```
# unzipping dataset
from google.colab import drive
drive.mount('/content/gdrive')
!unzip gdrive/MyDrive/FaceRecognition/archive2.zip
      inflating: __MACOSX/archive2/UTKFace/._70_0_0_20170111203945060.jpg.chip.jpg
      inflating: archive2/UTKFace/7_1_0_20170109205356415.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._7_1_0_20170109205356415.jpg.chip.jpg
      inflating: archive2/UTKFace/6_0_4_20161221202547897.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._6_0_4_20161221202547897.jpg.chip.jpg
      inflating: archive2/UTKFace/1 1 0 20170109194432489.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._1_1_0_20170109194432489.jpg.chip.jpg
      inflating: archive2/UTKFace/8_1_0_20170109200918322.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._8_1_0_20170109200918322.jpg.chip.jpg
      inflating: archive2/UTKFace/30_0_0_20170104201841050.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._30_0_0_20170104201841050.jpg.chip.jpg
      inflating: archive2/UTKFace/29_0_3_20170104220737116.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._29_0_3_20170104220737116.jpg.chip.jpg
      inflating: archive2/UTKFace/24_1_3_20170104231911026.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._24_1_3_20170104231911026.jpg.chip.jpg
      inflating: archive2/UTKFace/26 1 2 20170116184131117.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._26_1_2_20170116184131117.jpg.chip.jpg
      inflating: archive2/UTKFace/27 1 0 20170116173136538.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._27_1_0_20170116173136538.jpg.chip.jpg
      inflating: archive2/UTKFace/54 1 0 20170117191523029.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._54_1_0_20170117191523029.jpg.chip.jpg
      inflating: archive2/UTKFace/39_1_2_20170103182754641.jpg.chip.jpg
      inflating: MACOSX/archive2/UTKFace/. 39 1 2 20170103182754641.jpg.chip.jpg
      inflating: archive2/UTKFace/35_1_0_20170117154643574.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._35_1_0_20170117154643574.jpg.chip.jpg
      inflating: archive2/UTKFace/85_1_0_20170110181946121.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._85_1_0_20170110181946121.jpg.chip.jpg
      inflating: archive2/UTKFace/53_0_4_20170109012743281.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._53_0_4_20170109012743281.jpg.chip.jpg
      inflating: archive2/UTKFace/26 0 3 20170117154627472.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._26_0_3_20170117154627472.jpg.chip.jpg
      inflating: archive2/UTKFace/25 1 3 20170104222459455.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._25_1_3_20170104222459455.jpg.chip.jpg
      inflating: archive2/UTKFace/57_0_0_20170112203923347.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._57_0_0_20170112203923347.jpg.chip.jpg
      inflating: archive2/UTKFace/26_1_2_20170116184831737.jpg.chip.jpg
      inflating: MACOSX/archive2/UTKFace/. 26 1 2 20170116184831737.jpg.chip.jpg
      inflating: archive2/UTKFace/14_1_0_20170109203625481.jpg.chip.jpg
      inflating: MACOSX/archive2/UTKFace/._14_1_0_20170109203625481.jpg.chip.jpg
      inflating: archive2/UTKFace/33_0_1_20170114034227427.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._33_0_1_20170114034227427.jpg.chip.jpg
      inflating: archive2/UTKFace/54_0_0_20170116193518166.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._54_0_0_20170116193518166.jpg.chip.jpg
      inflating: archive2/UTKFace/30 1 0 20170117181416245.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._30_1_0_20170117181416245.jpg.chip.jpg
      inflating: archive2/UTKFace/89_0_2_20170112220351776.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._89_0_2_20170112220351776.jpg.chip.jpg
      inflating: archive2/UTKFace/51_0_0_20170117160748775.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._51_0_0_20170117160748775.jpg.chip.jpg
      inflating: archive2/UTKFace/24_1_0_20170116222405565.jpg.chip.jpg
      inflating: MACOSX/archive2/UTKFace/. 24 1 0 20170116222405565.jpg.chip.jpg
      inflating: archive2/UTKFace/8_1_0_20170109202318713.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._8_1_0_20170109202318713.jpg.chip.jpg
      inflating: archive2/UTKFace/9 0 0 20170110225030430.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._9_0_0_20170110225030430.jpg.chip.jpg
      inflating: archive2/UTKFace/36_0_1_20170113182511607.jpg.chip.jpg
      inflating: __MACOSX/archive2/UTKFace/._36_0_1_20170113182511607.jpg.chip.jpg
      inflating: archive2/UTKFace/69_1_0_20170110141835631.jpg.chip.jpg
# import modules
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.pvplot as plt
import numpy as np
import os
import cv2
```

```
from PIL import Image
 import matplotlib.pyplot as plt
 from keras.utils import np_utils
{\tt from \ sklearn.model\_selection \ import \ train\_test\_split}
 from keras.applications.vgg16 import VGG16
{\tt 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
path = "/content/archive2/UTKFace/"
amounts = [0,0,0,0,0]
training = []
count = 0
 for img in os.listdir(path):
                  if img != ".DS_Store":
                          classification = img.split("_")
                          category = classification[2]
                           \texttt{if} \cdot \texttt{category} \cdot \texttt{!=} \cdot \texttt{"20170116174525125.jpg.chip.jpg"} \cdot \texttt{and} \cdot \texttt{category} \cdot \texttt{!=} \cdot \texttt{"20170109142408075.jpg.chip.jpg"} : \texttt{and} \cdot \texttt{and} \cdot \texttt{category} \cdot \texttt{!=} \cdot \texttt{"20170109142408075.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.chip.jpg.ch
                                 category = int(category)
                                 if amounts[category] < 2000:</pre>
                                       amounts[category] +=1
                                        img_array = cv2.imread(os.path.join(path, img))
                                       img_array = cv2.cvtColor(img_array,cv2.COLOR_RGB2BGR)
                                       imga = cv2.resize(img_array,dsize = (100,100), interpolation = cv2.INTER_CUBIC)
                                       training.append([imga, category])
# randomizes data
 random.shuffle(training)
plt.imshow(training[0][0])
print(amounts)
                [2000, 2000, 2000, 2000, 1692]
                       0
                   20
                    40
                   60
                   80
                                                        20
                                                                                                                 60
                                                                                                                                              80
X = []
 for features, label in training:
            X.append(features)
             y.append(label)
Xarray = np.array(X).reshape(-1, 100, 100, 3)
Xarray.shape
Xarraydiv = Xarray/ 255
```

```
Xarray = np.array(X).reshape(-1, 100, 100, 3)
Xarray.shape
Xarraydiv = Xarraydiv.astype('float32')
nexty = np_utils.to_categorical(y,5)

X_train, X_test, y_train, y_test = train_test_split(Xarraytype,nexty,test_size = .2, random_state = 5)
input_shape = (100,100,3)
model6 = keras.Sequential(
```

```
[
        keras.Input(shape=input shape),
        layers.RandomFlip("horizontal"),
        layers.Conv2D(32, kernel_size=(3, 3), activation="relu"),
        layers.MaxPooling2D(pool_size=(2, 2)),
        layers.Dropout(0.2),
        layers.Conv2D(64, kernel size=(3, 3), activation="relu"),
        layers.MaxPooling2D(pool_size=(2, 2)),
        layers.RandomRotation(0.1),
        layers.Conv2D(128,kernel_size=(3, 3), activation="relu"),
        layers.MaxPooling2D(pool_size=(2,2)),
        layers.Dropout(0.2),
        layers.Conv2D(256,kernel_size=(3,3), activation = "relu"),
        layers.MaxPooling2D(pool_size=(2,2)),
        layers.Conv2D(512,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(5, activation="sigmoid"),
model6.summary()
```

Model: "sequential"

model: sequential		
Layer (type)	Output Shape	Param #
random_flip (RandomFlip)		0
conv2d (Conv2D)	(None, 98, 98, 32)	896
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 49, 49, 32)	0
dropout (Dropout)	(None, 49, 49, 32)	0
conv2d_1 (Conv2D)	(None, 47, 47, 64)	18496
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(None, 23, 23, 64)	0
<pre>random_rotation (RandomRota tion)</pre>	(None, 23, 23, 64)	0
conv2d_2 (Conv2D)	(None, 21, 21, 128)	73856
<pre>max_pooling2d_2 (MaxPooling 2D)</pre>	(None, 10, 10, 128)	0
dropout_1 (Dropout)	(None, 10, 10, 128)	0
conv2d_3 (Conv2D)	(None, 8, 8, 256)	295168
<pre>max_pooling2d_3 (MaxPooling 2D)</pre>	(None, 4, 4, 256)	0
conv2d_4 (Conv2D)	(None, 2, 2, 512)	1180160
<pre>max_pooling2d_4 (MaxPooling 2D)</pre>	(None, 1, 1, 512)	0
flatten (Flatten)	(None, 512)	0
dropout_2 (Dropout)	(None, 512)	0
dense (Dense)	(None, 100)	51300
dense_1 (Dense)	(None, 5)	505

Total params: 1,620,381 Trainable params: 1,620,381 Non-trainable params: 0

```
model6.compile(loss="categorical_crossentropy", optimizer="adam", metrics=["accuracy"])
model6.fit(X_train, y_train, batch_size=batch_size, epochs=epochs, validation_data =(X_test,y_test))
    Epoch 1/20
    243/243 [==
                              =======] - 9s 24ms/step - loss: 0.8421 - accuracy: 0.6770 - val_loss: 0.8718 - val_accuracy:
    Epoch 2/20
    243/243 [==
                              :=======] - 5s 22ms/step - loss: 0.8397 - accuracy: 0.6819 - val_loss: 0.8895 - val_accuracy:
    Epoch 3/20
    243/243 [============== ] - 5s 21ms/step - loss: 0.8254 - accuracy: 0.6871 - val loss: 0.8250 - val accuracy:
    Epoch 4/20
    243/243 [==
                             ========] - 5s 20ms/step - loss: 0.8067 - accuracy: 0.6919 - val_loss: 0.8198 - val_accuracy:
    Epoch 5/20
    243/243 [=====
                         =========== | - 5s 21ms/step - loss: 0.8096 - accuracy: 0.6957 - val loss: 0.8755 - val accuracy:
    Epoch 6/20
                                 =======] - 5s 21ms/step - loss: 0.7975 - accuracy: 0.6938 - val_loss: 0.8301 - val accuracy:
    243/243 [==
    Epoch 7/20
    243/243 [==:
                             ========1 - 5s 22ms/step - loss: 0.8045 - accuracy: 0.6944 - val loss: 0.8280 - val accuracy:
    Epoch 8/20
                                  =======] - 5s 20ms/step - loss: 0.7898 - accuracy: 0.6986 - val_loss: 0.8608 - val_accuracy:
    243/243 [==
    Epoch 9/20
    243/243 [=================== ] - 5s 20ms/step - loss: 0.7818 - accuracy: 0.7053 - val_loss: 0.8096 - val_accuracy:
    Epoch 10/20
    243/243 [============== ] - 5s 21ms/step - loss: 0.7716 - accuracy: 0.7147 - val loss: 0.8170 - val accuracy:
    Epoch 11/20
    243/243 [===
                            ========= ] - 5s 20ms/step - loss: 0.7759 - accuracy: 0.7066 - val loss: 0.7924 - val accuracy:
    Epoch 12/20
    243/243 [============= ] - 5s 21ms/step - loss: 0.7571 - accuracy: 0.7103 - val loss: 0.8215 - val accuracy:
    Epoch 13/20
    243/243 [===
                                   ======] - 5s 21ms/step - loss: 0.7606 - accuracy: 0.7153 - val_loss: 0.8434 - val_accuracy:
    Epoch 14/20
    243/243 [=================== ] - 5s 21ms/step - loss: 0.7527 - accuracy: 0.7140 - val_loss: 0.8231 - val_accuracy:
    Epoch 15/20
    243/243 [===
                             ========] - 5s 20ms/step - loss: 0.7456 - accuracy: 0.7177 - val_loss: 0.8443 - val_accuracy:
    Epoch 16/20
    243/243 [=====
                       ==========] - 5s 22ms/step - loss: 0.7440 - accuracy: 0.7178 - val_loss: 0.8115 - val_accuracy:
    Epoch 17/20
    243/243 [===
                                ======= ] - 5s 21ms/step - loss: 0.7341 - accuracy: 0.7232 - val loss: 0.8724 - val accuracy:
    Epoch 18/20
    243/243 [===
                                ======== ] - 5s 22ms/step - loss: 0.7197 - accuracy: 0.7269 - val loss: 0.8117 - val accuracy:
    Epoch 19/20
    243/243 [===
                                ======== ] - 5s 20ms/step - loss: 0.7328 - accuracy: 0.7294 - val loss: 0.8283 - val accuracy:
    Epoch 20/20
                                =======] - 5s 20ms/step - loss: 0.7077 - accuracy: 0.7318 - val_loss: 0.8005 - val_accuracy:
    243/243 [==:
    <keras.callbacks.History at 0x7f9f8410e9e0>
```

batch_size = 32
epochs = 20

path = "/content/gdrive/MyDrive/RaceRecognition"

model6.save(path)

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convo

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