```
import cv2
import matplotlib.pyplot as plt
from google.colab import drive
import zipfile
from google.colab import drive
drive.mount('/content/drive/')
     Mounted at /content/drive/
zip_ref = zipfile.ZipFile("/content/drive/MyDrive/Face-Images.zip", 'r')
zip_ref.extractall("/content/")
zip_ref.close()
from keras.preprocessing.image import ImageDataGenerator
\ensuremath{\mathtt{\#}} Defining pre-processing transformations on raw images of training data
# These hyper parameters helps to generate slightly twisted versions
# of the original image, which leads to a better model, since it learns
# on the good and bad mix of images
train_datagen = ImageDataGenerator(
        shear_range=0.1,
        zoom_range=0.1,
        horizontal_flip=True)
#we wont apply any pre processimg on the raw images of the test dataset
test_datagen = ImageDataGenerator()
trainingImagePath = '/content/Face Images/Final Training Images'
testImagePath = '/content/Face Images/Final Testing Images'
# Generating the Training Data
training_set = train_datagen.flow_from_directory(
        trainingImagePath,
        target_size=(64, 64),
        batch_size=32,
        class_mode='categorical')
     Found 256 images belonging to 17 classes.
# Generating the Testing Data
test_set = test_datagen.flow_from_directory(
        testImagePath,
        target_size=(64, 64),
        batch_size=32,
        class_mode='categorical')
     Found 66 images belonging to 17 classes.
# Printing class labels for each face
test_set.class_indices
     {'face1': 0,
       'face10': 1,
      'face11': 2,
      'face12': 3,
      'face13': 4,
      'face14': 5,
      'face15': 6,
      'face16': 7,
      'face17': 8,
      'face2': 9,
      'face3': 10,
      'face4': 11,
      'face5': 12,
      'face6': 13,
'face7': 14,
'face8': 15,
      'face9': 16}
```

```
# class_indices have the numeric tag for each face
TrainClasses=training_set.class_indices
# Storing the face and the numeric tag for future reference
ResultMap={}
for faceValue,faceName in zip(TrainClasses.values(),TrainClasses.keys()):
    ResultMap[faceValue]=faceName
# Saving the face map for future reference
import pickle
with open("ResultsMap.pkl", 'wb') as fileWriteStream:
    pickle.dump(ResultMap, fileWriteStream)
# The model will give answer as a numeric tag
# This mapping will help to get the corresponding face name for it
print("Mapping of Face and its ID", ResultMap)
# The number of neurons for the output layer is equal to the number of faces
OutputNeurons=len(ResultMap)
print('\n The Number of output neurons: ', OutputNeurons)
     Mapping of Face and its ID {0: 'face1', 1: 'face10', 2: 'face11', 3: 'face12', 4: 'face13', 5: 'face14', 6: 'face15', 7: 'face16',
      The Number of output neurons: 17
# so in our CNN model we would have:
#2 hideen convolutional layers
#2 hidden pooling layers
#16 neurons in the output layer since we have 17 classes
# and 1 flattening layer
from keras.models import Sequential
from keras.layers import Convolution2D
from keras.layers import MaxPool2D
from keras.layers import Flatten
from keras.layers import Dense
'''Initializing the Convolutional Neural Network'''
classifier= Sequential()
''' STEP--1 Convolution
# Adding the first layer of CNN
# we are using the format (64,64,3) because we are using TensorFlow backend
# It means 3 matrix of size (64X64) pixels representing Red, Green and Blue components of pixels
classifier.add(Convolution2D(32, kernel_size=(5, 5), strides=(1, 1), input_shape=(64,64,3), activation='relu'))
'''STEP--2 MAX Pooling'''
classifier.add(MaxPool2D(pool_size=(2,2)))
''' ADDITIONAL LAYER of CONVOLUTION for better accuracy'''
classifier.add(Convolution2D(64, kernel_size=(5, 5), strides=(1, 1), activation='relu'))
classifier.add(MaxPool2D(pool_size=(2,2)))
''' STEP--3 FLattening'''
classifier.add(Flatten())
'''STEP--4 Fully Connected Neural Network'''
classifier.add(Dense(64, activation='relu'))
classifier.add(Dense(OutputNeurons, activation='softmax'))
'''Compiling the CNN'''
#classifier.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
classifier.compile(loss='categorical_crossentropy', optimizer = 'adam', metrics=["accuracy"])
# Starting the model training
classifier.fit(
    training_set,
    steps_per_epoch=8,
                              #number of steps per epoch = (Total number of training samples / Batch size), here i have 244 training i
    epochs=15,
    validation_data=test_set,
    validation_steps=10
)
     Epoch 1/15
     8/8 [============] - ETA: 0s - loss: 92.6023 - accuracy: 0.0859 WARNING:tensorflow:Your input ran out of data; in
```

```
8/8 [==============] - 6s 563ms/step - loss: 92.6023 - accuracy: 0.0859 - val_loss: 10.5710 - val_accuracy: 0.0455
    Epoch 2/15
    8/8 [================= ] - 3s 412ms/step - loss: 4.2431 - accuracy: 0.1211
    Epoch 3/15
    8/8 [=============== ] - 5s 659ms/step - loss: 2.6973 - accuracy: 0.1836
    Epoch 4/15
    8/8 [=============== ] - 3s 397ms/step - loss: 2.4474 - accuracy: 0.2852
    Epoch 5/15
    8/8 [============= ] - 4s 426ms/step - loss: 2.2642 - accuracy: 0.3242
    Epoch 6/15
    8/8 [=============== ] - 5s 619ms/step - loss: 1.8446 - accuracy: 0.4727
    Epoch 7/15
    8/8 [================= ] - 4s 454ms/step - loss: 1.2340 - accuracy: 0.6289
    Epoch 8/15
    8/8 [============== ] - 3s 404ms/step - loss: 0.6389 - accuracy: 0.8008
    Epoch 9/15
    8/8 [=============== ] - 4s 517ms/step - loss: 0.3629 - accuracy: 0.8867
    Epoch 10/15
    8/8 [============== ] - 4s 493ms/step - loss: 0.2584 - accuracy: 0.9297
    Epoch 11/15
    Epoch 12/15
    8/8 [================ ] - 4s 465ms/step - loss: 0.2003 - accuracy: 0.9414
    Epoch 13/15
    8/8 [=============== ] - 4s 427ms/step - loss: 0.1248 - accuracy: 0.9648
    Epoch 14/15
    8/8 [============== ] - 4s 441ms/step - loss: 0.0588 - accuracy: 0.9922
    Epoch 15/15
    8/8 [=============== ] - 4s 411ms/step - loss: 0.0548 - accuracy: 0.9844
    <keras.src.callbacks.History at 0x7dfea3438880>
from google.colab.patches import cv2_imshow
 ''Making single predictions'
import numpy as np
from keras.preprocessing import image
ImagePath='/content/Face Images/Final Testing Images/face17/IMG_20190606_145627.jpg'
```

test_image=image.load_img(ImagePath,target_size=(64, 64))

 $image path for output = r' \underline{/content/Face} \ \ Images/Final \ \ Testing \ \ Images/face 17/IMG_20190606_145627.jpg'$

test_image=image.img_to_array(test_image)
test_image=np.expand_dims(test_image,axis=0)
result=classifier.predict(test_image,verbose=0)

imageforoutput = cv2.imread(imagepathforoutput)

print('Prediction is: ',ResultMap[np.argmax(result)])

#print(training_set.class_indices)

cv2 imshow(imageforoutput)



Prediction is: face17