from google.colab import drive drive.mount('/content/gdrive') !unzip gdrive/MyDrive/FaceRecognition/archive.zip Streaming output truncated to the last 5000 lines. inflating: MMAFEDB/valid/neutral/8867Exp6fighting_expression_295.jpg inflating: MMAFEDB/valid/neutral/8870Exp6fighting expression 308.jpg inflating: MMAFEDB/valid/neutral/8871Exp6fighting_expression_308.jpg inflating: MMAFEDB/valid/neutral/8872Exp6fighting_expression_320.jpg inflating: MMAFEDB/valid/neutral/8873Exp6fighting_expression_326.jpg inflating: MMAFEDB/valid/neutral/8875Exp6fighting expression 342.jpg inflating: MMAFEDB/valid/neutral/8876Exp6fighting_expression_346.jpg inflating: MMAFEDB/valid/neutral/8878Exp6fighting_expression_395.jpg inflating: MMAFEDB/valid/neutral/8880Exp6fighting_expression_413.jpg inflating: MMAFEDB/valid/neutral/8881Exp6fighting_expression_45.jpg inflating: MMAFEDB/valid/neutral/8882Exp6fighting expression 46.jpg inflating: MMAFEDB/valid/neutral/8887Exp6fighting_expression_578.jpg inflating: MMAFEDB/valid/neutral/8888Exp6fighting_expression_58.jpg inflating: MMAFEDB/valid/neutral/8891Exp6fighting_expression_592.jpg inflating: MMAFEDB/valid/neutral/8892Exp6fighting_expression_601.jpg inflating: MMAFEDB/valid/neutral/8894Exp6fighting_expression_623.jpg inflating: MMAFEDB/valid/neutral/8895Exp6fighting_expression_623.jpg inflating: MMAFEDB/valid/neutral/8898Exp6fighting_expression_642.jpg inflating: MMAFEDB/valid/neutral/8899Exp6fighting expression 651.jpg inflating: MMAFEDB/valid/neutral/889Exp6angry european 197.jpg inflating: MMAFEDB/valid/neutral/88Exp6angry actor 611.jpg inflating: MMAFEDB/valid/neutral/8900Exp6fighting_expression_667.jpg inflating: MMAFEDB/valid/neutral/8901Exp6fighting expression 679.jpg inflating: MMAFEDB/valid/neutral/8904Exp6fighting_expression_712.jpg inflating: MMAFEDB/valid/neutral/8905Exp6fighting expression 716.jpg inflating: MMAFEDB/valid/neutral/8906Exp6fighting_expression_716.jpg inflating: MMAFEDB/valid/neutral/890Exp6angry_european_197.jpg inflating: MMAFEDB/valid/neutral/8910Exp6fighting expression 752.jpg inflating: MMAFEDB/valid/neutral/8911Exp6fighting_expression_752.jpg inflating: MMAFEDB/valid/neutral/8912Exp6fighting_expression_762.jpg inflating: MMAFEDB/valid/neutral/8913Exp6fighting_expression_762.jpg inflating: MMAFEDB/valid/neutral/8915Exp6fighting_expression_782.jpg inflating: MMAFEDB/valid/neutral/8917Exp6fighting_expression_807.jpg inflating: MMAFEDB/valid/neutral/8918Exp6fighting_expression_818.jpg inflating: MMAFEDB/valid/neutral/8919Exp6fighting_expression_827.jpg inflating: MMAFEDB/valid/neutral/891Exp6angry_european_197.jpg inflating: MMAFEDB/valid/neutral/8920Exp6fighting_expression_869.jpg inflating: MMAFEDB/valid/neutral/8922Exp6fighting_expression_90.jpg inflating: MMAFEDB/valid/neutral/8923Exp6fighting_face_105.jpg inflating: MMAFEDB/valid/neutral/8924Exp6fighting_face_108.jpg inflating: MMAFEDB/valid/neutral/8926Exp6fighting_face_128.jpg inflating: MMAFEDB/valid/neutral/8928Exp6fighting face 153.jpg inflating: MMAFEDB/valid/neutral/892Exp6angry_european_197.jpg inflating: MMAFEDB/valid/neutral/8930Exp6fighting_face_156.jpg inflating: MMAFEDB/valid/neutral/8931Exp6fighting face 162.jpg inflating: MMAFEDB/valid/neutral/8932Exp6fighting_face_174.jpg inflating: MMAFEDB/valid/neutral/8934Exp6fighting face 190.jpg inflating: MMAFEDB/valid/neutral/8935Exp6fighting_face_197.jpg inflating: MMAFEDB/valid/neutral/8937Exp6fighting_face_223.jpg inflating: MMAFEDB/valid/neutral/8938Exp6fighting face 23.jpg inflating: MMAFEDB/valid/neutral/8939Exp6fighting face 239.jpg inflating: MMAFEDB/valid/neutral/893Exp6angry european 197.jpg inflating: MMAFEDB/valid/neutral/8940Exp6fighting face 242.jpg inflating: MMAFEDB/valid/neutral/8941Exp6fighting face 246.jpg inflating: MMAFEDB/valid/neutral/8942Exp6fighting_face_267.jpg inflating: MMAFEDB/valid/neutral/8945Exp6fighting face 281.jpg inflating: MMAFEDB/valid/neutral/8946Exp6fighting_face_282.jpg # testing import random import keras from tensorflow.keras import layers from keras.models import Sequential import tensorflow as tf import tensorflow datasets as tfds from keras.layers.core import Dense, Activation, Dropout, Flatten from keras.layers.convolutional import Convolution2D, MaxPooling2D from keras.optimizers import SGD, RMSprop, adam from keras.utils import np utils from sklearn.tree import DecisionTreeClassifier # Import Decision Tree Classifier import matplotlib.pyplot as plt

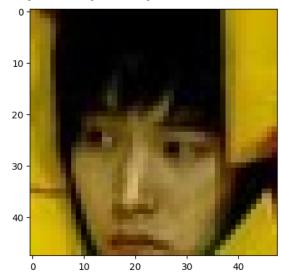
import numpy as np

from PIL import Image

import os import cv2

```
import matplotlib.pyplot as plt
from keras.utils import np utils
from sklearn.model_selection import train_test_split
from keras.applications.vgg16 import VGG16
from keras.applications.vgg16 import VGG16
from keras.models import Model
from tensorflow.keras import layers
from tensorflow.keras.applications.vgg16 import preprocess_input
import tensorflow_datasets as tfds
# setting up training data
path_train = "/content/MMAFEDB/train"
training = []
EMOTIONS = ["angry", "disgust", "fear", "happy", "neutral", "sad", "surprise"]
for emotion in EMOTIONS:
   path = os.path.join(path_train, emotion)
   class num = EMOTIONS.index(emotion)
    count = 0
    for img in os.listdir(path):
       if count > 3200:
         break
        count+=1
        img_array = cv2.imread(os.path.join(path, img))
        img_array = cv2.cvtColor(img_array,cv2.COLOR_RGB2BGR)
        training.append([img array, class num])
# randomizes data
random.shuffle(training)
plt.imshow(training[0][0])
```

<matplotlib.image.AxesImage at 0x7fba195293f0>



```
!ls
    gdrive MMAFEDB sample_data

# assign labels
X = []
y = []

for features, label in training:
    X.append(features)
    y.append(label)
X = np.array(X).reshape(-1, 48, 48, 3)
```

```
# normalizing x
X = X.astype('float32')
x /= 255
Y = np_utils.to_categorical(y, 7)
print(Y[100])
print(Y.shape)
     [0. 0. 0. 1. 0. 0. 0.]
     (22407, 7)
# splitting X and Y
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size = 0.2, random_state = 7)
# model1
input_shape = (48, 48, 3)
model1 = keras.Sequential(
   [
        keras.Input(shape=input_shape),
        layers.Conv2D(32, kernel_size=(3, 3), activation="relu"),
        layers.MaxPooling2D(pool_size=(2, 2)),
        layers.Conv2D(64, kernel_size=(3, 3), activation="relu"),
        layers.MaxPooling2D(pool size=(2, 2)),
        layers.Dropout(0.2),
        layers.Conv2D(128, kernel_size=(3, 3), activation="relu"),
        layers.MaxPooling2D(pool_size=(2, 2)),
        layers.Conv2D(256, kernel_size=(3, 3), activation="relu"),
        layers.MaxPooling2D(pool_size=(2, 2)),
        layers.Flatten(),
        layers.Dropout(0.5),
        layers.Dense(100, activation="sigmoid"),
        layers.Dense(7, activation="sigmoid"),
model1.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
conv2d_8 (Conv2D)		
<pre>max_pooling2d_8 (MaxPooling 2D)</pre>	(None, 23, 23, 32)	0
conv2d_9 (Conv2D)	(None, 21, 21, 64)	18496
<pre>max_pooling2d_9 (MaxPooling 2D)</pre>	(None, 10, 10, 64)	0
dropout_4 (Dropout)	(None, 10, 10, 64)	0
conv2d_10 (Conv2D)	(None, 8, 8, 128)	73856
<pre>max_pooling2d_10 (MaxPoolin g2D)</pre>	(None, 4, 4, 128)	0
conv2d_11 (Conv2D)	(None, 2, 2, 256)	295168
<pre>max_pooling2d_11 (MaxPoolin g2D)</pre>	(None, 1, 1, 256)	0
flatten_2 (Flatten)	(None, 256)	0
dropout_5 (Dropout)	(None, 256)	0
dense_4 (Dense)	(None, 100)	25700
dense_5 (Dense)	(None, 7)	707

```
batch size = 32
epochs = 15
model1.compile(loss="categorical crossentropy", optimizer="adam", metrics=["accuracy"])
model1.fit(X train, y train, batch size=batch size, epochs=epochs, validation data =(X test,y test))
   Epoch 1/15
   561/561 [=========] - 6s 8ms/step - loss: 1.8916 - accuracy: 0.2015 - val loss: 1.8450 - val accuracy: (
   Epoch 2/15
   561/561 [==:
                  =========] - 4s 8ms/step - loss: 1.7733 - accuracy: 0.2789 - val_loss: 1.7126 - val_accuracy: (
   Epoch 3/15
   561/561 [===
                  ========== ] - 4s 7ms/step - loss: 1.7070 - accuracy: 0.3135 - val_loss: 1.6551 - val_accuracy: (
   Epoch 4/15
   561/561 [=========] - 4s 7ms/step - loss: 1.6396 - accuracy: 0.3507 - val loss: 1.5834 - val accuracy: (
   Epoch 5/15
   561/561 [=====
               Epoch 6/15
   561/561 [=========] - 4s 8ms/step - loss: 1.5153 - accuracy: 0.4099 - val_loss: 1.5074 - val_accuracy: (
   Epoch 7/15
   561/561 r==
                  ==========] - 3s 6ms/step - loss: 1.4699 - accuracy: 0.4311 - val_loss: 1.4586 - val_accuracy: (
   Epoch 8/15
   561/561 [=========] - 3s 6ms/step - loss: 1.4377 - accuracy: 0.4460 - val_loss: 1.4278 - val_accuracy: (
   Epoch 9/15
   561/561 [====
               Epoch 10/15
   561/561 [===========] - 4s 7ms/step - loss: 1.3575 - accuracy: 0.4763 - val loss: 1.4011 - val accuracy: (
   Epoch 11/15
   561/561 [=========] - 3s 6ms/step - loss: 1.3272 - accuracy: 0.4899 - val loss: 1.4026 - val accuracy: (
   Epoch 12/15
   561/561 [============] - 4s 7ms/step - loss: 1.2958 - accuracy: 0.4979 - val_loss: 1.3896 - val_accuracy: (
   Epoch 13/15
   561/561 [==========] - 4s 7ms/step - loss: 1.2603 - accuracy: 0.5194 - val_loss: 1.3873 - val_accuracy: (
   Epoch 14/15
   561/561 [===
                 Epoch 15/15
   561/561 [===========] - 4s 7ms/step - loss: 1.2029 - accuracy: 0.5444 - val loss: 1.4260 - val accuracy: (
   <keras.callbacks.History at 0x7fb9c815bd30>
path = "/content/gdrive/MyDrive/EmotionRecognition"
```

model1.save(path)

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op, _jit_convolution_op, _jit_convoluti