.0682 - val_accuracy: 0.5750 - val_loss: 4.1215
       25/25
       Epoch 199/200
       25/25
                                 5s 190ms/step - accuracy: 0.9963 - loss: 0.0230 - val_accuracy: 0.5950 - val_loss: 3.8465
       Epoch 200/200
       25/25
                                 5s 189ms/step - accuracy: 0.9853 - loss: 0.0466 - val accuracy: 0.5800 - val loss: 3.8862
In [8]: print(f"Execution time: {elapsed_time:.2f} seconds")
       Execution time: 967.01 seconds
```

In [9]: def append_core_data(score_path, num_cores, elapsed_time):
 # Check if the file already exists

Check if the file already exists
file_exists = os.path.exists(score_path)

Open the file in append mode
with open(score_path, mode='a', newline='') as file:

```
writer = csv.writer(file)

# If the file is new, write the header
if not file_exists:
    writer.writerow(["Number of Cores", "Elapsed Time"])

# Write the new data
writer.writerow([num_cores, elapsed_time])
```

```
In [1]: %matplotlib inline
        import shutil
        import random
        import numpy as np
        from warnings import filterwarnings
        filterwarnings('ignore')
        from tensorflow.keras import layers, regularizers, optimizers
        from tensorflow.keras import models
        from tensorflow.keras.models import Sequential, Model
        from tensorflow.keras.layers import LeakyReLU,Dense, Activation, Flatten, Dropout, BatchNormalization,Conv2D, MaxPooling2D
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import tensorflow as tf
        import os
        import time
        import csv
        from matplotlib import figure
```

Define 6 worker

```
In [2]: # Set the number of threads
number_of_worker = 6
os.environ['OMP_NUM_THREADS'] = '6' # OpenMP threads for parallelism
os.environ['TF_NUM_INTEROP_THREADS'] = '6' # Threads for inter-operation parallelism
os.environ['TF_NUM_INTRAOP_THREADS'] = '6' # Threads for intra-operation parallelism
# Confirm TensorFlow is using the specified number of threads
tf.config.threading.set_inter_op_parallelism_threads(number_of_worker)
tf.config.threading.set_intra_op_parallelism_threads(number_of_worker)
```

```
In [3]: source_dir = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DATA\Convert_Audio_File_to_jpg_file"
        target_dir = r'genres_train_val_split_data'
        split_ratio = 0.8
        def Train_Test_Split(source_dir,target_dir,split_ratio):
            # Define source and target directories
            train_dir = os.path.join(target_dir, 'train')
            val_dir = os.path.join(target_dir, 'val')
            # Create target directories if they don't exist
            os.makedirs(train_dir, exist_ok=True)
            os.makedirs(val_dir, exist_ok=True)
            # Get the list of class directories
            classes = [d for d in os.listdir(source_dir) if os.path.isdir(os.path.join(source_dir, d))]
            for class name in classes:
                # Create class directories in train and val folders
                os.makedirs(os.path.join(train_dir, class_name), exist_ok=True)
                os.makedirs(os.path.join(val_dir, class_name), exist_ok=True)
                # Get list of images in the class directory
                class_dir = os.path.join(source_dir, class_name)
                images = [f for f in os.listdir(class_dir) if os.path.isfile(os.path.join(class_dir, f))]
                # Shuffle the images
                random.shuffle(images)
                # Compute the split point
                split_point = int(len(images) * split_ratio)
                # Split the images into training and validation sets
                train images = images[:split point]
                val_images = images[split_point:]
```

Data split completed successfully!

Load the Data

```
In [5]: WIDTH = 64
        HEIGHT = 64
        BATCH_SIZE = 32
        TRAIN_DIR=r'genres_train_val_split_data/train'
        val_dir = r'genres_train_val_split_data/val'
        # data prep
        train_datagen = ImageDataGenerator(
            rescale=1./255.,validation_split=0.25)
        train_generator = train_datagen.flow_from_directory(
            TRAIN_DIR,
            target size=(HEIGHT, WIDTH),
                batch_size=BATCH_SIZE,
                class_mode='categorical')
        validation_gen = train_datagen.flow_from_directory(
            val_dir,target_size = (HEIGHT,WIDTH),
            batch_size = BATCH_SIZE,
            class_mode = 'categorical'
```

Found 800 images belonging to 10 classes. Found 200 images belonging to 10 classes.

Model Architecture

```
In [6]: model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                         input_shape=(64,64,3)))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(128, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(128, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(10, activation='softmax'))
        model.compile(optimizers.RMSprop(learning_rate=0.0005, decay=1e-6),loss="categorical_crossentropy",metrics=["accuracy"])
        model.summary()
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 64, 64, 32)	896
activation (Activation)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 62, 62, 64)	18,496
activation_1 (Activation)	(None, 62, 62, 64)	0
max_pooling2d (MaxPooling2D)	(None, 31, 31, 64)	0
dropout (Dropout)	(None, 31, 31, 64)	0
conv2d_2 (Conv2D)	(None, 31, 31, 64)	36,928
activation_2 (Activation)	(None, 31, 31, 64)	0
conv2d_3 (Conv2D)	(None, 29, 29, 64)	36,928
activation_3 (Activation)	(None, 29, 29, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_1 (Dropout)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 14, 14, 128)	73,856
activation_4 (Activation)	(None, 14, 14, 128)	0
conv2d_5 (Conv2D)	(None, 12, 12, 128)	147,584
activation_5 (Activation)	(None, 12, 12, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_2 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2,359,808
activation_6 (Activation)	(None, 512)	0
dropout_3 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5,130

```
In [7]: STEP_SIZE_TRAIN=train_generator.n//train_generator.batch_size
# Measure the execution time
start_time = time.time()

model.fit(train_generator,validation_data=validation_gen,epochs=200)

end_time = time.time()
elapsed_time = end_time - start_time
```

```
Epoch 1/200
                           5s 157ms/step - accuracy: 0.0890 - loss: 2.3322 - val_accuracy: 0.1000 - val_loss: 2.3011
25/25
Epoch 2/200
25/25 -
                          4s 154ms/step - accuracy: 0.1340 - loss: 2.3001 - val_accuracy: 0.1550 - val_loss: 2.2372
Epoch 3/200
25/25
                          4s 162ms/step - accuracy: 0.1880 - loss: 2.2027 - val_accuracy: 0.2400 - val_loss: 2.0769
Epoch 4/200
                          4s 160ms/step - accuracy: 0.2299 - loss: 2.1084 - val_accuracy: 0.2600 - val_loss: 2.0161
25/25
Epoch 5/200
25/25
                          - 4s 167ms/step - accuracy: 0.2161 - loss: 2.0444 - val_accuracy: 0.2450 - val_loss: 1.9046
Epoch 6/200
25/25
                          4s 160ms/step - accuracy: 0.2236 - loss: 2.0755 - val_accuracy: 0.2700 - val_loss: 1.9300
Epoch 7/200
25/25
                          4s 157ms/step - accuracy: 0.2998 - loss: 1.9565 - val_accuracy: 0.2900 - val_loss: 1.9200
Epoch 8/200
25/25
                          4s 163ms/step - accuracy: 0.2854 - loss: 1.9678 - val_accuracy: 0.2200 - val_loss: 2.0452
Epoch 9/200
                          4s 161ms/step - accuracy: 0.3138 - loss: 1.9735 - val_accuracy: 0.3500 - val_loss: 1.8268
25/25
Epoch 10/200
25/25
                          4s 166ms/step - accuracy: 0.3174 - loss: 1.8635 - val_accuracy: 0.3950 - val_loss: 1.7900
Epoch 11/200
25/25
                          4s 166ms/step - accuracy: 0.3454 - loss: 1.8097 - val_accuracy: 0.4200 - val_loss: 1.6308
Epoch 12/200
25/25
                          4s 167ms/step - accuracy: 0.3527 - loss: 1.7176 - val_accuracy: 0.4450 - val_loss: 1.5843
Epoch 13/200
                          4s 164ms/step - accuracy: 0.3828 - loss: 1.6564 - val_accuracy: 0.4500 - val_loss: 1.5476
25/25
Epoch 14/200
25/25
                          4s 167ms/step - accuracy: 0.3882 - loss: 1.6838 - val_accuracy: 0.4000 - val_loss: 1.6897
Epoch 15/200
                          4s 168ms/step - accuracy: 0.3887 - loss: 1.6812 - val_accuracy: 0.4350 - val_loss: 1.5513
25/25
Epoch 16/200
25/25
                          - 4s 169ms/step - accuracy: 0.4168 - loss: 1.6188 - val_accuracy: 0.3950 - val_loss: 1.5964
Epoch 17/200
25/25
                          4s 164ms/step - accuracy: 0.3895 - loss: 1.6571 - val_accuracy: 0.4650 - val_loss: 1.5162
Epoch 18/200
25/25
                          4s 170ms/step - accuracy: 0.4429 - loss: 1.5384 - val_accuracy: 0.5150 - val_loss: 1.4634
Epoch 19/200
25/25
                          4s 165ms/step - accuracy: 0.4582 - loss: 1.4640 - val_accuracy: 0.4400 - val_loss: 1.4517
Epoch 20/200
25/25
                          4s 168ms/step - accuracy: 0.4804 - loss: 1.4429 - val_accuracy: 0.4650 - val_loss: 1.4854
Epoch 21/200
25/25
                          4s 170ms/step - accuracy: 0.4517 - loss: 1.4612 - val_accuracy: 0.5150 - val_loss: 1.4165
Epoch 22/200
25/25
                          4s 170ms/step - accuracy: 0.4987 - loss: 1.3627 - val_accuracy: 0.4500 - val_loss: 1.4091
Epoch 23/200
25/25
                          4s 165ms/step - accuracy: 0.4607 - loss: 1.4274 - val_accuracy: 0.5600 - val_loss: 1.2584
Epoch 24/200
25/25
                          4s 163ms/step - accuracy: 0.5193 - loss: 1.3213 - val_accuracy: 0.5650 - val_loss: 1.2483
Epoch 25/200
                          4s 163ms/step - accuracy: 0.4835 - loss: 1.3475 - val_accuracy: 0.5500 - val_loss: 1.2676
25/25
Epoch 26/200
25/25
                          • 4s 165ms/step - accuracy: 0.5028 - loss: 1.3596 - val_accuracy: 0.5300 - val_loss: 1.2933
Epoch 27/200
25/25
                          4s 168ms/step - accuracy: 0.5224 - loss: 1.3174 - val_accuracy: 0.5300 - val_loss: 1.2774
Epoch 28/200
25/25
                          4s 168ms/step - accuracy: 0.5612 - loss: 1.2203 - val_accuracy: 0.5550 - val_loss: 1.2307
Epoch 29/200
25/25 -
                          4s 166ms/step - accuracy: 0.5548 - loss: 1.1674 - val_accuracy: 0.5950 - val_loss: 1.2334
Epoch 30/200
25/25
                          4s 167ms/step - accuracy: 0.5841 - loss: 1.1709 - val_accuracy: 0.5500 - val_loss: 1.1717
Epoch 31/200
25/25
                          4s 165ms/step - accuracy: 0.6043 - loss: 1.0935 - val_accuracy: 0.5400 - val_loss: 1.2270
Epoch 32/200
25/25
                          4s 169ms/step - accuracy: 0.5900 - loss: 1.1565 - val accuracy: 0.6050 - val loss: 1.1707
Epoch 33/200
25/25
                          4s 166ms/step - accuracy: 0.6057 - loss: 1.0924 - val_accuracy: 0.5700 - val_loss: 1.1845
Epoch 34/200
                          4s 166ms/step - accuracy: 0.6117 - loss: 1.0508 - val_accuracy: 0.5650 - val_loss: 1.2611
25/25
Epoch 35/200
25/25
                          4s 162ms/step - accuracy: 0.6288 - loss: 1.0134 - val_accuracy: 0.6050 - val_loss: 1.1702
Epoch 36/200
25/25
                          4s 167ms/step - accuracy: 0.6574 - loss: 1.0158 - val accuracy: 0.5050 - val loss: 1.3790
Epoch 37/200
25/25
                          - 4s 173ms/step - accuracy: 0.6466 - loss: 0.9846 - val_accuracy: 0.4950 - val_loss: 1.3185
Epoch 38/200
25/25
                          4s 168ms/step - accuracy: 0.6530 - loss: 0.9448 - val_accuracy: 0.6300 - val_loss: 1.1059
Epoch 39/200
25/25
                           4s 169ms/step - accuracy: 0.6636 - loss: 0.9406 - val_accuracy: 0.5450 - val_loss: 1.3051
Epoch 40/200
25/25
                          4s 171ms/step - accuracy: 0.7135 - loss: 0.7819 - val_accuracy: 0.5100 - val_loss: 1.4909
Epoch 41/200
25/25
                          4s 171ms/step - accuracy: 0.6594 - loss: 0.9155 - val_accuracy: 0.6400 - val_loss: 1.0680
```

```
Epoch 42/200
25/25
                          4s 165ms/step - accuracy: 0.7064 - loss: 0.8548 - val_accuracy: 0.6250 - val_loss: 1.1288
Epoch 43/200
25/25
                          4s 166ms/step - accuracy: 0.7434 - loss: 0.7080 - val_accuracy: 0.6400 - val_loss: 1.0650
Epoch 44/200
25/25
                          4s 168ms/step - accuracy: 0.7365 - loss: 0.7392 - val_accuracy: 0.6400 - val_loss: 1.2224
Epoch 45/200
                          4s 165ms/step - accuracy: 0.7206 - loss: 0.7465 - val_accuracy: 0.6050 - val_loss: 1.1413
25/25
Epoch 46/200
25/25
                          • 4s 167ms/step - accuracy: 0.7372 - loss: 0.7719 - val_accuracy: 0.6600 - val_loss: 1.0741
Epoch 47/200
25/25
                          4s 166ms/step - accuracy: 0.7668 - loss: 0.6776 - val_accuracy: 0.6050 - val_loss: 1.2668
Epoch 48/200
25/25
                          4s 172ms/step - accuracy: 0.7406 - loss: 0.6342 - val_accuracy: 0.6450 - val_loss: 1.0855
Epoch 49/200
25/25
                          4s 167ms/step - accuracy: 0.7495 - loss: 0.6687 - val_accuracy: 0.6200 - val_loss: 1.1426
Epoch 50/200
                          4s 174ms/step - accuracy: 0.7787 - loss: 0.5902 - val_accuracy: 0.5900 - val_loss: 1.3110
25/25
Epoch 51/200
25/25
                          4s 162ms/step - accuracy: 0.7930 - loss: 0.6049 - val_accuracy: 0.6400 - val_loss: 1.2216
Epoch 52/200
25/25
                          4s 167ms/step - accuracy: 0.8118 - loss: 0.5545 - val_accuracy: 0.6900 - val_loss: 1.1773
Epoch 53/200
25/25
                          4s 162ms/step - accuracy: 0.7822 - loss: 0.5856 - val_accuracy: 0.6450 - val_loss: 1.1829
Epoch 54/200
25/25
                          4s 169ms/step - accuracy: 0.7902 - loss: 0.5876 - val_accuracy: 0.6500 - val_loss: 1.1793
Epoch 55/200
25/25
                          4s 172ms/step - accuracy: 0.7871 - loss: 0.5644 - val_accuracy: 0.5750 - val_loss: 1.3634
Epoch 56/200
                          4s 174ms/step - accuracy: 0.8249 - loss: 0.4743 - val_accuracy: 0.6400 - val_loss: 1.3440
25/25
Epoch 57/200
25/25
                          - 4s 167ms/step - accuracy: 0.7946 - loss: 0.5888 - val_accuracy: 0.6550 - val_loss: 1.2579
Epoch 58/200
25/25
                          4s 169ms/step - accuracy: 0.8377 - loss: 0.4342 - val_accuracy: 0.6600 - val_loss: 1.2937
Epoch 59/200
25/25
                          4s 167ms/step - accuracy: 0.8479 - loss: 0.4293 - val_accuracy: 0.6600 - val_loss: 1.2260
Epoch 60/200
25/25
                          4s 169ms/step - accuracy: 0.8441 - loss: 0.4288 - val_accuracy: 0.6800 - val_loss: 1.1514
Epoch 61/200
25/25
                          4s 172ms/step - accuracy: 0.8759 - loss: 0.3626 - val_accuracy: 0.6300 - val_loss: 1.5372
Epoch 62/200
25/25
                          4s 166ms/step - accuracy: 0.8100 - loss: 0.5374 - val_accuracy: 0.6400 - val_loss: 1.2935
Epoch 63/200
25/25
                          4s 170ms/step - accuracy: 0.8421 - loss: 0.4204 - val_accuracy: 0.6200 - val_loss: 1.5336
Epoch 64/200
25/25
                           4s 170ms/step - accuracy: 0.8930 - loss: 0.2955 - val_accuracy: 0.6700 - val_loss: 1.3578
Epoch 65/200
25/25
                          4s 167ms/step - accuracy: 0.8889 - loss: 0.3113 - val_accuracy: 0.6150 - val_loss: 1.3991
Epoch 66/200
                          4s 172ms/step - accuracy: 0.8777 - loss: 0.3702 - val_accuracy: 0.6750 - val_loss: 1.3092
25/25
Epoch 67/200
25/25
                          • 4s 170ms/step - accuracy: 0.8840 - loss: 0.3203 - val_accuracy: 0.6650 - val_loss: 1.3457
Epoch 68/200
25/25
                          4s 171ms/step - accuracy: 0.8963 - loss: 0.2672 - val_accuracy: 0.6700 - val_loss: 1.3466
Epoch 69/200
25/25
                          4s 174ms/step - accuracy: 0.8528 - loss: 0.4241 - val_accuracy: 0.6050 - val_loss: 1.6286
Epoch 70/200
25/25 -
                          4s 170ms/step - accuracy: 0.9103 - loss: 0.2695 - val_accuracy: 0.6650 - val_loss: 1.4309
Epoch 71/200
25/25
                          4s 170ms/step - accuracy: 0.9254 - loss: 0.2268 - val_accuracy: 0.6350 - val_loss: 1.4934
Epoch 72/200
25/25
                           4s 169ms/step - accuracy: 0.9097 - loss: 0.2662 - val_accuracy: 0.6550 - val_loss: 1.5303
Epoch 73/200
25/25
                          4s 162ms/step - accuracy: 0.8869 - loss: 0.2817 - val accuracy: 0.6650 - val loss: 1.4049
Epoch 74/200
25/25
                          4s 172ms/step - accuracy: 0.9100 - loss: 0.2291 - val_accuracy: 0.6550 - val_loss: 1.3215
Epoch 75/200
                          4s 169ms/step - accuracy: 0.9192 - loss: 0.2422 - val_accuracy: 0.6750 - val_loss: 1.2551
25/25
Epoch 76/200
25/25
                          4s 162ms/step - accuracy: 0.9105 - loss: 0.2636 - val_accuracy: 0.6300 - val_loss: 1.5186
Epoch 77/200
25/25
                          4s 166ms/step - accuracy: 0.9350 - loss: 0.1721 - val accuracy: 0.5850 - val loss: 1.7975
Epoch 78/200
25/25
                          - 4s 170ms/step - accuracy: 0.9183 - loss: 0.2491 - val_accuracy: 0.6500 - val_loss: 1.4510
Epoch 79/200
25/25
                          4s 159ms/step - accuracy: 0.9141 - loss: 0.2119 - val_accuracy: 0.6500 - val_loss: 1.5305
Epoch 80/200
25/25
                           4s 168ms/step - accuracy: 0.9163 - loss: 0.2393 - val_accuracy: 0.6850 - val_loss: 1.3357
Epoch 81/200
25/25
                          4s 159ms/step - accuracy: 0.9439 - loss: 0.1721 - val_accuracy: 0.6800 - val_loss: 1.5198
Epoch 82/200
25/25
                          4s 166ms/step - accuracy: 0.9051 - loss: 0.2558 - val_accuracy: 0.6450 - val_loss: 1.4418
```

```
Epoch 83/200
                          4s 172ms/step - accuracy: 0.9331 - loss: 0.1749 - val_accuracy: 0.6750 - val_loss: 1.4238
25/25
Epoch 84/200
25/25
                          4s 170ms/step - accuracy: 0.9438 - loss: 0.1506 - val_accuracy: 0.6850 - val_loss: 1.3142
Epoch 85/200
25/25
                          4s 171ms/step - accuracy: 0.9479 - loss: 0.1356 - val_accuracy: 0.6800 - val_loss: 1.5173
Epoch 86/200
25/25
                          4s 171ms/step - accuracy: 0.9248 - loss: 0.2362 - val_accuracy: 0.6600 - val_loss: 1.7096
Epoch 87/200
25/25
                          4s 172ms/step - accuracy: 0.9312 - loss: 0.1948 - val_accuracy: 0.6550 - val_loss: 1.7363
Epoch 88/200
25/25
                          4s 169ms/step - accuracy: 0.9397 - loss: 0.1930 - val_accuracy: 0.6350 - val_loss: 1.7357
Epoch 89/200
25/25
                          4s 171ms/step - accuracy: 0.9558 - loss: 0.1363 - val_accuracy: 0.6750 - val_loss: 1.4225
Epoch 90/200
25/25
                          4s 169ms/step - accuracy: 0.9226 - loss: 0.2255 - val_accuracy: 0.6750 - val_loss: 1.5259
Epoch 91/200
                          4s 173ms/step - accuracy: 0.9393 - loss: 0.1695 - val_accuracy: 0.6550 - val_loss: 1.8268
25/25
Epoch 92/200
25/25
                          4s 161ms/step - accuracy: 0.9457 - loss: 0.1703 - val_accuracy: 0.6550 - val_loss: 1.9608
Epoch 93/200
25/25
                          4s 167ms/step - accuracy: 0.9414 - loss: 0.1481 - val_accuracy: 0.6550 - val_loss: 1.5784
Epoch 94/200
25/25
                          4s 170ms/step - accuracy: 0.9607 - loss: 0.1032 - val_accuracy: 0.6550 - val_loss: 1.7475
Epoch 95/200
25/25
                          4s 165ms/step - accuracy: 0.9778 - loss: 0.1090 - val_accuracy: 0.6700 - val_loss: 1.4802
Epoch 96/200
25/25
                          4s 172ms/step - accuracy: 0.9754 - loss: 0.1131 - val_accuracy: 0.6850 - val_loss: 1.4780
Epoch 97/200
25/25
                          4s 171ms/step - accuracy: 0.9543 - loss: 0.1551 - val_accuracy: 0.4850 - val_loss: 3.1356
Epoch 98/200
25/25
                          4s 163ms/step - accuracy: 0.9005 - loss: 0.3302 - val_accuracy: 0.6600 - val_loss: 1.5196
Epoch 99/200
25/25
                          4s 164ms/step - accuracy: 0.9756 - loss: 0.0736 - val_accuracy: 0.6750 - val_loss: 1.6067
Epoch 100/200
25/25
                          4s 170ms/step - accuracy: 0.9593 - loss: 0.1224 - val_accuracy: 0.6150 - val_loss: 2.1983
Epoch 101/200
25/25
                          4s 173ms/step - accuracy: 0.9357 - loss: 0.2089 - val_accuracy: 0.6800 - val_loss: 1.7204
Epoch 102/200
25/25
                          4s 173ms/step - accuracy: 0.9600 - loss: 0.1106 - val_accuracy: 0.6600 - val_loss: 2.1862
Epoch 103/200
25/25
                          4s 173ms/step - accuracy: 0.9742 - loss: 0.1175 - val_accuracy: 0.6700 - val_loss: 1.7248
Epoch 104/200
25/25
                          4s 171ms/step - accuracy: 0.9673 - loss: 0.1144 - val_accuracy: 0.6800 - val_loss: 1.7310
Epoch 105/200
25/25
                          4s 170ms/step - accuracy: 0.9703 - loss: 0.0887 - val_accuracy: 0.6900 - val_loss: 1.7657
Epoch 106/200
25/25
                          4s 170ms/step - accuracy: 0.9544 - loss: 0.1503 - val_accuracy: 0.6500 - val_loss: 1.8447
Epoch 107/200
                          4s 171ms/step - accuracy: 0.9348 - loss: 0.1630 - val_accuracy: 0.6800 - val_loss: 1.7305
25/25
Epoch 108/200
25/25
                          • 4s 178ms/step - accuracy: 0.9695 - loss: 0.1021 - val_accuracy: 0.6550 - val_loss: 2.1379
Epoch 109/200
25/25
                          4s 161ms/step - accuracy: 0.9433 - loss: 0.1890 - val_accuracy: 0.6500 - val_loss: 1.6394
Epoch 110/200
25/25
                          4s 166ms/step - accuracy: 0.9777 - loss: 0.1011 - val_accuracy: 0.6550 - val_loss: 1.5713
Epoch 111/200
25/25
                          4s 171ms/step - accuracy: 0.9575 - loss: 0.1313 - val_accuracy: 0.6700 - val_loss: 1.6176
Epoch 112/200
25/25
                          4s 169ms/step - accuracy: 0.9760 - loss: 0.0796 - val_accuracy: 0.6400 - val_loss: 2.2218
Epoch 113/200
25/25
                          4s 155ms/step - accuracy: 0.9694 - loss: 0.0870 - val_accuracy: 0.6450 - val_loss: 2.2486
Epoch 114/200
25/25
                          4s 160ms/step - accuracy: 0.9743 - loss: 0.0628 - val accuracy: 0.5700 - val loss: 3.4683
Epoch 115/200
25/25
                          4s 169ms/step - accuracy: 0.9345 - loss: 0.2422 - val_accuracy: 0.6000 - val_loss: 2.5655
Epoch 116/200
25/25
                          4s 168ms/step - accuracy: 0.9648 - loss: 0.1994 - val_accuracy: 0.6300 - val_loss: 2.0663
Epoch 117/200
25/25
                          4s 174ms/step - accuracy: 0.9458 - loss: 0.1140 - val_accuracy: 0.6750 - val_loss: 1.9091
Epoch 118/200
25/25
                          4s 174ms/step - accuracy: 0.9762 - loss: 0.0969 - val accuracy: 0.7050 - val loss: 1.7098
Epoch 119/200
25/25
                          - 4s 169ms/step - accuracy: 0.9849 - loss: 0.0512 - val_accuracy: 0.6600 - val_loss: 2.0054
Epoch 120/200
25/25
                          4s 169ms/step - accuracy: 0.9551 - loss: 0.1360 - val_accuracy: 0.6650 - val_loss: 1.9079
Epoch 121/200
25/25
                           4s 170ms/step - accuracy: 0.9772 - loss: 0.0688 - val_accuracy: 0.6650 - val_loss: 1.8168
Epoch 122/200
25/25
                          4s 169ms/step - accuracy: 0.9620 - loss: 0.1115 - val_accuracy: 0.6750 - val_loss: 1.8947
Epoch 123/200
25/25
                          4s 179ms/step - accuracy: 0.9783 - loss: 0.0906 - val_accuracy: 0.6550 - val_loss: 1.8924
```

```
Epoch 124/200
                          4s 163ms/step - accuracy: 0.9669 - loss: 0.1133 - val_accuracy: 0.6750 - val_loss: 1.7190
25/25
Epoch 125/200
25/25
                          5s 180ms/step - accuracy: 0.9892 - loss: 0.0311 - val_accuracy: 0.6550 - val_loss: 2.1296
Epoch 126/200
25/25
                          4s 169ms/step - accuracy: 0.9659 - loss: 0.0754 - val_accuracy: 0.6700 - val_loss: 1.9735
Epoch 127/200
25/25
                          4s 177ms/step - accuracy: 0.9602 - loss: 0.1226 - val_accuracy: 0.6650 - val_loss: 2.0814
Epoch 128/200
25/25
                          • 4s 173ms/step - accuracy: 0.9858 - loss: 0.0539 - val_accuracy: 0.6800 - val_loss: 2.1406
Epoch 129/200
25/25
                          4s 169ms/step - accuracy: 0.9681 - loss: 0.1102 - val_accuracy: 0.6750 - val_loss: 1.9733
Epoch 130/200
25/25
                          4s 174ms/step - accuracy: 0.9774 - loss: 0.0680 - val_accuracy: 0.6700 - val_loss: 1.7611
Epoch 131/200
25/25
                          4s 174ms/step - accuracy: 0.9811 - loss: 0.0621 - val_accuracy: 0.6500 - val_loss: 1.8567
Epoch 132/200
                          4s 168ms/step - accuracy: 0.9886 - loss: 0.0347 - val_accuracy: 0.6050 - val_loss: 2.9189
25/25
Epoch 133/200
25/25
                          4s 171ms/step - accuracy: 0.9355 - loss: 0.3112 - val_accuracy: 0.6600 - val_loss: 1.9797
Epoch 134/200
25/25
                          4s 172ms/step - accuracy: 0.9788 - loss: 0.0517 - val_accuracy: 0.6500 - val_loss: 2.3874
Epoch 135/200
25/25
                          4s 173ms/step - accuracy: 0.9656 - loss: 0.0846 - val_accuracy: 0.6650 - val_loss: 2.0185
Epoch 136/200
                          4s 171ms/step - accuracy: 0.9800 - loss: 0.0580 - val_accuracy: 0.6900 - val_loss: 1.9009
25/25
Epoch 137/200
25/25
                          4s 159ms/step - accuracy: 0.9758 - loss: 0.0743 - val_accuracy: 0.6350 - val_loss: 2.5351
Epoch 138/200
                          4s 161ms/step - accuracy: 0.9412 - loss: 0.1937 - val_accuracy: 0.6750 - val_loss: 2.3021
25/25
Epoch 139/200
25/25
                          - 4s 172ms/step - accuracy: 0.9578 - loss: 0.0986 - val_accuracy: 0.6600 - val_loss: 2.4281
Epoch 140/200
25/25
                          4s 171ms/step - accuracy: 0.9918 - loss: 0.0322 - val_accuracy: 0.6400 - val_loss: 2.8026
Epoch 141/200
25/25
                          4s 167ms/step - accuracy: 0.9650 - loss: 0.0813 - val_accuracy: 0.6900 - val_loss: 2.1600
Epoch 142/200
25/25
                          4s 173ms/step - accuracy: 0.9815 - loss: 0.0500 - val_accuracy: 0.6800 - val_loss: 2.1603
Epoch 143/200
25/25
                          4s 173ms/step - accuracy: 0.9883 - loss: 0.0346 - val_accuracy: 0.6850 - val_loss: 2.0507
Epoch 144/200
25/25
                          4s 169ms/step - accuracy: 0.9832 - loss: 0.0488 - val_accuracy: 0.6600 - val_loss: 2.2622
Epoch 145/200
25/25
                          4s 171ms/step - accuracy: 0.9690 - loss: 0.1287 - val_accuracy: 0.6600 - val_loss: 2.0851
Epoch 146/200
25/25
                          4s 172ms/step - accuracy: 0.9703 - loss: 0.0971 - val_accuracy: 0.6600 - val_loss: 2.2161
Epoch 147/200
25/25
                          4s 172ms/step - accuracy: 0.9647 - loss: 0.1197 - val_accuracy: 0.6900 - val_loss: 1.9852
Epoch 148/200
                          4s 170ms/step - accuracy: 0.9848 - loss: 0.0470 - val_accuracy: 0.6800 - val_loss: 2.1950
25/25
Epoch 149/200
25/25
                          • 4s 168ms/step - accuracy: 0.9646 - loss: 0.1005 - val_accuracy: 0.6750 - val_loss: 2.2800
Epoch 150/200
25/25
                          4s 171ms/step - accuracy: 0.9792 - loss: 0.0724 - val_accuracy: 0.6750 - val_loss: 2.0512
Epoch 151/200
25/25
                          4s 170ms/step - accuracy: 0.9825 - loss: 0.0804 - val accuracy: 0.6900 - val loss: 2.4662
Epoch 152/200
25/25
                          4s 171ms/step - accuracy: 0.9782 - loss: 0.0690 - val_accuracy: 0.6450 - val_loss: 2.5868
Epoch 153/200
25/25
                          4s 171ms/step - accuracy: 0.9741 - loss: 0.0586 - val_accuracy: 0.6700 - val_loss: 2.2021
Epoch 154/200
25/25
                          4s 173ms/step - accuracy: 0.9772 - loss: 0.0831 - val_accuracy: 0.6850 - val_loss: 2.1732
Epoch 155/200
25/25
                          4s 174ms/step - accuracy: 0.9813 - loss: 0.0443 - val accuracy: 0.6250 - val loss: 2.8519
Epoch 156/200
25/25
                          4s 168ms/step - accuracy: 0.9704 - loss: 0.0755 - val_accuracy: 0.6450 - val_loss: 2.3332
Epoch 157/200
25/25
                          4s 173ms/step - accuracy: 0.9592 - loss: 0.1600 - val_accuracy: 0.6850 - val_loss: 2.0250
Epoch 158/200
25/25
                          4s 172ms/step - accuracy: 0.9815 - loss: 0.0503 - val_accuracy: 0.6800 - val_loss: 2.1698
Epoch 159/200
25/25
                          4s 169ms/step - accuracy: 0.9784 - loss: 0.0438 - val accuracy: 0.7050 - val loss: 2.1671
Epoch 160/200
                          - 4s 169ms/step - accuracy: 0.9854 - loss: 0.0685 - val_accuracy: 0.6750 - val_loss: 2.2213
25/25
Epoch 161/200
25/25
                          4s 175ms/step - accuracy: 0.9817 - loss: 0.0532 - val_accuracy: 0.6800 - val_loss: 2.1180
Epoch 162/200
25/25
                           5s 179ms/step - accuracy: 0.9793 - loss: 0.0800 - val_accuracy: 0.6650 - val_loss: 2.0562
Epoch 163/200
25/25
                          4s 168ms/step - accuracy: 0.9934 - loss: 0.0224 - val_accuracy: 0.6800 - val_loss: 2.3128
Epoch 164/200
25/25
                          4s 172ms/step - accuracy: 0.9781 - loss: 0.0955 - val_accuracy: 0.6850 - val_loss: 2.1782
```

```
Epoch 165/200
                                 4s 161ms/step - accuracy: 0.9667 - loss: 0.1743 - val_accuracy: 0.6650 - val_loss: 2.3627
       25/25
       Epoch 166/200
       25/25
                                 • 4s 171ms/step - accuracy: 0.9924 - loss: 0.0248 - val_accuracy: 0.6650 - val_loss: 2.4867
       Epoch 167/200
      25/25
                                 4s 171ms/step - accuracy: 0.9799 - loss: 0.0824 - val_accuracy: 0.6800 - val_loss: 2.2316
       Epoch 168/200
      25/25
                                 • 4s 169ms/step - accuracy: 0.9718 - loss: 0.0875 - val_accuracy: 0.6750 - val_loss: 2.1804
       Epoch 169/200
      25/25
                                 - 4s 174ms/step - accuracy: 0.9887 - loss: 0.0427 - val_accuracy: 0.6750 - val_loss: 2.0417
      Epoch 170/200
      25/25
                                 - 4s 167ms/step - accuracy: 0.9892 - loss: 0.0395 - val_accuracy: 0.6750 - val_loss: 2.4242
       Epoch 171/200
      25/25
                                 4s 168ms/step - accuracy: 0.9685 - loss: 0.1369 - val_accuracy: 0.6550 - val_loss: 2.5412
       Epoch 172/200
                                 - 4s 175ms/step - accuracy: 0.9850 - loss: 0.0492 - val_accuracy: 0.7000 - val_loss: 1.9487
       25/25
       Epoch 173/200
                                 4s 170ms/step - accuracy: 0.9841 - loss: 0.0448 - val_accuracy: 0.6200 - val_loss: 2.5740
      25/25
      Epoch 174/200
       25/25
                                 - 4s 177ms/step - accuracy: 0.9827 - loss: 0.0601 - val_accuracy: 0.6850 - val_loss: 1.9880
      Epoch 175/200
       25/25
                                 4s 172ms/step - accuracy: 0.9967 - loss: 0.0140 - val_accuracy: 0.6900 - val_loss: 2.1357
       Epoch 176/200
      25/25
                                 4s 168ms/step - accuracy: 0.9829 - loss: 0.0802 - val_accuracy: 0.6800 - val_loss: 2.0864
       Epoch 177/200
                                 • 4s 167ms/step - accuracy: 0.9862 - loss: 0.0423 - val_accuracy: 0.6900 - val_loss: 2.1834
       25/25
       Epoch 178/200
      25/25
                                 4s 171ms/step - accuracy: 0.9574 - loss: 0.1224 - val_accuracy: 0.6650 - val_loss: 2.3492
      Epoch 179/200
                                 4s 172ms/step - accuracy: 0.9945 - loss: 0.0241 - val_accuracy: 0.6250 - val_loss: 3.3972
       25/25
       Epoch 180/200
      25/25
                                - 4s 171ms/step - accuracy: 0.9831 - loss: 0.0915 - val_accuracy: 0.6700 - val_loss: 3.0905
       Epoch 181/200
      25/25
                                 - 4s 162ms/step - accuracy: 0.9797 - loss: 0.0585 - val_accuracy: 0.6150 - val_loss: 3.0855
       Epoch 182/200
      25/25
                                 - 4s 166ms/step - accuracy: 0.9878 - loss: 0.0432 - val_accuracy: 0.6750 - val_loss: 2.1840
       Epoch 183/200
       25/25
                                 4s 173ms/step - accuracy: 0.9706 - loss: 0.1064 - val_accuracy: 0.6550 - val_loss: 2.7121
       Epoch 184/200
      25/25
                                 - 4s 168ms/step - accuracy: 0.9745 - loss: 0.0734 - val_accuracy: 0.6900 - val_loss: 2.2102
       Epoch 185/200
                                 - 4s 172ms/step - accuracy: 0.9871 - loss: 0.0432 - val_accuracy: 0.6950 - val_loss: 2.0016
       25/25
       Epoch 186/200
      25/25
                                 - 4s 165ms/step - accuracy: 0.9841 - loss: 0.0458 - val_accuracy: 0.6650 - val_loss: 2.3818
       Epoch 187/200
       25/25 -
                                 4s 171ms/step - accuracy: 0.9763 - loss: 0.0931 - val_accuracy: 0.6950 - val_loss: 2.1277
       Epoch 188/200
      25/25
                                 - 4s 168ms/step - accuracy: 0.9905 - loss: 0.0339 - val_accuracy: 0.6950 - val_loss: 2.1814
       Epoch 189/200
      25/25
                                 - 4s 172ms/step - accuracy: 0.9924 - loss: 0.0474 - val_accuracy: 0.6600 - val_loss: 2.8345
      Epoch 190/200
      25/25
                                 - 4s 169ms/step - accuracy: 0.9727 - loss: 0.0834 - val_accuracy: 0.6600 - val_loss: 2.1733
      Epoch 191/200
       25/25
                                 4s 171ms/step - accuracy: 0.9894 - loss: 0.0371 - val_accuracy: 0.6700 - val_loss: 2.3174
       Epoch 192/200
      25/25
                                 - 4s 170ms/step - accuracy: 0.9936 - loss: 0.0223 - val_accuracy: 0.6850 - val_loss: 2.5772
       Epoch 193/200
       25/25
                                 - 4s 160ms/step - accuracy: 0.9917 - loss: 0.0226 - val_accuracy: 0.6650 - val_loss: 2.3985
       Epoch 194/200
      25/25
                                 4s 167ms/step - accuracy: 0.9880 - loss: 0.0382 - val_accuracy: 0.6750 - val_loss: 2.5574
      Epoch 195/200
       25/25
                                 4s 170ms/step - accuracy: 0.9776 - loss: 0.1042 - val_accuracy: 0.6600 - val_loss: 2.7016
      Epoch 196/200
      25/25
                                 • 4s 175ms/step - accuracy: 0.9847 - loss: 0.0479 - val_accuracy: 0.6950 - val_loss: 2.6320
       Epoch 197/200
      25/25
                                 • 4s 170ms/step - accuracy: 0.9786 - loss: 0.0357 - val_accuracy: 0.6550 - val_loss: 2.3906
       Epoch 198/200
       25/25
                                 • 4s 174ms/step - accuracy: 0.9901 - loss: 0.0448 - val_accuracy: 0.6600 - val_loss: 2.7241
       Epoch 199/200
       25/25
                                 4s 167ms/step - accuracy: 0.9870 - loss: 0.0350 - val_accuracy: 0.6250 - val_loss: 2.9793
       Epoch 200/200
       25/25
                                 • 4s 171ms/step - accuracy: 0.9906 - loss: 0.0226 - val accuracy: 0.6500 - val loss: 2.4609
In [8]: print(f"Execution time: {elapsed_time:.2f} seconds")
```

Execution time: 850.77 seconds

```
In [9]: def append_core_data(score_path, num_cores, elapsed_time):
    # Check if the file already exists
    file_exists = os.path.exists(score_path)

# Open the file in append mode
with open(score_path, mode='a', newline='') as file:
```

```
writer = csv.writer(file)

# If the file is new, write the header
if not file_exists:
    writer.writerow(["Number of Cores", "Elapsed Time"])

# Write the new data
writer.writerow([num_cores, elapsed_time])
```

```
In [1]: %matplotlib inline
        import shutil
        import random
        import numpy as np
        from warnings import filterwarnings
        filterwarnings('ignore')
        from tensorflow.keras import layers, regularizers, optimizers
        from tensorflow.keras import models
        from tensorflow.keras.models import Sequential, Model
        from tensorflow.keras.layers import LeakyReLU,Dense, Activation, Flatten, Dropout, BatchNormalization,Conv2D, MaxPooling2D
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import tensorflow as tf
        import os
        import time
        import csv
        from matplotlib import figure
```

Define 7 worker

```
In [2]: # Set the number of threads
number_of_worker = 7
os.environ['OMP_NUM_THREADS'] = '7' # OpenMP threads for parallelism
os.environ['TF_NUM_INTEROP_THREADS'] = '7' # Threads for inter-operation parallelism
os.environ['TF_NUM_INTRAOP_THREADS'] = '7' # Threads for intra-operation parallelism
# Confirm TensorFlow is using the specified number of threads
tf.config.threading.set_inter_op_parallelism_threads(number_of_worker)
tf.config.threading.set_intra_op_parallelism_threads(number_of_worker)
```

```
In [3]: source_dir = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DATA\Convert_Audio_File_to_jpg_file"
        target_dir = r'genres_train_val_split_data'
        split_ratio = 0.8
        def Train_Test_Split(source_dir,target_dir,split_ratio):
            # Define source and target directories
            train_dir = os.path.join(target_dir, 'train')
            val_dir = os.path.join(target_dir, 'val')
            # Create target directories if they don't exist
            os.makedirs(train_dir, exist_ok=True)
            os.makedirs(val_dir, exist_ok=True)
            # Get the list of class directories
            classes = [d for d in os.listdir(source_dir) if os.path.isdir(os.path.join(source_dir, d))]
            for class name in classes:
                # Create class directories in train and val folders
                os.makedirs(os.path.join(train_dir, class_name), exist_ok=True)
                os.makedirs(os.path.join(val_dir, class_name), exist_ok=True)
                # Get list of images in the class directory
                class_dir = os.path.join(source_dir, class_name)
                images = [f for f in os.listdir(class_dir) if os.path.isfile(os.path.join(class_dir, f))]
                # Shuffle the images
                random.shuffle(images)
                # Compute the split point
                split_point = int(len(images) * split_ratio)
                # Split the images into training and validation sets
                train images = images[:split point]
                val_images = images[split_point:]
```

Data split completed successfully!

Load the Data

```
In [5]: WIDTH = 64
        HEIGHT = 64
        BATCH_SIZE = 32
        TRAIN_DIR=r'genres_train_val_split_data/train'
        val_dir = r'genres_train_val_split_data/val'
        # data prep
        train_datagen = ImageDataGenerator(
            rescale=1./255.,validation_split=0.25)
        train_generator = train_datagen.flow_from_directory(
            TRAIN_DIR,
            target size=(HEIGHT, WIDTH),
                batch_size=BATCH_SIZE,
                class_mode='categorical')
        validation_gen = train_datagen.flow_from_directory(
            val_dir,target_size = (HEIGHT,WIDTH),
            batch_size = BATCH_SIZE,
            class_mode = 'categorical'
```

Found 800 images belonging to 10 classes. Found 200 images belonging to 10 classes.

Model Architecture

```
In [6]: model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                         input_shape=(64,64,3)))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(128, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(128, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(10, activation='softmax'))
        model.compile(optimizers.RMSprop(learning_rate=0.0005, decay=1e-6),loss="categorical_crossentropy",metrics=["accuracy"])
        model.summary()
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 64, 64, 32)	896
activation (Activation)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 62, 62, 64)	18,496
activation_1 (Activation)	(None, 62, 62, 64)	0
max_pooling2d (MaxPooling2D)	(None, 31, 31, 64)	0
dropout (Dropout)	(None, 31, 31, 64)	0
conv2d_2 (Conv2D)	(None, 31, 31, 64)	36,928
activation_2 (Activation)	(None, 31, 31, 64)	0
conv2d_3 (Conv2D)	(None, 29, 29, 64)	36,928
activation_3 (Activation)	(None, 29, 29, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_1 (Dropout)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 14, 14, 128)	73,856
activation_4 (Activation)	(None, 14, 14, 128)	0
conv2d_5 (Conv2D)	(None, 12, 12, 128)	147,584
activation_5 (Activation)	(None, 12, 12, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_2 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2,359,808
activation_6 (Activation)	(None, 512)	0
dropout_3 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5,130

```
In [7]: STEP_SIZE_TRAIN=train_generator.n//train_generator.batch_size
# Measure the execution time
start_time = time.time()

model.fit(train_generator,validation_data=validation_gen,epochs=200)

end_time = time.time()
elapsed_time = end_time - start_time
```

```
Epoch 1/200
                          4s 123ms/step - accuracy: 0.0818 - loss: 2.3262 - val_accuracy: 0.1050 - val_loss: 2.2680
25/25
Epoch 2/200
25/25 -
                          4s 150ms/step - accuracy: 0.1573 - loss: 2.2889 - val_accuracy: 0.2100 - val_loss: 2.0738
Epoch 3/200
25/25
                          4s 149ms/step - accuracy: 0.1911 - loss: 2.1370 - val_accuracy: 0.2250 - val_loss: 2.1721
Epoch 4/200
                          4s 145ms/step - accuracy: 0.2274 - loss: 2.1387 - val_accuracy: 0.2350 - val_loss: 2.0582
25/25
Epoch 5/200
25/25
                          - 4s 149ms/step - accuracy: 0.2377 - loss: 2.0395 - val_accuracy: 0.3100 - val_loss: 1.9587
Epoch 6/200
25/25
                          4s 150ms/step - accuracy: 0.2534 - loss: 1.9749 - val_accuracy: 0.3600 - val_loss: 2.0144
Epoch 7/200
25/25
                          4s 153ms/step - accuracy: 0.2613 - loss: 1.9955 - val_accuracy: 0.3100 - val_loss: 1.8442
Epoch 8/200
25/25
                          4s 149ms/step - accuracy: 0.2949 - loss: 1.8974 - val_accuracy: 0.3950 - val_loss: 1.8752
Epoch 9/200
                          4s 152ms/step - accuracy: 0.3669 - loss: 1.8716 - val_accuracy: 0.4100 - val_loss: 1.8567
25/25
Epoch 10/200
25/25
                          4s 154ms/step - accuracy: 0.3370 - loss: 1.8430 - val_accuracy: 0.4150 - val_loss: 1.7267
Epoch 11/200
25/25
                          4s 148ms/step - accuracy: 0.3315 - loss: 1.8672 - val_accuracy: 0.3900 - val_loss: 1.6877
Epoch 12/200
25/25
                          4s 154ms/step - accuracy: 0.4074 - loss: 1.7262 - val_accuracy: 0.4200 - val_loss: 1.7082
Epoch 13/200
                          4s 155ms/step - accuracy: 0.3605 - loss: 1.7425 - val_accuracy: 0.4300 - val_loss: 1.6156
25/25
Epoch 14/200
25/25
                          4s 155ms/step - accuracy: 0.3964 - loss: 1.6583 - val_accuracy: 0.4500 - val_loss: 1.5646
Epoch 15/200
25/25
                          4s 150ms/step - accuracy: 0.3550 - loss: 1.7073 - val_accuracy: 0.4550 - val_loss: 1.5227
Epoch 16/200
25/25
                          - 4s 153ms/step - accuracy: 0.4600 - loss: 1.5362 - val_accuracy: 0.4300 - val_loss: 1.5923
Epoch 17/200
25/25
                          4s 151ms/step - accuracy: 0.4731 - loss: 1.5131 - val_accuracy: 0.4800 - val_loss: 1.4602
Epoch 18/200
25/25
                          4s 153ms/step - accuracy: 0.4707 - loss: 1.4784 - val_accuracy: 0.4700 - val_loss: 1.4989
Epoch 19/200
25/25
                          4s 147ms/step - accuracy: 0.5048 - loss: 1.4154 - val_accuracy: 0.4900 - val_loss: 1.4474
Epoch 20/200
25/25
                          4s 152ms/step - accuracy: 0.5050 - loss: 1.4192 - val_accuracy: 0.4300 - val_loss: 1.5239
Epoch 21/200
25/25
                          4s 154ms/step - accuracy: 0.4976 - loss: 1.4264 - val_accuracy: 0.5150 - val_loss: 1.3873
Epoch 22/200
25/25
                          4s 151ms/step - accuracy: 0.5021 - loss: 1.3814 - val_accuracy: 0.5700 - val_loss: 1.3150
Epoch 23/200
25/25
                          4s 152ms/step - accuracy: 0.5420 - loss: 1.3249 - val_accuracy: 0.5050 - val_loss: 1.3644
Epoch 24/200
25/25
                          4s 153ms/step - accuracy: 0.5389 - loss: 1.2833 - val_accuracy: 0.5650 - val_loss: 1.2784
Epoch 25/200
25/25
                          4s 150ms/step - accuracy: 0.5266 - loss: 1.2446 - val_accuracy: 0.5300 - val_loss: 1.3330
Epoch 26/200
25/25
                          • 4s 155ms/step - accuracy: 0.5608 - loss: 1.2418 - val_accuracy: 0.5250 - val_loss: 1.3601
Epoch 27/200
25/25
                          4s 154ms/step - accuracy: 0.6144 - loss: 1.1044 - val_accuracy: 0.5150 - val_loss: 1.3303
Epoch 28/200
25/25
                          4s 154ms/step - accuracy: 0.5713 - loss: 1.1832 - val_accuracy: 0.5700 - val_loss: 1.2238
Epoch 29/200
25/25 -
                          4s 152ms/step - accuracy: 0.5994 - loss: 1.0920 - val_accuracy: 0.5350 - val_loss: 1.3166
Epoch 30/200
25/25
                          4s 152ms/step - accuracy: 0.6277 - loss: 1.0387 - val_accuracy: 0.5650 - val_loss: 1.3302
Epoch 31/200
25/25
                          4s 150ms/step - accuracy: 0.6816 - loss: 0.9319 - val_accuracy: 0.5200 - val_loss: 1.3509
Epoch 32/200
25/25
                          4s 153ms/step - accuracy: 0.6372 - loss: 1.0041 - val_accuracy: 0.5250 - val_loss: 1.3733
Epoch 33/200
25/25
                          4s 157ms/step - accuracy: 0.6570 - loss: 0.9495 - val_accuracy: 0.6100 - val_loss: 1.1951
Epoch 34/200
25/25
                          4s 154ms/step - accuracy: 0.7004 - loss: 0.8949 - val_accuracy: 0.5900 - val_loss: 1.2032
Epoch 35/200
25/25
                          4s 153ms/step - accuracy: 0.7214 - loss: 0.7897 - val_accuracy: 0.5300 - val_loss: 1.3989
Epoch 36/200
25/25
                          4s 155ms/step - accuracy: 0.6723 - loss: 0.9051 - val accuracy: 0.5700 - val loss: 1.3947
Epoch 37/200
25/25
                          - 4s 157ms/step - accuracy: 0.7025 - loss: 0.8452 - val_accuracy: 0.5950 - val_loss: 1.4084
Epoch 38/200
25/25
                          4s 152ms/step - accuracy: 0.7058 - loss: 0.8214 - val_accuracy: 0.6000 - val_loss: 1.3120
Epoch 39/200
25/25
                           4s 154ms/step - accuracy: 0.7523 - loss: 0.7262 - val_accuracy: 0.6000 - val_loss: 1.3909
Epoch 40/200
25/25
                          4s 153ms/step - accuracy: 0.7442 - loss: 0.6782 - val_accuracy: 0.6300 - val_loss: 1.2302
Epoch 41/200
25/25
                          4s 154ms/step - accuracy: 0.7542 - loss: 0.6868 - val_accuracy: 0.5750 - val_loss: 1.3111
```

```
Epoch 42/200
                          4s 150ms/step - accuracy: 0.7372 - loss: 0.7429 - val_accuracy: 0.6000 - val_loss: 1.3485
25/25
Epoch 43/200
25/25
                          4s 151ms/step - accuracy: 0.7675 - loss: 0.6476 - val_accuracy: 0.5900 - val_loss: 1.5109
Epoch 44/200
25/25
                          4s 151ms/step - accuracy: 0.7961 - loss: 0.6001 - val_accuracy: 0.6050 - val_loss: 1.4317
Epoch 45/200
                          4s 154ms/step - accuracy: 0.8014 - loss: 0.5804 - val_accuracy: 0.6100 - val_loss: 1.2404
25/25
Epoch 46/200
25/25
                          • 4s 153ms/step - accuracy: 0.7928 - loss: 0.5734 - val_accuracy: 0.6250 - val_loss: 1.4807
Epoch 47/200
25/25
                          4s 152ms/step - accuracy: 0.8146 - loss: 0.5881 - val_accuracy: 0.6300 - val_loss: 1.3544
Epoch 48/200
25/25
                          4s 155ms/step - accuracy: 0.8335 - loss: 0.4861 - val_accuracy: 0.6150 - val_loss: 1.5126
Epoch 49/200
25/25
                          4s 151ms/step - accuracy: 0.8099 - loss: 0.5352 - val_accuracy: 0.5750 - val_loss: 1.4304
Epoch 50/200
                          4s 153ms/step - accuracy: 0.8612 - loss: 0.4229 - val_accuracy: 0.5750 - val_loss: 1.5456
25/25
Epoch 51/200
25/25
                          4s 150ms/step - accuracy: 0.8352 - loss: 0.5294 - val_accuracy: 0.5600 - val_loss: 1.8977
Epoch 52/200
25/25
                          4s 150ms/step - accuracy: 0.8287 - loss: 0.4984 - val_accuracy: 0.6050 - val_loss: 1.4981
Epoch 53/200
25/25
                          4s 156ms/step - accuracy: 0.8452 - loss: 0.4254 - val_accuracy: 0.6400 - val_loss: 1.5920
Epoch 54/200
25/25
                          4s 154ms/step - accuracy: 0.8350 - loss: 0.4264 - val_accuracy: 0.6150 - val_loss: 1.4441
Epoch 55/200
25/25
                          4s 157ms/step - accuracy: 0.9038 - loss: 0.3091 - val_accuracy: 0.6400 - val_loss: 1.4488
Epoch 56/200
                          4s 152ms/step - accuracy: 0.8716 - loss: 0.3435 - val_accuracy: 0.6200 - val_loss: 1.4544
25/25
Epoch 57/200
25/25
                          - 4s 156ms/step - accuracy: 0.8840 - loss: 0.3251 - val_accuracy: 0.6000 - val_loss: 1.4671
Epoch 58/200
25/25
                          4s 155ms/step - accuracy: 0.8853 - loss: 0.3201 - val_accuracy: 0.6200 - val_loss: 1.4904
Epoch 59/200
25/25
                          4s 154ms/step - accuracy: 0.8687 - loss: 0.3920 - val_accuracy: 0.5850 - val_loss: 1.6867
Epoch 60/200
25/25
                          4s 152ms/step - accuracy: 0.8751 - loss: 0.3240 - val_accuracy: 0.6350 - val_loss: 1.5783
Epoch 61/200
25/25
                          4s 150ms/step - accuracy: 0.8843 - loss: 0.3223 - val_accuracy: 0.6150 - val_loss: 1.7564
Epoch 62/200
25/25
                          4s 155ms/step - accuracy: 0.8901 - loss: 0.3179 - val_accuracy: 0.6050 - val_loss: 1.6935
Epoch 63/200
25/25
                          4s 156ms/step - accuracy: 0.9084 - loss: 0.2651 - val_accuracy: 0.6450 - val_loss: 1.5367
Epoch 64/200
25/25
                           4s 159ms/step - accuracy: 0.9011 - loss: 0.2681 - val_accuracy: 0.6550 - val_loss: 1.5078
Epoch 65/200
25/25
                          4s 153ms/step - accuracy: 0.9111 - loss: 0.2012 - val_accuracy: 0.6550 - val_loss: 1.7399
Epoch 66/200
25/25
                          4s 154ms/step - accuracy: 0.9056 - loss: 0.2551 - val_accuracy: 0.6450 - val_loss: 1.7057
Epoch 67/200
25/25
                          • 4s 153ms/step - accuracy: 0.9123 - loss: 0.2318 - val_accuracy: 0.6200 - val_loss: 1.7583
Epoch 68/200
25/25
                          4s 148ms/step - accuracy: 0.9229 - loss: 0.2184 - val_accuracy: 0.6200 - val_loss: 1.7297
Epoch 69/200
25/25
                          4s 151ms/step - accuracy: 0.9322 - loss: 0.1998 - val_accuracy: 0.6450 - val_loss: 1.6918
Epoch 70/200
25/25
                          4s 153ms/step - accuracy: 0.9346 - loss: 0.1579 - val_accuracy: 0.6300 - val_loss: 1.9413
Epoch 71/200
25/25
                          4s 149ms/step - accuracy: 0.9307 - loss: 0.1901 - val_accuracy: 0.6100 - val_loss: 1.6549
Epoch 72/200
25/25
                           4s 154ms/step - accuracy: 0.8986 - loss: 0.2894 - val_accuracy: 0.6350 - val_loss: 1.7343
Epoch 73/200
25/25
                          4s 156ms/step - accuracy: 0.9419 - loss: 0.1680 - val accuracy: 0.6350 - val loss: 1.8229
Epoch 74/200
25/25
                          4s 152ms/step - accuracy: 0.9522 - loss: 0.1810 - val_accuracy: 0.6350 - val_loss: 1.8504
Epoch 75/200
25/25
                          4s 152ms/step - accuracy: 0.9320 - loss: 0.2007 - val_accuracy: 0.5950 - val_loss: 1.8805
Epoch 76/200
25/25
                          4s 152ms/step - accuracy: 0.9319 - loss: 0.1726 - val_accuracy: 0.6050 - val_loss: 2.3080
Epoch 77/200
25/25
                          4s 152ms/step - accuracy: 0.9378 - loss: 0.1729 - val accuracy: 0.5850 - val loss: 2.5177
Epoch 78/200
25/25
                          - 4s 153ms/step - accuracy: 0.9300 - loss: 0.2347 - val_accuracy: 0.6300 - val_loss: 1.8858
Epoch 79/200
25/25
                          4s 150ms/step - accuracy: 0.9570 - loss: 0.1181 - val_accuracy: 0.6500 - val_loss: 1.7856
Epoch 80/200
25/25
                           4s 156ms/step - accuracy: 0.9564 - loss: 0.1307 - val_accuracy: 0.6600 - val_loss: 1.9016
Epoch 81/200
25/25
                          4s 153ms/step - accuracy: 0.9573 - loss: 0.1220 - val_accuracy: 0.6350 - val_loss: 2.2473
Epoch 82/200
25/25
                          4s 147ms/step - accuracy: 0.9390 - loss: 0.1833 - val_accuracy: 0.6300 - val_loss: 1.9236
```

```
Epoch 83/200
25/25
                          4s 148ms/step - accuracy: 0.9190 - loss: 0.2433 - val_accuracy: 0.6200 - val_loss: 1.9956
Epoch 84/200
25/25
                          4s 148ms/step - accuracy: 0.9646 - loss: 0.0943 - val_accuracy: 0.6400 - val_loss: 1.9079
Epoch 85/200
25/25
                          4s 151ms/step - accuracy: 0.9253 - loss: 0.2035 - val_accuracy: 0.6650 - val_loss: 1.9023
Epoch 86/200
25/25
                          4s 151ms/step - accuracy: 0.9570 - loss: 0.1210 - val_accuracy: 0.6300 - val_loss: 2.2414
Epoch 87/200
25/25
                          4s 155ms/step - accuracy: 0.9634 - loss: 0.1130 - val_accuracy: 0.6800 - val_loss: 1.7807
Epoch 88/200
25/25
                          4s 152ms/step - accuracy: 0.9449 - loss: 0.1721 - val_accuracy: 0.5950 - val_loss: 2.3276
Epoch 89/200
25/25
                          4s 158ms/step - accuracy: 0.9632 - loss: 0.1476 - val_accuracy: 0.6400 - val_loss: 1.9755
Epoch 90/200
25/25
                          4s 146ms/step - accuracy: 0.9655 - loss: 0.1029 - val_accuracy: 0.6450 - val_loss: 2.1367
Epoch 91/200
                          4s 147ms/step - accuracy: 0.9527 - loss: 0.1291 - val_accuracy: 0.6400 - val_loss: 2.1940
25/25
Epoch 92/200
25/25
                          4s 148ms/step - accuracy: 0.9660 - loss: 0.0966 - val_accuracy: 0.6100 - val_loss: 2.1825
Epoch 93/200
25/25
                          4s 148ms/step - accuracy: 0.9590 - loss: 0.1117 - val_accuracy: 0.6500 - val_loss: 2.4047
Epoch 94/200
25/25
                          4s 153ms/step - accuracy: 0.9492 - loss: 0.1480 - val_accuracy: 0.6200 - val_loss: 2.1937
Epoch 95/200
                          4s 153ms/step - accuracy: 0.9725 - loss: 0.0931 - val_accuracy: 0.5950 - val_loss: 2.4867
25/25
Epoch 96/200
25/25
                          4s 155ms/step - accuracy: 0.9345 - loss: 0.1980 - val_accuracy: 0.5950 - val_loss: 2.5017
Epoch 97/200
25/25
                          4s 156ms/step - accuracy: 0.9536 - loss: 0.1214 - val_accuracy: 0.6600 - val_loss: 2.3244
Epoch 98/200
25/25
                          - 4s 153ms/step - accuracy: 0.9634 - loss: 0.1116 - val_accuracy: 0.6050 - val_loss: 2.7636
Epoch 99/200
25/25
                          4s 151ms/step - accuracy: 0.9595 - loss: 0.1351 - val_accuracy: 0.6100 - val_loss: 2.1780
Epoch 100/200
25/25
                          4s 154ms/step - accuracy: 0.9683 - loss: 0.0903 - val_accuracy: 0.6450 - val_loss: 2.2073
Epoch 101/200
25/25
                          4s 150ms/step - accuracy: 0.9749 - loss: 0.0874 - val_accuracy: 0.6300 - val_loss: 2.6755
Epoch 102/200
25/25
                          4s 150ms/step - accuracy: 0.9796 - loss: 0.0895 - val_accuracy: 0.6300 - val_loss: 2.2189
Epoch 103/200
25/25
                          4s 149ms/step - accuracy: 0.9510 - loss: 0.1476 - val_accuracy: 0.5800 - val_loss: 2.4473
Epoch 104/200
25/25
                          4s 157ms/step - accuracy: 0.9651 - loss: 0.1142 - val_accuracy: 0.6400 - val_loss: 2.1473
Epoch 105/200
25/25
                          4s 151ms/step - accuracy: 0.9715 - loss: 0.1023 - val_accuracy: 0.5900 - val_loss: 2.3329
Epoch 106/200
25/25
                          4s 154ms/step - accuracy: 0.9530 - loss: 0.1820 - val_accuracy: 0.6000 - val_loss: 2.7490
Epoch 107/200
25/25
                          4s 152ms/step - accuracy: 0.9564 - loss: 0.1376 - val_accuracy: 0.6500 - val_loss: 2.3251
Epoch 108/200
25/25
                          • 4s 154ms/step - accuracy: 0.9596 - loss: 0.1519 - val_accuracy: 0.6550 - val_loss: 2.4093
Epoch 109/200
25/25
                          4s 152ms/step - accuracy: 0.9653 - loss: 0.0969 - val_accuracy: 0.5850 - val_loss: 2.8712
Epoch 110/200
25/25
                          4s 157ms/step - accuracy: 0.9487 - loss: 0.1415 - val_accuracy: 0.6350 - val_loss: 2.0791
Epoch 111/200
25/25
                          4s 153ms/step - accuracy: 0.9632 - loss: 0.1062 - val_accuracy: 0.6350 - val_loss: 2.1070
Epoch 112/200
25/25
                          4s 154ms/step - accuracy: 0.9727 - loss: 0.0791 - val_accuracy: 0.6250 - val_loss: 2.1753
Epoch 113/200
25/25
                          4s 150ms/step - accuracy: 0.9769 - loss: 0.0716 - val_accuracy: 0.6500 - val_loss: 2.0458
Epoch 114/200
25/25
                          4s 152ms/step - accuracy: 0.9788 - loss: 0.0582 - val accuracy: 0.6400 - val loss: 2.2886
Epoch 115/200
25/25
                          4s 156ms/step - accuracy: 0.9330 - loss: 0.2581 - val_accuracy: 0.6450 - val_loss: 2.3485
Epoch 116/200
25/25
                          4s 151ms/step - accuracy: 0.9665 - loss: 0.1004 - val_accuracy: 0.6450 - val_loss: 2.5679
Epoch 117/200
25/25
                          4s 150ms/step - accuracy: 0.9841 - loss: 0.0572 - val_accuracy: 0.6350 - val_loss: 2.4322
Epoch 118/200
25/25
                          4s 153ms/step - accuracy: 0.9653 - loss: 0.0872 - val accuracy: 0.6350 - val loss: 2.3860
Epoch 119/200
25/25
                          - 4s 152ms/step - accuracy: 0.9739 - loss: 0.0839 - val_accuracy: 0.6250 - val_loss: 2.6356
Epoch 120/200
25/25
                          4s 147ms/step - accuracy: 0.9751 - loss: 0.0728 - val_accuracy: 0.6400 - val_loss: 2.5431
Epoch 121/200
25/25
                           4s 157ms/step - accuracy: 0.9766 - loss: 0.0772 - val_accuracy: 0.6450 - val_loss: 2.2422
Epoch 122/200
25/25
                          4s 152ms/step - accuracy: 0.9807 - loss: 0.0660 - val_accuracy: 0.6250 - val_loss: 2.4464
Epoch 123/200
25/25
                          4s 156ms/step - accuracy: 0.9760 - loss: 0.0668 - val_accuracy: 0.6100 - val_loss: 2.8307
```

```
Epoch 124/200
                          4s 153ms/step - accuracy: 0.9725 - loss: 0.0801 - val_accuracy: 0.6250 - val_loss: 2.3358
25/25
Epoch 125/200
25/25
                          4s 155ms/step - accuracy: 0.9702 - loss: 0.1228 - val_accuracy: 0.6500 - val_loss: 2.3098
Epoch 126/200
25/25
                          4s 152ms/step - accuracy: 0.9603 - loss: 0.1196 - val_accuracy: 0.6550 - val_loss: 2.4535
Epoch 127/200
25/25
                          4s 152ms/step - accuracy: 0.9828 - loss: 0.0520 - val_accuracy: 0.6400 - val_loss: 2.5397
Epoch 128/200
25/25
                          • 4s 153ms/step - accuracy: 0.9824 - loss: 0.0516 - val_accuracy: 0.6250 - val_loss: 2.7237
Epoch 129/200
25/25
                          4s 155ms/step - accuracy: 0.9866 - loss: 0.0471 - val_accuracy: 0.6250 - val_loss: 2.4941
Epoch 130/200
25/25
                          4s 154ms/step - accuracy: 0.9766 - loss: 0.0721 - val_accuracy: 0.6350 - val_loss: 2.6602
Epoch 131/200
25/25
                          4s 153ms/step - accuracy: 0.9856 - loss: 0.0327 - val_accuracy: 0.6400 - val_loss: 3.0325
Epoch 132/200
                          4s 151ms/step - accuracy: 0.9794 - loss: 0.0530 - val_accuracy: 0.6200 - val_loss: 2.9336
25/25
Epoch 133/200
25/25
                          4s 153ms/step - accuracy: 0.9736 - loss: 0.0823 - val_accuracy: 0.6100 - val_loss: 2.8365
Epoch 134/200
25/25
                          4s 154ms/step - accuracy: 0.9675 - loss: 0.1192 - val_accuracy: 0.6150 - val_loss: 2.5089
Epoch 135/200
25/25
                          4s 156ms/step - accuracy: 0.9666 - loss: 0.1056 - val_accuracy: 0.6550 - val_loss: 2.5584
Epoch 136/200
                          4s 156ms/step - accuracy: 0.9863 - loss: 0.0400 - val_accuracy: 0.6100 - val_loss: 2.6453
25/25
Epoch 137/200
25/25
                          4s 157ms/step - accuracy: 0.9780 - loss: 0.0677 - val_accuracy: 0.6500 - val_loss: 2.3392
Epoch 138/200
                          4s 154ms/step - accuracy: 0.9820 - loss: 0.0461 - val_accuracy: 0.6400 - val_loss: 2.6747
25/25
Epoch 139/200
25/25
                          - 4s 151ms/step - accuracy: 0.9669 - loss: 0.1188 - val_accuracy: 0.6150 - val_loss: 2.7924
Epoch 140/200
25/25
                          4s 152ms/step - accuracy: 0.9767 - loss: 0.0923 - val_accuracy: 0.6300 - val_loss: 2.4425
Epoch 141/200
25/25
                          4s 152ms/step - accuracy: 0.9895 - loss: 0.0273 - val_accuracy: 0.6000 - val_loss: 2.8783
Epoch 142/200
25/25
                          4s 159ms/step - accuracy: 0.9818 - loss: 0.0588 - val_accuracy: 0.6200 - val_loss: 3.2892
Epoch 143/200
25/25
                          4s 152ms/step - accuracy: 0.9920 - loss: 0.0224 - val_accuracy: 0.6350 - val_loss: 2.8748
Epoch 144/200
25/25
                          4s 152ms/step - accuracy: 0.9902 - loss: 0.0372 - val_accuracy: 0.6450 - val_loss: 2.9861
Epoch 145/200
25/25
                          4s 149ms/step - accuracy: 0.9819 - loss: 0.0749 - val_accuracy: 0.6250 - val_loss: 2.6517
Epoch 146/200
25/25
                          4s 154ms/step - accuracy: 0.9738 - loss: 0.0858 - val_accuracy: 0.6450 - val_loss: 2.7020
Epoch 147/200
25/25
                          4s 155ms/step - accuracy: 0.9814 - loss: 0.0641 - val_accuracy: 0.6250 - val_loss: 2.7762
Epoch 148/200
25/25
                          4s 152ms/step - accuracy: 0.9870 - loss: 0.0443 - val_accuracy: 0.5850 - val_loss: 3.0334
Epoch 149/200
25/25
                          • 4s 156ms/step - accuracy: 0.9825 - loss: 0.0449 - val_accuracy: 0.6450 - val_loss: 2.9334
Epoch 150/200
25/25
                          4s 153ms/step - accuracy: 0.9744 - loss: 0.1010 - val_accuracy: 0.6250 - val_loss: 2.8921
Epoch 151/200
25/25
                          4s 152ms/step - accuracy: 0.9817 - loss: 0.0701 - val_accuracy: 0.6300 - val_loss: 2.8880
Epoch 152/200
25/25
                          4s 155ms/step - accuracy: 0.9804 - loss: 0.0619 - val_accuracy: 0.6450 - val_loss: 2.9605
Epoch 153/200
25/25
                          4s 153ms/step - accuracy: 0.9932 - loss: 0.0343 - val_accuracy: 0.6550 - val_loss: 2.7758
Epoch 154/200
25/25
                          4s 148ms/step - accuracy: 0.9638 - loss: 0.1059 - val_accuracy: 0.6250 - val_loss: 3.0840
Epoch 155/200
25/25
                          4s 152ms/step - accuracy: 0.9836 - loss: 0.0825 - val accuracy: 0.5850 - val loss: 3.1181
Epoch 156/200
25/25
                          4s 156ms/step - accuracy: 0.9827 - loss: 0.1365 - val_accuracy: 0.6250 - val_loss: 2.3661
Epoch 157/200
25/25
                          4s 156ms/step - accuracy: 0.9843 - loss: 0.0540 - val_accuracy: 0.6200 - val_loss: 3.2481
Epoch 158/200
25/25
                          4s 157ms/step - accuracy: 0.9880 - loss: 0.0368 - val_accuracy: 0.6100 - val_loss: 2.9156
Epoch 159/200
25/25
                          4s 156ms/step - accuracy: 0.9795 - loss: 0.0625 - val accuracy: 0.6350 - val loss: 2.9362
Epoch 160/200
                          - 4s 154ms/step - accuracy: 0.9871 - loss: 0.0501 - val_accuracy: 0.6100 - val_loss: 2.6436
25/25
Epoch 161/200
25/25
                          4s 155ms/step - accuracy: 0.9893 - loss: 0.0645 - val_accuracy: 0.6000 - val_loss: 3.0421
Epoch 162/200
25/25
                           4s 155ms/step - accuracy: 0.9654 - loss: 0.0766 - val_accuracy: 0.6750 - val_loss: 2.8712
Epoch 163/200
25/25
                          4s 161ms/step - accuracy: 0.9946 - loss: 0.0277 - val_accuracy: 0.6450 - val_loss: 3.0525
Epoch 164/200
25/25
                          4s 151ms/step - accuracy: 0.9867 - loss: 0.0390 - val_accuracy: 0.6050 - val_loss: 3.7059
```

```
Epoch 165/200
                                  4s 151ms/step - accuracy: 0.9782 - loss: 0.0850 - val_accuracy: 0.6400 - val_loss: 3.2004
       25/25
       Epoch 166/200
       25/25
                                 • 4s 150ms/step - accuracy: 0.9659 - loss: 0.1274 - val_accuracy: 0.6400 - val_loss: 2.7282
       Epoch 167/200
      25/25
                                 4s 156ms/step - accuracy: 0.9869 - loss: 0.0290 - val_accuracy: 0.6350 - val_loss: 2.9333
       Epoch 168/200
      25/25
                                 • 4s 151ms/step - accuracy: 0.9765 - loss: 0.0736 - val_accuracy: 0.6200 - val_loss: 3.0869
       Epoch 169/200
      25/25
                                 - 4s 153ms/step - accuracy: 0.9709 - loss: 0.0626 - val_accuracy: 0.6200 - val_loss: 3.1077
      Epoch 170/200
      25/25
                                 - 4s 152ms/step - accuracy: 0.9798 - loss: 0.0398 - val_accuracy: 0.6150 - val_loss: 3.1138
       Epoch 171/200
      25/25
                                 4s 152ms/step - accuracy: 0.9849 - loss: 0.0470 - val_accuracy: 0.6300 - val_loss: 3.1697
       Epoch 172/200
       25/25
                                 • 4s 153ms/step - accuracy: 0.9905 - loss: 0.0274 - val_accuracy: 0.6200 - val_loss: 2.9240
       Epoch 173/200
                                 4s 151ms/step - accuracy: 0.9799 - loss: 0.0688 - val_accuracy: 0.6250 - val_loss: 3.0502
      25/25
      Epoch 174/200
       25/25
                                 - 4s 152ms/step - accuracy: 0.9785 - loss: 0.0580 - val_accuracy: 0.5950 - val_loss: 3.1900
      Epoch 175/200
       25/25
                                 4s 153ms/step - accuracy: 0.9910 - loss: 0.0383 - val_accuracy: 0.6300 - val_loss: 3.1558
       Epoch 176/200
      25/25
                                 4s 152ms/step - accuracy: 0.9687 - loss: 0.0718 - val_accuracy: 0.6350 - val_loss: 3.1910
       Epoch 177/200
                                 4s 153ms/step - accuracy: 0.9696 - loss: 0.0787 - val_accuracy: 0.6450 - val_loss: 2.9201
       25/25
       Epoch 178/200
      25/25
                                 • 4s 155ms/step - accuracy: 0.9831 - loss: 0.0540 - val_accuracy: 0.6400 - val_loss: 2.9415
      Epoch 179/200
                                 • 4s 152ms/step - accuracy: 0.9862 - loss: 0.0781 - val_accuracy: 0.6650 - val_loss: 3.0149
       25/25
       Epoch 180/200
      25/25
                                - 4s 152ms/step - accuracy: 0.9834 - loss: 0.0509 - val_accuracy: 0.6450 - val_loss: 2.8482
       Epoch 181/200
      25/25
                                 - 4s 152ms/step - accuracy: 0.9862 - loss: 0.0647 - val_accuracy: 0.6500 - val_loss: 3.4474
       Epoch 182/200
      25/25 •
                                 - 4s 155ms/step - accuracy: 0.9794 - loss: 0.1101 - val_accuracy: 0.6500 - val_loss: 2.7705
       Epoch 183/200
       25/25
                                 4s 152ms/step - accuracy: 0.9850 - loss: 0.0286 - val_accuracy: 0.6600 - val_loss: 3.5766
       Epoch 184/200
      25/25
                                 - 4s 156ms/step - accuracy: 0.9922 - loss: 0.0405 - val_accuracy: 0.6650 - val_loss: 2.8988
       Epoch 185/200
                                 - 4s 151ms/step - accuracy: 0.9823 - loss: 0.0602 - val_accuracy: 0.6500 - val_loss: 2.8050
       25/25
       Epoch 186/200
      25/25
                                 • 4s 156ms/step - accuracy: 0.9909 - loss: 0.0293 - val_accuracy: 0.6450 - val_loss: 2.7298
       Epoch 187/200
       25/25 -
                                 4s 154ms/step - accuracy: 0.9860 - loss: 0.0411 - val_accuracy: 0.6450 - val_loss: 3.0310
       Epoch 188/200
      25/25
                                 - 4s 151ms/step - accuracy: 0.9940 - loss: 0.0242 - val_accuracy: 0.6600 - val_loss: 3.1644
       Epoch 189/200
      25/25
                                 - 4s 151ms/step - accuracy: 0.9837 - loss: 0.0473 - val_accuracy: 0.6550 - val_loss: 2.8926
      Epoch 190/200
      25/25
                                 - 4s 156ms/step - accuracy: 0.9790 - loss: 0.1066 - val_accuracy: 0.6300 - val_loss: 2.9581
      Epoch 191/200
       25/25
                                 4s 152ms/step - accuracy: 0.9876 - loss: 0.0275 - val_accuracy: 0.6350 - val_loss: 3.0488
       Epoch 192/200
      25/25
                                 - 4s 153ms/step - accuracy: 0.9851 - loss: 0.0696 - val_accuracy: 0.6200 - val_loss: 3.1620
       Epoch 193/200
       25/25
                                 - 4s 154ms/step - accuracy: 0.9921 - loss: 0.0154 - val_accuracy: 0.6500 - val_loss: 3.8708
       Epoch 194/200
      25/25
                                 4s 153ms/step - accuracy: 0.9761 - loss: 0.1093 - val_accuracy: 0.6750 - val_loss: 2.9113
      Epoch 195/200
       25/25
                                 4s 151ms/step - accuracy: 0.9773 - loss: 0.0973 - val_accuracy: 0.6550 - val_loss: 3.0089
      Epoch 196/200
      25/25
                                 • 4s 158ms/step - accuracy: 0.9829 - loss: 0.0561 - val_accuracy: 0.6550 - val_loss: 2.9303
       Epoch 197/200
      25/25
                                 • 4s 152ms/step - accuracy: 0.9827 - loss: 0.0508 - val_accuracy: 0.6300 - val_loss: 3.1155
       Epoch 198/200
       25/25
                                 4s 155ms/step - accuracy: 0.9807 - loss: 0.0829 - val_accuracy: 0.6350 - val_loss: 2.8135
       Epoch 199/200
       25/25
                                 4s 155ms/step - accuracy: 0.9841 - loss: 0.0454 - val_accuracy: 0.6450 - val_loss: 3.5844
       Epoch 200/200
       25/25
                                 4s 148ms/step - accuracy: 0.9732 - loss: 0.0892 - val accuracy: 0.6450 - val loss: 2.8868
In [8]: print(f"Execution time: {elapsed_time:.2f} seconds")
```

Execution time: 770.62 seconds

```
In [9]: def append_core_data(score_path, num_cores, elapsed_time):
    # Check if the file already exists
    file_exists = os.path.exists(score_path)

# Open the file in append mode
with open(score_path, mode='a', newline='') as file:
```

```
writer = csv.writer(file)

# If the file is new, write the header
if not file_exists:
    writer.writerow(["Number of Cores", "Elapsed Time"])

# Write the new data
writer.writerow([num_cores, elapsed_time])
```

In [10]: score_path = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DEEP LEARNING WITH HPSC\core_data.txt"
 append_core_data(score_path, number_of_worker, elapsed_time)

Import Req Lib

```
In [1]: %matplotlib inline
        import shutil
        import random
        import numpy as np
        from warnings import filterwarnings
        filterwarnings('ignore')
        from tensorflow.keras import layers, regularizers, optimizers
        from tensorflow.keras import models
        from tensorflow.keras.models import Sequential, Model
        from tensorflow.keras.layers import LeakyReLU,Dense, Activation, Flatten, Dropout, BatchNormalization,Conv2D, MaxPooling2D
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import tensorflow as tf
        import os
        import time
        import csv
        from matplotlib import figure
```

Define 8 worker

```
In [2]: # Set the number of threads
number_of_worker = 8
os.environ['OMP_NUM_THREADS'] = '8' # OpenMP threads for parallelism
os.environ['TF_NUM_INTEROP_THREADS'] = '8' # Threads for inter-operation parallelism
os.environ['TF_NUM_INTRAOP_THREADS'] = '8' # Threads for intra-operation parallelism
# Confirm TensorFlow is using the specified number of threads
tf.config.threading.set_inter_op_parallelism_threads(number_of_worker)
tf.config.threading.set_intra_op_parallelism_threads(number_of_worker)
```

Train Val data Split

```
In [3]: source_dir = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DATA\Convert_Audio_File_to_jpg_file"
        target_dir = r'genres_train_val_split_data'
        split ratio = 0.8
        def Train_Test_Split(source_dir,target_dir,split_ratio):
            # Define source and target directories
            train_dir = os.path.join(target_dir, 'train')
            val_dir = os.path.join(target_dir, 'val')
            # Create target directories if they don't exist
            os.makedirs(train_dir, exist_ok=True)
            os.makedirs(val_dir, exist_ok=True)
            # Get the list of class directories
            classes = [d for d in os.listdir(source_dir) if os.path.isdir(os.path.join(source_dir, d))]
            for class_name in classes:
                # Create class directories in train and val folders
                os.makedirs(os.path.join(train_dir, class_name), exist_ok=True)
                os.makedirs(os.path.join(val_dir, class_name), exist_ok=True)
                # Get list of images in the class directory
                class_dir = os.path.join(source_dir, class_name)
                images = [f for f in os.listdir(class_dir) if os.path.isfile(os.path.join(class_dir, f))]
                # Shuffle the images
                random.shuffle(images)
                # Compute the split point
                split_point = int(len(images) * split_ratio)
                # Split the images into training and validation sets
                train_images = images[:split_point]
                val_images = images[split_point:]
```

Data split completed successfully!

Load the Data

```
In [5]: WIDTH = 64
        HEIGHT = 64
        BATCH_SIZE = 32
        TRAIN_DIR=r'genres_train_val_split_data/train'
        val_dir = r'genres_train_val_split_data/val'
        # data prep
        train_datagen = ImageDataGenerator(
            rescale=1./255.,validation_split=0.25)
        train_generator = train_datagen.flow_from_directory(
            TRAIN DIR,
            target_size=(HEIGHT, WIDTH),
                batch_size=BATCH_SIZE,
                class_mode='categorical')
        validation_gen = train_datagen.flow_from_directory(
            val_dir,target_size = (HEIGHT,WIDTH),
            batch_size = BATCH_SIZE,
            class_mode = 'categorical'
         )
```

Found 800 images belonging to 10 classes. Found 200 images belonging to 10 classes.

Model Architecture

```
In [6]: model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                         input_shape=(64,64,3)))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(128, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(128, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(10, activation='softmax'))
        model.compile(optimizers.RMSprop(learning_rate=0.0005, decay=1e-6),loss="categorical_crossentropy",metrics=["accuracy"])
        model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 64, 64, 32)	896
activation (Activation)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 62, 62, 64)	18,496
activation_1 (Activation)	(None, 62, 62, 64)	0
max_pooling2d (MaxPooling2D)	(None, 31, 31, 64)	0
dropout (Dropout)	(None, 31, 31, 64)	0
conv2d_2 (Conv2D)	(None, 31, 31, 64)	36,928
activation_2 (Activation)	(None, 31, 31, 64)	0
conv2d_3 (Conv2D)	(None, 29, 29, 64)	36,928
activation_3 (Activation)	(None, 29, 29, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_1 (Dropout)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 14, 14, 128)	73,856
activation_4 (Activation)	(None, 14, 14, 128)	0
conv2d_5 (Conv2D)	(None, 12, 12, 128)	147,584
activation_5 (Activation)	(None, 12, 12, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_2 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2,359,808
activation_6 (Activation)	(None, 512)	0
dropout_3 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5,130

Total params: 2,679,626 (10.22 MB)

Trainable params: 2,679,626 (10.22 MB)

Non-trainable params: 0 (0.00 B)

```
In [7]: STEP_SIZE_TRAIN=train_generator.n//train_generator.batch_size
# Measure the execution time
start_time = time.time()

model.fit(train_generator,validation_data=validation_gen,epochs=200)

end_time = time.time()
elapsed_time = end_time - start_time
```

```
Epoch 1/200
                          4s 126ms/step - accuracy: 0.0908 - loss: 2.3262 - val_accuracy: 0.1000 - val_loss: 2.3015
25/25
Epoch 2/200
25/25 -
                          3s 135ms/step - accuracy: 0.1362 - loss: 2.2996 - val_accuracy: 0.1950 - val_loss: 2.2538
Epoch 3/200
25/25
                          3s 136ms/step - accuracy: 0.1901 - loss: 2.2619 - val_accuracy: 0.1550 - val_loss: 2.1788
Epoch 4/200
25/25
                          3s 136ms/step - accuracy: 0.2647 - loss: 2.0942 - val_accuracy: 0.1650 - val_loss: 2.1146
Epoch 5/200
25/25
                          - 3s 135ms/step - accuracy: 0.2221 - loss: 2.0821 - val_accuracy: 0.2550 - val_loss: 2.0688
Epoch 6/200
25/25
                          4s 140ms/step - accuracy: 0.2638 - loss: 1.9729 - val_accuracy: 0.1850 - val_loss: 2.1266
Epoch 7/200
25/25
                          4s 140ms/step - accuracy: 0.2670 - loss: 2.0135 - val_accuracy: 0.3050 - val_loss: 1.9701
Epoch 8/200
25/25
                          3s 137ms/step - accuracy: 0.3221 - loss: 1.8983 - val_accuracy: 0.3200 - val_loss: 1.9209
Epoch 9/200
                          4s 142ms/step - accuracy: 0.3465 - loss: 1.8112 - val_accuracy: 0.3850 - val_loss: 1.7874
25/25
Epoch 10/200
25/25
                          3s 134ms/step - accuracy: 0.3333 - loss: 1.7993 - val_accuracy: 0.3800 - val_loss: 1.7778
Epoch 11/200
25/25
                          4s 141ms/step - accuracy: 0.3197 - loss: 1.7694 - val_accuracy: 0.3800 - val_loss: 1.8889
Epoch 12/200
25/25
                          4s 139ms/step - accuracy: 0.4073 - loss: 1.7037 - val_accuracy: 0.3950 - val_loss: 1.7497
Epoch 13/200
                          4s 139ms/step - accuracy: 0.4157 - loss: 1.6517 - val_accuracy: 0.4050 - val_loss: 1.6553
25/25
Epoch 14/200
25/25
                          4s 139ms/step - accuracy: 0.4353 - loss: 1.5729 - val_accuracy: 0.4100 - val_loss: 1.5987
Epoch 15/200
                          4s 139ms/step - accuracy: 0.4026 - loss: 1.6035 - val_accuracy: 0.4650 - val_loss: 1.5758
25/25
Epoch 16/200
25/25
                          - 4s 142ms/step - accuracy: 0.4446 - loss: 1.5463 - val_accuracy: 0.4250 - val_loss: 1.5314
Epoch 17/200
25/25
                          4s 142ms/step - accuracy: 0.4380 - loss: 1.5158 - val_accuracy: 0.3150 - val_loss: 1.9509
Epoch 18/200
25/25
                          4s 140ms/step - accuracy: 0.4421 - loss: 1.5638 - val_accuracy: 0.5050 - val_loss: 1.5129
Epoch 19/200
25/25
                          3s 137ms/step - accuracy: 0.4866 - loss: 1.4481 - val_accuracy: 0.4600 - val_loss: 1.4984
Epoch 20/200
25/25
                          4s 141ms/step - accuracy: 0.5063 - loss: 1.3240 - val_accuracy: 0.5150 - val_loss: 1.4143
Epoch 21/200
25/25
                          4s 139ms/step - accuracy: 0.4910 - loss: 1.3623 - val_accuracy: 0.5200 - val_loss: 1.4117
Epoch 22/200
25/25
                          4s 144ms/step - accuracy: 0.5047 - loss: 1.3392 - val_accuracy: 0.4800 - val_loss: 1.5531
Epoch 23/200
25/25
                          3s 138ms/step - accuracy: 0.5424 - loss: 1.2446 - val_accuracy: 0.5400 - val_loss: 1.3136
Epoch 24/200
25/25
                          4s 140ms/step - accuracy: 0.5654 - loss: 1.2154 - val_accuracy: 0.4050 - val_loss: 1.7990
Epoch 25/200
                          4s 140ms/step - accuracy: 0.5409 - loss: 1.2645 - val_accuracy: 0.5100 - val_loss: 1.5017
25/25
Epoch 26/200
25/25
                          • 4s 139ms/step - accuracy: 0.6027 - loss: 1.1757 - val_accuracy: 0.5600 - val_loss: 1.3537
Epoch 27/200
25/25
                          4s 140ms/step - accuracy: 0.5567 - loss: 1.1760 - val_accuracy: 0.5450 - val_loss: 1.3311
Epoch 28/200
25/25
                          4s 140ms/step - accuracy: 0.6314 - loss: 1.0326 - val_accuracy: 0.4900 - val_loss: 1.6634
Epoch 29/200
25/25 -
                          3s 137ms/step - accuracy: 0.5486 - loss: 1.2524 - val_accuracy: 0.5700 - val_loss: 1.2929
Epoch 30/200
25/25
                          3s 138ms/step - accuracy: 0.6574 - loss: 1.0052 - val_accuracy: 0.5450 - val_loss: 1.3549
Epoch 31/200
25/25
                          3s 139ms/step - accuracy: 0.6195 - loss: 1.0483 - val_accuracy: 0.5400 - val_loss: 1.2411
Epoch 32/200
25/25
                          4s 142ms/step - accuracy: 0.6598 - loss: 0.9453 - val accuracy: 0.5750 - val loss: 1.2204
Epoch 33/200
25/25
                          4s 141ms/step - accuracy: 0.6279 - loss: 1.0213 - val_accuracy: 0.6000 - val_loss: 1.2158
Epoch 34/200
25/25
                          4s 138ms/step - accuracy: 0.6599 - loss: 0.9575 - val_accuracy: 0.6000 - val_loss: 1.2094
Epoch 35/200
25/25
                          4s 139ms/step - accuracy: 0.6886 - loss: 0.8812 - val_accuracy: 0.5850 - val_loss: 1.1563
Epoch 36/200
25/25
                          4s 139ms/step - accuracy: 0.6898 - loss: 0.8930 - val accuracy: 0.5550 - val loss: 1.2696
Epoch 37/200
25/25
                         - 3s 137ms/step - accuracy: 0.6576 - loss: 0.8930 - val_accuracy: 0.5500 - val_loss: 1.2612
Epoch 38/200
25/25
                          3s 139ms/step - accuracy: 0.6722 - loss: 0.9251 - val_accuracy: 0.5900 - val_loss: 1.2177
Epoch 39/200
25/25
                           4s 141ms/step - accuracy: 0.6881 - loss: 0.8160 - val_accuracy: 0.6000 - val_loss: 1.1883
Epoch 40/200
25/25
                          3s 138ms/step - accuracy: 0.7392 - loss: 0.7417 - val_accuracy: 0.5600 - val_loss: 1.1867
Epoch 41/200
25/25
                          3s 138ms/step - accuracy: 0.6914 - loss: 0.8287 - val_accuracy: 0.5600 - val_loss: 1.3192
```

```
Epoch 42/200
                          3s 137ms/step - accuracy: 0.7687 - loss: 0.6921 - val_accuracy: 0.5750 - val_loss: 1.3051
25/25
Epoch 43/200
25/25
                          4s 140ms/step - accuracy: 0.7411 - loss: 0.7306 - val_accuracy: 0.5950 - val_loss: 1.1532
Epoch 44/200
25/25
                          4s 141ms/step - accuracy: 0.7491 - loss: 0.6981 - val_accuracy: 0.5800 - val_loss: 1.4488
Epoch 45/200
                          4s 139ms/step - accuracy: 0.7577 - loss: 0.6257 - val_accuracy: 0.5050 - val_loss: 1.5136
25/25
Epoch 46/200
25/25
                          - 3s 137ms/step - accuracy: 0.7645 - loss: 0.6471 - val_accuracy: 0.5900 - val_loss: 1.2315
Epoch 47/200
25/25
                          4s 140ms/step - accuracy: 0.7872 - loss: 0.5983 - val_accuracy: 0.6150 - val_loss: 1.2239
Epoch 48/200
25/25
                          4s 142ms/step - accuracy: 0.7524 - loss: 0.7278 - val_accuracy: 0.6250 - val_loss: 1.1958
Epoch 49/200
25/25
                          4s 142ms/step - accuracy: 0.7851 - loss: 0.6215 - val_accuracy: 0.6300 - val_loss: 1.2098
Epoch 50/200
                          3s 138ms/step - accuracy: 0.8122 - loss: 0.5169 - val_accuracy: 0.6100 - val_loss: 1.3742
25/25
Epoch 51/200
25/25
                          4s 141ms/step - accuracy: 0.8208 - loss: 0.4952 - val_accuracy: 0.6300 - val_loss: 1.3932
Epoch 52/200
25/25
                          4s 139ms/step - accuracy: 0.8011 - loss: 0.5288 - val_accuracy: 0.5900 - val_loss: 1.3119
Epoch 53/200
25/25
                          4s 140ms/step - accuracy: 0.8117 - loss: 0.4985 - val_accuracy: 0.6000 - val_loss: 1.4959
Epoch 54/200
25/25
                          4s 139ms/step - accuracy: 0.8080 - loss: 0.4890 - val_accuracy: 0.6050 - val_loss: 1.4572
Epoch 55/200
25/25
                          4s 141ms/step - accuracy: 0.8633 - loss: 0.3825 - val_accuracy: 0.6000 - val_loss: 1.4302
Epoch 56/200
                          4s 141ms/step - accuracy: 0.8359 - loss: 0.4682 - val_accuracy: 0.5900 - val_loss: 1.4071
25/25
Epoch 57/200
25/25
                          - 4s 140ms/step - accuracy: 0.8595 - loss: 0.4078 - val_accuracy: 0.5450 - val_loss: 1.7839
Epoch 58/200
25/25
                          4s 142ms/step - accuracy: 0.8430 - loss: 0.3921 - val_accuracy: 0.5900 - val_loss: 1.5943
Epoch 59/200
25/25
                          4s 140ms/step - accuracy: 0.8386 - loss: 0.4653 - val_accuracy: 0.6600 - val_loss: 1.2770
Epoch 60/200
25/25
                          4s 141ms/step - accuracy: 0.8801 - loss: 0.3475 - val_accuracy: 0.5850 - val_loss: 1.5956
Epoch 61/200
25/25
                          4s 141ms/step - accuracy: 0.8934 - loss: 0.3076 - val_accuracy: 0.6100 - val_loss: 1.5560
Epoch 62/200
25/25
                          4s 141ms/step - accuracy: 0.8729 - loss: 0.2943 - val_accuracy: 0.5500 - val_loss: 2.1760
Epoch 63/200
25/25
                          3s 133ms/step - accuracy: 0.8659 - loss: 0.3484 - val_accuracy: 0.6050 - val_loss: 1.3787
Epoch 64/200
25/25
                          4s 139ms/step - accuracy: 0.8979 - loss: 0.3263 - val_accuracy: 0.6150 - val_loss: 1.2772
Epoch 65/200
25/25
                          4s 139ms/step - accuracy: 0.9240 - loss: 0.2520 - val_accuracy: 0.6150 - val_loss: 1.5082
Epoch 66/200
25/25
                          4s 143ms/step - accuracy: 0.9125 - loss: 0.2537 - val_accuracy: 0.6450 - val_loss: 1.6815
Epoch 67/200
25/25
                          • 4s 142ms/step - accuracy: 0.9034 - loss: 0.3102 - val_accuracy: 0.6150 - val_loss: 1.6648
Epoch 68/200
25/25
                          4s 142ms/step - accuracy: 0.8977 - loss: 0.2868 - val_accuracy: 0.6150 - val_loss: 1.5470
Epoch 69/200
25/25
                          4s 144ms/step - accuracy: 0.9120 - loss: 0.2509 - val_accuracy: 0.6050 - val_loss: 1.7457
Epoch 70/200
25/25
                          4s 141ms/step - accuracy: 0.8936 - loss: 0.2831 - val_accuracy: 0.5750 - val_loss: 2.1970
Epoch 71/200
                          4s 140ms/step - accuracy: 0.8854 - loss: 0.3099 - val_accuracy: 0.6350 - val_loss: 1.7903
25/25
Epoch 72/200
25/25
                          4s 139ms/step - accuracy: 0.9302 - loss: 0.2035 - val_accuracy: 0.6450 - val_loss: 1.5624
Epoch 73/200
25/25
                          3s 139ms/step - accuracy: 0.9376 - loss: 0.1596 - val accuracy: 0.6400 - val loss: 1.7233
Epoch 74/200
25/25
                          4s 140ms/step - accuracy: 0.9390 - loss: 0.1822 - val_accuracy: 0.6400 - val_loss: 1.7426
Epoch 75/200
25/25
                          3s 138ms/step - accuracy: 0.9161 - loss: 0.2388 - val_accuracy: 0.6150 - val_loss: 1.8065
Epoch 76/200
25/25
                          3s 138ms/step - accuracy: 0.9217 - loss: 0.2217 - val_accuracy: 0.5850 - val_loss: 2.0870
Epoch 77/200
25/25
                          4s 139ms/step - accuracy: 0.9474 - loss: 0.1783 - val accuracy: 0.5600 - val loss: 1.9619
Epoch 78/200
25/25
                          - 4s 140ms/step - accuracy: 0.9547 - loss: 0.1430 - val_accuracy: 0.6400 - val_loss: 1.7234
Epoch 79/200
25/25
                          4s 139ms/step - accuracy: 0.9492 - loss: 0.1798 - val_accuracy: 0.6450 - val_loss: 1.7415
Epoch 80/200
25/25
                          3s 139ms/step - accuracy: 0.9351 - loss: 0.1791 - val_accuracy: 0.6250 - val_loss: 2.0626
Epoch 81/200
25/25
                          4s 141ms/step - accuracy: 0.9433 - loss: 0.1763 - val_accuracy: 0.6250 - val_loss: 1.8965
Epoch 82/200
25/25
                          4s 140ms/step - accuracy: 0.9281 - loss: 0.2056 - val_accuracy: 0.6350 - val_loss: 1.8599
```

```
Epoch 83/200
                          4s 142ms/step - accuracy: 0.9496 - loss: 0.1491 - val_accuracy: 0.5900 - val_loss: 2.2457
25/25
Epoch 84/200
25/25
                          4s 142ms/step - accuracy: 0.9162 - loss: 0.2485 - val_accuracy: 0.5650 - val_loss: 2.3358
Epoch 85/200
25/25
                          4s 144ms/step - accuracy: 0.9598 - loss: 0.1338 - val_accuracy: 0.6200 - val_loss: 2.0118
Epoch 86/200
                          4s 141ms/step - accuracy: 0.9405 - loss: 0.1782 - val_accuracy: 0.6250 - val_loss: 1.8883
25/25
Epoch 87/200
25/25
                          4s 139ms/step - accuracy: 0.9804 - loss: 0.0627 - val_accuracy: 0.6350 - val_loss: 2.0973
Epoch 88/200
25/25
                          4s 141ms/step - accuracy: 0.9515 - loss: 0.1667 - val_accuracy: 0.6250 - val_loss: 2.0799
Epoch 89/200
25/25
                          3s 138ms/step - accuracy: 0.9499 - loss: 0.1645 - val_accuracy: 0.6100 - val_loss: 2.1547
Epoch 90/200
25/25
                          4s 140ms/step - accuracy: 0.9439 - loss: 0.1523 - val_accuracy: 0.6250 - val_loss: 2.1661
Epoch 91/200
                          3s 138ms/step - accuracy: 0.9245 - loss: 0.1971 - val_accuracy: 0.5800 - val_loss: 2.1904
25/25
Epoch 92/200
25/25
                          4s 141ms/step - accuracy: 0.9467 - loss: 0.1535 - val_accuracy: 0.6200 - val_loss: 2.0070
Epoch 93/200
25/25
                          4s 141ms/step - accuracy: 0.9467 - loss: 0.2095 - val_accuracy: 0.6500 - val_loss: 1.8759
Epoch 94/200
25/25
                          3s 135ms/step - accuracy: 0.9725 - loss: 0.0735 - val_accuracy: 0.6050 - val_loss: 2.1894
Epoch 95/200
                          3s 139ms/step - accuracy: 0.9641 - loss: 0.1138 - val_accuracy: 0.6100 - val_loss: 2.1479
25/25
Epoch 96/200
25/25
                          4s 139ms/step - accuracy: 0.9625 - loss: 0.1299 - val_accuracy: 0.5650 - val_loss: 1.9899
Epoch 97/200
                          4s 142ms/step - accuracy: 0.9669 - loss: 0.1340 - val_accuracy: 0.6200 - val_loss: 2.0378
25/25
Epoch 98/200
25/25
                          - 4s 140ms/step - accuracy: 0.9557 - loss: 0.1493 - val_accuracy: 0.5800 - val_loss: 2.0380
Epoch 99/200
25/25
                          3s 137ms/step - accuracy: 0.9268 - loss: 0.1841 - val_accuracy: 0.6200 - val_loss: 2.1253
Epoch 100/200
25/25
                          4s 141ms/step - accuracy: 0.9663 - loss: 0.1215 - val_accuracy: 0.6300 - val_loss: 2.0655
Epoch 101/200
25/25
                          4s 140ms/step - accuracy: 0.9568 - loss: 0.1693 - val_accuracy: 0.6250 - val_loss: 1.7176
Epoch 102/200
25/25
                          4s 142ms/step - accuracy: 0.9743 - loss: 0.0622 - val_accuracy: 0.5800 - val_loss: 2.7079
Epoch 103/200
25/25
                          4s 142ms/step - accuracy: 0.9500 - loss: 0.1408 - val_accuracy: 0.6400 - val_loss: 2.3660
Epoch 104/200
25/25
                          4s 140ms/step - accuracy: 0.9346 - loss: 0.1862 - val_accuracy: 0.6000 - val_loss: 2.5210
Epoch 105/200
25/25
                          4s 143ms/step - accuracy: 0.9648 - loss: 0.0955 - val_accuracy: 0.5850 - val_loss: 2.5043
Epoch 106/200
25/25
                          4s 142ms/step - accuracy: 0.9383 - loss: 0.2160 - val_accuracy: 0.6200 - val_loss: 2.3773
Epoch 107/200
                          3s 137ms/step - accuracy: 0.9772 - loss: 0.0925 - val_accuracy: 0.6300 - val_loss: 2.3062
25/25
Epoch 108/200
25/25
                          • 4s 140ms/step - accuracy: 0.9854 - loss: 0.0618 - val_accuracy: 0.5800 - val_loss: 2.2823
Epoch 109/200
25/25
                          3s 137ms/step - accuracy: 0.9648 - loss: 0.1287 - val_accuracy: 0.6500 - val_loss: 2.2313
Epoch 110/200
25/25
                          4s 139ms/step - accuracy: 0.9834 - loss: 0.0841 - val_accuracy: 0.6150 - val_loss: 2.8426
Epoch 111/200
25/25
                          4s 140ms/step - accuracy: 0.9567 - loss: 0.1696 - val_accuracy: 0.6000 - val_loss: 2.2645
Epoch 112/200
25/25
                          4s 143ms/step - accuracy: 0.9763 - loss: 0.0623 - val_accuracy: 0.5850 - val_loss: 3.1558
Epoch 113/200
25/25
                          4s 142ms/step - accuracy: 0.9779 - loss: 0.0871 - val_accuracy: 0.6000 - val_loss: 2.3316
Epoch 114/200
25/25
                          3s 139ms/step - accuracy: 0.9771 - loss: 0.0701 - val accuracy: 0.6100 - val loss: 2.5346
Epoch 115/200
25/25
                          4s 140ms/step - accuracy: 0.9910 - loss: 0.0450 - val_accuracy: 0.6200 - val_loss: 2.6165
Epoch 116/200
25/25
                          3s 133ms/step - accuracy: 0.9698 - loss: 0.0784 - val_accuracy: 0.6100 - val_loss: 2.5954
Epoch 117/200
25/25
                          4s 141ms/step - accuracy: 0.9836 - loss: 0.0577 - val_accuracy: 0.5850 - val_loss: 2.7277
Epoch 118/200
25/25
                          4s 139ms/step - accuracy: 0.9925 - loss: 0.0346 - val accuracy: 0.6300 - val loss: 2.8001
Epoch 119/200
25/25
                         - 3s 139ms/step - accuracy: 0.9619 - loss: 0.1764 - val_accuracy: 0.6100 - val_loss: 2.6465
Epoch 120/200
25/25
                          4s 141ms/step - accuracy: 0.9778 - loss: 0.0528 - val_accuracy: 0.5600 - val_loss: 3.3701
Epoch 121/200
25/25
                           4s 139ms/step - accuracy: 0.9516 - loss: 0.1404 - val_accuracy: 0.6050 - val_loss: 2.2734
Epoch 122/200
25/25
                          4s 144ms/step - accuracy: 0.9612 - loss: 0.1199 - val_accuracy: 0.5950 - val_loss: 3.1950
Epoch 123/200
25/25
                          3s 138ms/step - accuracy: 0.9744 - loss: 0.1086 - val_accuracy: 0.6000 - val_loss: 2.6465
```

```
Epoch 124/200
                          4s 143ms/step - accuracy: 0.9637 - loss: 0.0909 - val_accuracy: 0.6150 - val_loss: 2.5651
25/25
Epoch 125/200
25/25
                          4s 141ms/step - accuracy: 0.9712 - loss: 0.0857 - val_accuracy: 0.6150 - val_loss: 2.9187
Epoch 126/200
25/25
                          4s 139ms/step - accuracy: 0.9763 - loss: 0.0642 - val_accuracy: 0.6500 - val_loss: 2.5741
Epoch 127/200
25/25
                          3s 138ms/step - accuracy: 0.9774 - loss: 0.0712 - val_accuracy: 0.6100 - val_loss: 2.8007
Epoch 128/200
25/25
                          • 4s 143ms/step - accuracy: 0.9647 - loss: 0.1249 - val_accuracy: 0.6250 - val_loss: 2.3439
Epoch 129/200
25/25
                          4s 141ms/step - accuracy: 0.9923 - loss: 0.0272 - val_accuracy: 0.6050 - val_loss: 2.8320
Epoch 130/200
25/25
                          4s 142ms/step - accuracy: 0.9654 - loss: 0.0755 - val_accuracy: 0.6050 - val_loss: 2.5047
Epoch 131/200
25/25
                          3s 137ms/step - accuracy: 0.9896 - loss: 0.0388 - val_accuracy: 0.6000 - val_loss: 2.9230
Epoch 132/200
                          4s 139ms/step - accuracy: 0.9675 - loss: 0.1126 - val_accuracy: 0.5800 - val_loss: 2.9205
25/25
Epoch 133/200
25/25
                          3s 139ms/step - accuracy: 0.9697 - loss: 0.1045 - val_accuracy: 0.6300 - val_loss: 2.8700
Epoch 134/200
25/25
                          4s 141ms/step - accuracy: 0.9584 - loss: 0.1553 - val_accuracy: 0.6100 - val_loss: 3.2127
Epoch 135/200
25/25
                          4s 143ms/step - accuracy: 0.9559 - loss: 0.1463 - val_accuracy: 0.6400 - val_loss: 2.4648
Epoch 136/200
                          4s 143ms/step - accuracy: 0.9850 - loss: 0.0413 - val_accuracy: 0.6400 - val_loss: 2.4898
25/25
Epoch 137/200
25/25
                          4s 142ms/step - accuracy: 0.9781 - loss: 0.0685 - val_accuracy: 0.6100 - val_loss: 3.1533
Epoch 138/200
                          4s 140ms/step - accuracy: 0.9778 - loss: 0.0885 - val_accuracy: 0.6400 - val_loss: 2.9685
25/25
Epoch 139/200
25/25
                         - 3s 137ms/step - accuracy: 0.9756 - loss: 0.0563 - val_accuracy: 0.5950 - val_loss: 3.0030
Epoch 140/200
25/25
                          3s 140ms/step - accuracy: 0.9734 - loss: 0.1123 - val_accuracy: 0.6200 - val_loss: 2.6312
Epoch 141/200
25/25
                          4s 144ms/step - accuracy: 0.9894 - loss: 0.0447 - val_accuracy: 0.6000 - val_loss: 2.7892
Epoch 142/200
25/25
                          3s 136ms/step - accuracy: 0.9839 - loss: 0.0571 - val_accuracy: 0.6100 - val_loss: 3.5329
Epoch 143/200
25/25
                          4s 140ms/step - accuracy: 0.9584 - loss: 0.2007 - val_accuracy: 0.6000 - val_loss: 3.1208
Epoch 144/200
25/25
                          4s 140ms/step - accuracy: 0.9645 - loss: 0.0972 - val_accuracy: 0.6100 - val_loss: 3.0806
Epoch 145/200
25/25
                          4s 141ms/step - accuracy: 0.9682 - loss: 0.1397 - val_accuracy: 0.6200 - val_loss: 2.8241
Epoch 146/200
25/25
                          4s 141ms/step - accuracy: 0.9820 - loss: 0.0489 - val_accuracy: 0.6100 - val_loss: 2.8053
Epoch 147/200
25/25
                          4s 140ms/step - accuracy: 0.9878 - loss: 0.0593 - val_accuracy: 0.6150 - val_loss: 2.6795
Epoch 148/200
                          4s 145ms/step - accuracy: 0.9806 - loss: 0.0387 - val_accuracy: 0.6450 - val_loss: 3.2969
25/25
Epoch 149/200
25/25
                          • 4s 141ms/step - accuracy: 0.9720 - loss: 0.0867 - val_accuracy: 0.5750 - val_loss: 3.8260
Epoch 150/200
25/25
                          4s 144ms/step - accuracy: 0.9751 - loss: 0.0840 - val_accuracy: 0.6100 - val_loss: 3.1916
Epoch 151/200
25/25
                          4s 142ms/step - accuracy: 0.9577 - loss: 0.2086 - val_accuracy: 0.6250 - val_loss: 2.7172
Epoch 152/200
25/25
                          4s 141ms/step - accuracy: 0.9896 - loss: 0.0274 - val_accuracy: 0.6300 - val_loss: 2.7414
Epoch 153/200
25/25
                          4s 144ms/step - accuracy: 0.9864 - loss: 0.0404 - val_accuracy: 0.6200 - val_loss: 2.8900
Epoch 154/200
25/25
                          4s 140ms/step - accuracy: 0.9674 - loss: 0.1702 - val_accuracy: 0.6150 - val_loss: 2.7966
Epoch 155/200
25/25
                          3s 138ms/step - accuracy: 0.9900 - loss: 0.0298 - val accuracy: 0.6150 - val loss: 3.0291
Epoch 156/200
25/25
                          4s 141ms/step - accuracy: 0.9665 - loss: 0.1125 - val_accuracy: 0.6450 - val_loss: 2.6484
Epoch 157/200
25/25
                          3s 137ms/step - accuracy: 0.9833 - loss: 0.0632 - val_accuracy: 0.6200 - val_loss: 2.6650
Epoch 158/200
25/25
                          4s 140ms/step - accuracy: 0.9877 - loss: 0.0304 - val_accuracy: 0.5750 - val_loss: 3.2498
Epoch 159/200
25/25
                          4s 141ms/step - accuracy: 0.9653 - loss: 0.1523 - val accuracy: 0.6350 - val loss: 2.5158
Epoch 160/200
                         - 3s 138ms/step - accuracy: 0.9975 - loss: 0.0103 - val_accuracy: 0.6350 - val_loss: 2.9519
25/25
Epoch 161/200
25/25
                          4s 139ms/step - accuracy: 0.9862 - loss: 0.0531 - val_accuracy: 0.6300 - val_loss: 2.7204
Epoch 162/200
25/25
                          3s 138ms/step - accuracy: 0.9858 - loss: 0.0496 - val_accuracy: 0.6450 - val_loss: 3.1435
Epoch 163/200
25/25
                          4s 139ms/step - accuracy: 0.9840 - loss: 0.0454 - val_accuracy: 0.6350 - val_loss: 3.5224
Epoch 164/200
25/25
                          4s 139ms/step - accuracy: 0.9805 - loss: 0.0615 - val_accuracy: 0.6050 - val_loss: 3.2094
```

```
Epoch 165/200
       25/25
                                 4s 140ms/step - accuracy: 0.9871 - loss: 0.0618 - val_accuracy: 0.6350 - val_loss: 3.2888
       Epoch 166/200
       25/25
                                 • 4s 140ms/step - accuracy: 0.9696 - loss: 0.1007 - val_accuracy: 0.6400 - val_loss: 2.6891
       Epoch 167/200
      25/25
                                 4s 145ms/step - accuracy: 0.9872 - loss: 0.0435 - val_accuracy: 0.5900 - val_loss: 3.3853
       Epoch 168/200
      25/25
                                 • 4s 139ms/step - accuracy: 0.9914 - loss: 0.0244 - val_accuracy: 0.6200 - val_loss: 3.1923
       Epoch 169/200
      25/25
                                 - 4s 139ms/step - accuracy: 0.9791 - loss: 0.0804 - val_accuracy: 0.6300 - val_loss: 2.9425
      Epoch 170/200
      25/25
                                 - 3s 138ms/step - accuracy: 0.9858 - loss: 0.0313 - val_accuracy: 0.6400 - val_loss: 3.1421
       Epoch 171/200
      25/25
                                 4s 141ms/step - accuracy: 0.9780 - loss: 0.1068 - val_accuracy: 0.6050 - val_loss: 2.7188
       Epoch 172/200
       25/25
                                 • 4s 145ms/step - accuracy: 0.9848 - loss: 0.0615 - val_accuracy: 0.6200 - val_loss: 2.4824
       Epoch 173/200
                                 4s 143ms/step - accuracy: 0.9863 - loss: 0.0414 - val_accuracy: 0.6400 - val_loss: 2.7057
      25/25
      Epoch 174/200
       25/25
                                 - 4s 141ms/step - accuracy: 0.9884 - loss: 0.0382 - val_accuracy: 0.6700 - val_loss: 2.7627
      Epoch 175/200
       25/25
                                 4s 141ms/step - accuracy: 0.9889 - loss: 0.0458 - val_accuracy: 0.6450 - val_loss: 3.2791
       Epoch 176/200
      25/25
                                 4s 142ms/step - accuracy: 0.9926 - loss: 0.0161 - val_accuracy: 0.6350 - val_loss: 3.3782
       Epoch 177/200
                                 • 4s 140ms/step - accuracy: 0.9737 - loss: 0.0997 - val_accuracy: 0.5950 - val_loss: 2.9673
       25/25
       Epoch 178/200
      25/25
                                 • 4s 140ms/step - accuracy: 0.9740 - loss: 0.0582 - val_accuracy: 0.6250 - val_loss: 2.7655
      Epoch 179/200
                                 • 4s 140ms/step - accuracy: 0.9769 - loss: 0.0929 - val_accuracy: 0.6100 - val_loss: 3.2484
       25/25
       Epoch 180/200
      25/25
                                - 4s 140ms/step - accuracy: 0.9844 - loss: 0.0512 - val_accuracy: 0.6450 - val_loss: 3.1637
       Epoch 181/200
      25/25
                                 - 4s 141ms/step - accuracy: 0.9837 - loss: 0.0645 - val_accuracy: 0.6100 - val_loss: 3.5123
       Epoch 182/200
      25/25
                                 - 3s 138ms/step - accuracy: 0.9856 - loss: 0.0575 - val_accuracy: 0.6300 - val_loss: 3.5459
       Epoch 183/200
       25/25
                                 4s 139ms/step - accuracy: 0.9792 - loss: 0.0598 - val_accuracy: 0.6250 - val_loss: 3.2964
       Epoch 184/200
      25/25
                                 4s 140ms/step - accuracy: 0.9887 - loss: 0.0531 - val_accuracy: 0.6150 - val_loss: 3.4861
       Epoch 185/200
                                 - 4s 143ms/step - accuracy: 0.9784 - loss: 0.0492 - val_accuracy: 0.6350 - val_loss: 3.3608
       25/25
       Epoch 186/200
      25/25
                                 - 4s 141ms/step - accuracy: 0.9810 - loss: 0.0573 - val_accuracy: 0.6200 - val_loss: 3.4178
       Epoch 187/200
       25/25 -
                                 4s 139ms/step - accuracy: 0.9845 - loss: 0.0397 - val_accuracy: 0.6050 - val_loss: 3.8684
       Epoch 188/200
      25/25
                                 - 4s 141ms/step - accuracy: 0.9640 - loss: 0.1283 - val_accuracy: 0.6350 - val_loss: 3.2243
       Epoch 189/200
      25/25
                                 - 4s 140ms/step - accuracy: 0.9854 - loss: 0.0466 - val_accuracy: 0.6200 - val_loss: 4.0521
      Epoch 190/200
      25/25
                                - 3s 136ms/step - accuracy: 0.9810 - loss: 0.0668 - val_accuracy: 0.6050 - val_loss: 3.4468
      Epoch 191/200
       25/25
                                 4s 138ms/step - accuracy: 0.9773 - loss: 0.0749 - val_accuracy: 0.6200 - val_loss: 3.5472
       Epoch 192/200
      25/25
                                 - 4s 145ms/step - accuracy: 0.9875 - loss: 0.0332 - val_accuracy: 0.5800 - val_loss: 4.8018
       Epoch 193/200
       25/25
                                 - 4s 140ms/step - accuracy: 0.9606 - loss: 0.1226 - val_accuracy: 0.6250 - val_loss: 3.5109
       Epoch 194/200
      25/25
                                 3s 139ms/step - accuracy: 0.9761 - loss: 0.0790 - val_accuracy: 0.6300 - val_loss: 4.0045
      Epoch 195/200
       25/25
                                 4s 140ms/step - accuracy: 0.9658 - loss: 0.1862 - val_accuracy: 0.6150 - val_loss: 3.4841
      Epoch 196/200
      25/25
                                 • 4s 141ms/step - accuracy: 0.9821 - loss: 0.0515 - val_accuracy: 0.5900 - val_loss: 3.7746
       Epoch 197/200
      25/25
                                 4s 142ms/step - accuracy: 0.9773 - loss: 0.0986 - val_accuracy: 0.6100 - val_loss: 3.8439
       Epoch 198/200
       25/25
                                 • 4s 141ms/step - accuracy: 0.9911 - loss: 0.0289 - val_accuracy: 0.6600 - val_loss: 3.3768
       Epoch 199/200
       25/25
                                 4s 141ms/step - accuracy: 0.9782 - loss: 0.0690 - val_accuracy: 0.6250 - val_loss: 3.4474
       Epoch 200/200
       25/25
                                 • 4s 143ms/step - accuracy: 0.9942 - loss: 0.0190 - val accuracy: 0.6150 - val loss: 4.5955
In [8]: print(f"Execution time: {elapsed_time:.2f} seconds")
       Execution time: 707.66 seconds
```

In [9]: def append_core_data(score_path, num_cores, elapsed_time):

Check if the file already exists
file_exists = os.path.exists(score_path)

Open the file in append mode
with open(score_path, mode='a', newline='') as file:

```
writer = csv.writer(file)

# If the file is new, write the header
if not file_exists:
    writer.writerow(["Number of Cores", "Elapsed Time"])

# Write the new data
writer.writerow([num_cores, elapsed_time])
```

In [10]: score_path = r"C:\Users\nikhi\OneDrive\Desktop\Final Project\DEEP LEARNING WITH HPSC\core_data.txt"
 append_core_data(score_path, number_of_worker, elapsed_time)