

Installation

Installing DeepFace

```
!pip install deepface
```

Collecting deepface

Downloading deepface-0.0.68-py3-none-any.whl (61 kB)

Requirement already satisfied: Pillow>=5.2.0 in /usr/local/lib/python3.7/dist-packages (from deepface) (7.1.2)

Requirement already satisfied: tensorflow>=1.9.0 in /usr/local/lib/python3.7/dist-packages (from deepface) (2.6.0)

Collecting mtcnn>=0.1.0

Downloading mtcnn-0.1.1-py3-none-any.whl (2.3 MB)

Requirement already satisfied: tqdm>=4.30.0 in /usr/local/lib/python3.7/dist-packages (from deepface) (4.62.0)

Requirement already satisfied: keras>=2.2.0 in /usr/local/lib/python3.7/dist-packages (from deepface) (2.6.0)

Collecting retina-face>=0.0.1

Downloading retina_face-0.0.5-py3-none-any.whl (14 kB)

Requirement already satisfied: Flask>=1.1.2 in /usr/local/lib/python3.7/dist-packages (from deepface) (1.1.4)

Requirement already satisfied: pandas>=0.23.4 in /usr/local/lib/python3.7/dist-packages (from deepface) (1.1.5)

Collecting gdown>=3.10.1

Downloading gdown-3.13.1.tar.gz (10 kB)

Installing build dependencies ... ents to build wheel ...

etadata ... Requirement already satisfied: opencv-python>=3.4.4 in /usr/local/lib/python3.7/dist-packages (from deepface) (4.1.2.30)

Requirement already satisfied: numpy>=1.14.0 in /usr/local/lib/python3.7/dist-packages (from deepface) (1.19.5)

Requirement already satisfied: click<8.0,>=5.1 in /usr/local/lib/python3.7/dist-packages (from Flask>=1.1.2->deepface) (7.1.2)

Requirement already satisfied: itsdangerous<2.0,>=0.24 in /usr/local/lib/python3.7/dist-packages (from Flask>=1.1.2->deepface) (1.1.0)

Requirement already satisfied: Werkzeug<2.0,>=0.15 in /usr/local/lib/python3.7/dist-packages (from Flask>=1.1.2->deepface) (1.0.1)

Requirement already satisfied: Jinja2<3.0,>=2.10.1 in /usr/local/lib/python3.7/dist-packages (from Flask>=1.1.2->deepface) (2.11.3)

Requirement already satisfied: filelock in /usr/local/lib/python3.7/dist-packages (from gdown>=3.10.1->deepface) (3.0.12)

Requirement already satisfied: requests[socks]>=2.12.0 in /usr/local/lib/python3.7/dist-packages (from gdown>=3.10.1->deepface)

(2.23.0)
Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from gdown>=3.10.1->deepface) (1.15.0)
Requirement already satisfied: MarkupSafe>=0.23 in /usr/local/lib/python3.7/dist-packages (from Jinja2<3.0,>=2.10.1->Flask>=1.1.2->deepface) (2.0.1)
Requirement already satisfied: python-dateutil>=2.7.3 in /usr/local/lib/python3.7/dist-packages (from pandas>=0.23.4->deepface) (2.8.2)
Requirement already satisfied: pytz>=2017.2 in /usr/local/lib/python3.7/dist-packages (from pandas>=0.23.4->deepface) (2018.9)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests[socks]>=2.12.0->gdown>=3.10.1->deepface) (2021.5.30)
Requirement already satisfied: urllib3!=1.25.0,!1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-packages (from requests[socks]>=2.12.0->gdown>=3.10.1->deepface) (1.24.3)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests[socks]>=2.12.0->gdown>=3.10.1->deepface) (2.10)
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests[socks]>=2.12.0->gdown>=3.10.1->deepface) (3.0.4)
Requirement already satisfied: PySocks!=1.5.7,>=1.5.6 in /usr/local/lib/python3.7/dist-packages (from requests[socks]>=2.12.0->gdown>=3.10.1->deepface) (1.7.1)
Requirement already satisfied: termcolor~=1.1.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0->deepface) (1.1.0)
Requirement already satisfied: typing-extensions~=3.7.4 in /usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0->deepface) (3.7.4.3)
Requirement already satisfied: wheel~=0.35 in /usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0->deepface) (0.37.0)
Requirement already satisfied: opt-einsum~=3.3.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0->deepface) (3.3.0)
Requirement already satisfied: gast==0.4.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0->deepface) (0.4.0)
Requirement already satisfied: tensorboard~=2.6 in /usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0->deepface) (2.6.0)
Requirement already satisfied: astunparse~=1.6.3 in /usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0->deepface) (1.6.3)
Requirement already satisfied: absl-py~=0.10 in

```
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (0.12.0)
Requirement already satisfied: clang~=5.0 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (5.0)
Requirement already satisfied: wrapt~=1.12.1 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (1.12.1)
Requirement already satisfied: keras-preprocessing~=1.1.2 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (1.1.2)
Requirement already satisfied: grpcio<2.0,>=1.37.0 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (1.39.0)
Requirement already satisfied: tensorflow-estimator~=2.6 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (2.6.0)
Requirement already satisfied: google-pasta~=0.2 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (0.2.0)
Requirement already satisfied: h5py~=3.1.0 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (3.1.0)
Requirement already satisfied: flatbuffers~=1.12.0 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (1.12)
Requirement already satisfied: protobuf>=3.9.2 in
/usr/local/lib/python3.7/dist-packages (from tensorflow>=1.9.0-
>deepface) (3.17.3)
Requirement already satisfied: cached-property in
/usr/local/lib/python3.7/dist-packages (from h5py~=3.1.0-
>tensorflow>=1.9.0->deepface) (1.5.2)
Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in
/usr/local/lib/python3.7/dist-packages (from tensorboard~=2.6-
>tensorflow>=1.9.0->deepface) (0.4.5)
Requirement already satisfied: markdown>=2.6.8 in
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>tensorflow>=1.9.0->deepface) (3.3.4)
Requirement already satisfied: tensorboard-data-server<0.7.0,>=0.6.0
in /usr/local/lib/python3.7/dist-packages (from tensorboard~=2.6-
>tensorflow>=1.9.0->deepface) (0.6.1)
Requirement already satisfied: google-auth<2,>=1.6.3 in
/usr/local/lib/python3.7/dist-packages (from tensorboard~=2.6-
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Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in
/usr/local/lib/python3.7/dist-packages (from tensorboard~=2.6-
>tensorflow>=1.9.0->deepface) (1.8.0)
Requirement already satisfied: setuptools>=41.0.0 in
/usr/local/lib/python3.7/dist-packages (from tensorboard~=2.6-
```

```

>tensorflow>=1.9.0->deepface) (57.4.0)
Requirement already satisfied: pyasn1-modules>=0.2.1 in
/usr/local/lib/python3.7/dist-packages (from google-auth<2,>=1.6.3-
>tensorboard~=2.6->tensorflow>=1.9.0->deepface) (0.2.8)
Requirement already satisfied: rsa<5,>=3.1.4 in
/usr/local/lib/python3.7/dist-packages (from google-auth<2,>=1.6.3-
>tensorboard~=2.6->tensorflow>=1.9.0->deepface) (4.7.2)
Requirement already satisfied: cachetools<5.0,>=2.0.0 in
/usr/local/lib/python3.7/dist-packages (from google-auth<2,>=1.6.3-
>tensorboard~=2.6->tensorflow>=1.9.0->deepface) (4.2.2)
Requirement already satisfied: requests-oauthlib>=0.7.0 in
/usr/local/lib/python3.7/dist-packages (from google-auth-
oauthlib<0.5,>=0.4.1->tensorboard~=2.6->tensorflow>=1.9.0->deepface)
(1.3.0)
Requirement already satisfied: importlib-metadata in
/usr/local/lib/python3.7/dist-packages (from markdown>=2.6.8-
>tensorboard~=2.6->tensorflow>=1.9.0->deepface) (4.6.4)
Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in
/usr/local/lib/python3.7/dist-packages (from pyasn1-modules>=0.2.1-
>google-auth<2,>=1.6.3->tensorboard~=2.6->tensorflow>=1.9.0->deepface)
(0.4.8)
Requirement already satisfied: oauthlib>=3.0.0 in
/usr/local/lib/python3.7/dist-packages (from requests-oauthlib>=0.7.0-
>google-auth-oauthlib<0.5,>=0.4.1->tensorboard~=2.6-
>tensorflow>=1.9.0->deepface) (3.1.1)
Requirement already satisfied: zipp>=0.5 in
/usr/local/lib/python3.7/dist-packages (from importlib-metadata-
>markdown>=2.6.8->tensorboard~=2.6->tensorflow>=1.9.0->deepface)
(3.5.0)
Building wheels for collected packages: gdown
  Building wheel for gdown (PEP 517) ... e=gdown-3.13.1-py3-none-
any.whl size=9920
sha256=9328d22f889f29fa1ddfd4963ced0695650a242fc72e77bfa3bb90295e81ac4
e
  Stored in directory:
/root/.cache/pip/wheels/f2/8d/0b/2e7e6c725f898bd7ef654b660528e459a4d79
f3a68976ca9fc
Successfully built gdown
Installing collected packages: gdown, retina-face, mtcnn, deepface
  Attempting uninstall: gdown
    Found existing installation: gdown 3.6.4
    Uninstalling gdown-3.6.4:
      Successfully uninstalled gdown-3.6.4
Successfully installed deepface-0.0.68 gdown-3.13.1 mtcnn-0.1.1
retina-face-0.0.5

```

2.2 To Load test data

Since its a pre-trained model we will only test it with some random image.

```
from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
image = "/content/drive/MyDrive/Capstone Project 5/PreTRained Model  
Test Images/faces-of-smiling-happy-children-in-makueni-county-kenya-  
2C8A5PJ.jpg"
```

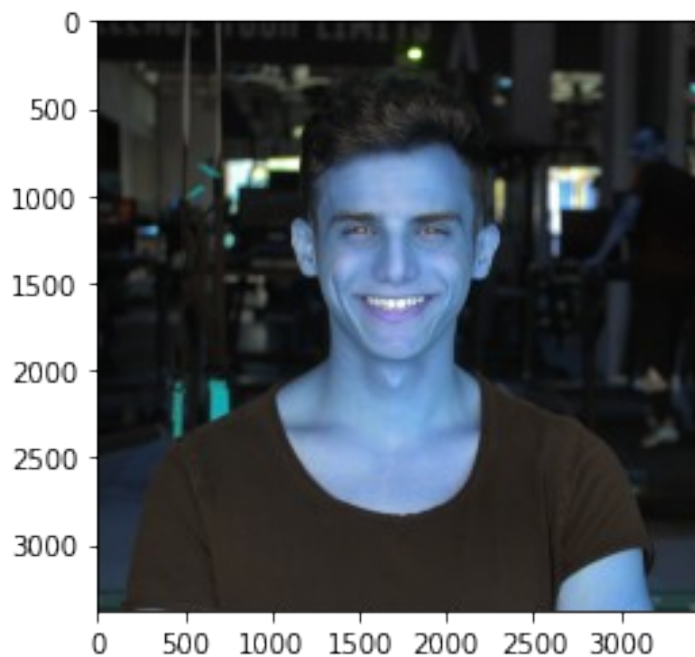
```
# Image Show
import cv2
import matplotlib.pyplot as plt
img_array=cv2.imread(image)
plt.imshow(img_array)
```

<matplotlib.image.AxesImage at 0x7efcc8d5fd90>



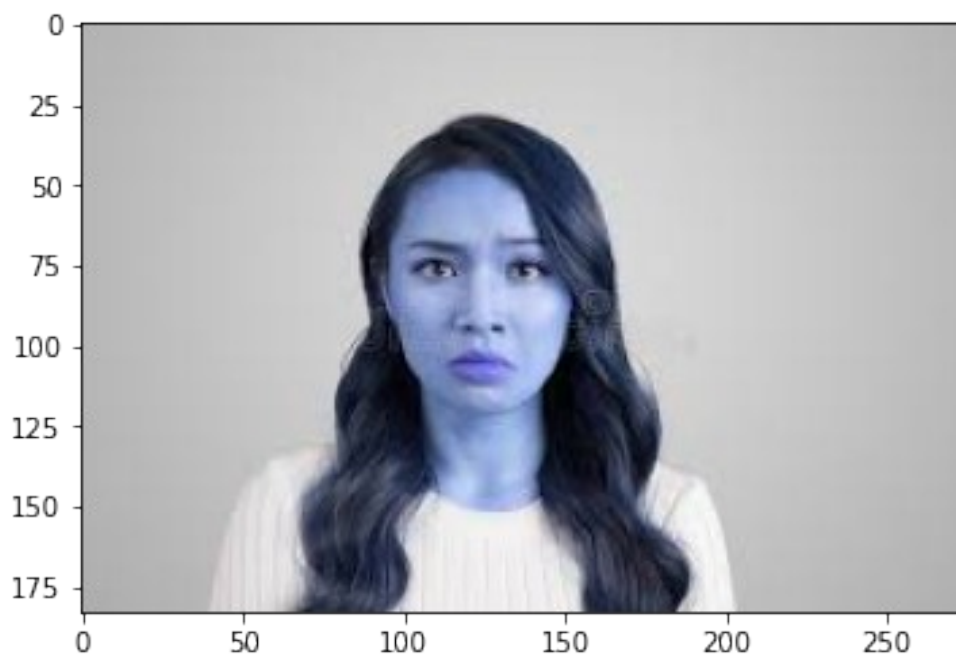
```
image1 = '/content/drive/MyDrive/Capstone Project 5/PreTRained Model  
Test Images/abdullah-ali-lw9I6H4aftw-unsplash.jpg'  
img_array1=cv2.imread(image1)  
plt.imshow(img_array1)
```

<matplotlib.image.AxesImage at 0x7efcc8add050>



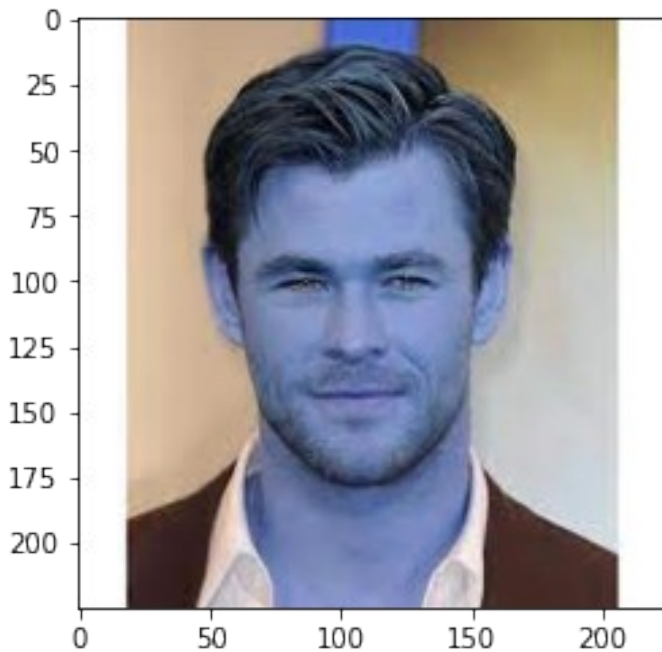
```
image2 = '/content/drive/MyDrive/Capstone Project 5/PreTrained Model  
Test Images/sad face.jpeg'  
img_array2=cv2.imread(image2)  
plt.imshow(img_array2)
```

<matplotlib.image.AxesImage at 0x7efcc99f0fd0>



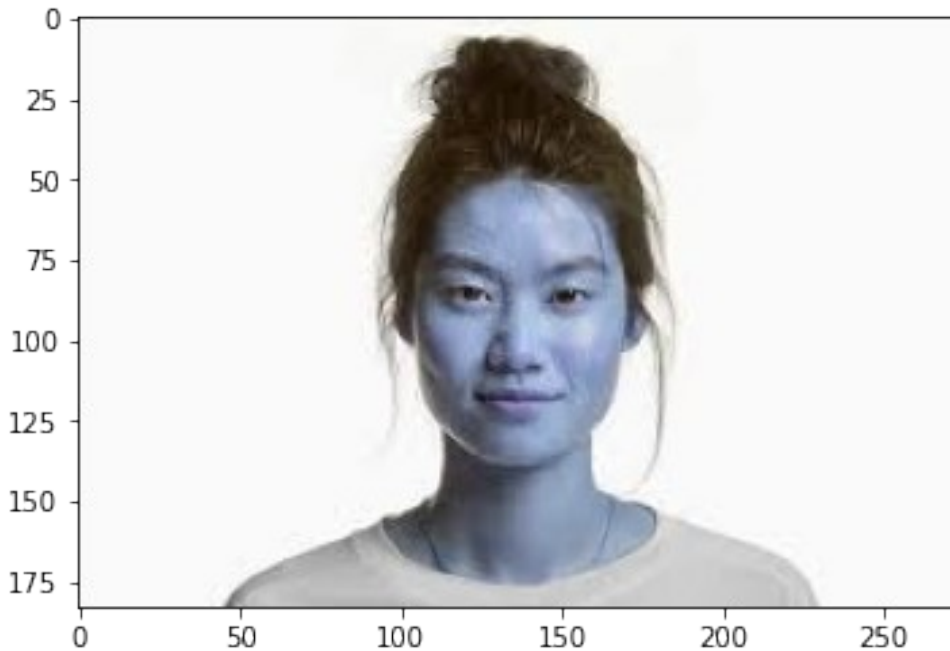
```
image3 = '/content/drive/MyDrive/Capstone Project 5/PreTRained Model  
Test Images/image 3.jpeg'  
img_array3=cv2.imread(image3)  
plt.imshow(img_array3)
```

<matplotlib.image.AxesImage at 0x7efcc99d0790>



```
image4 = '/content/drive/MyDrive/Capstone Project 5/PreTRained Model  
Test Images/image 4.jpeg'  
img_array4=cv2.imread(image4)  
plt.imshow(img_array4)
```

<matplotlib.image.AxesImage at 0x7efcc8d08510>



2.3 Analyze Data

```
obj1 = DeepFace.analyze(img_path = image1, actions = ['age', 'gender',  
'race', 'emotion'])  
print("Result for image1", obj1["age"]," years old  
",obj1["dominant_race"]," ",obj1["dominant_emotion"]," ",  
obj1["gender"])
```

Action: emotion: 100%|██████████| 4/4 [00:06<00:00, 1.74s/it]

Result for image1 28 years old white happy Man

```
obj2 = DeepFace.analyze(img_path = image2, actions = ['age', 'gender',  
'race', 'emotion'])  
print("Result for image2", obj2["age"]," years old  
",obj2["dominant_race"]," ",obj2["dominant_emotion"]," ",  
obj2["gender"])
```

Action: emotion: 100%|██████████| 4/4 [00:01<00:00, 2.29it/s]

Result for image2 31 years old latino hispanic sad Woman

```
obj3 = DeepFace.analyze(img_path = image3, actions = ['age', 'gender',  
'race', 'emotion'])  
print("Result for image3", obj3["age"]," years old  
",obj3["dominant_race"]," ",obj3["dominant_emotion"]," ",  
obj3["gender"])
```



```
Action: emotion: 100%|██████████| 4/4 [00:01<00:00, 2.24it/s]
```

```
Result for image3 23 years old white neutral Man
```

```
obj4 = DeepFace.analyze(img_path = image4, actions = ['age', 'gender',  
'race', 'emotion'])  
print("Result for image4", obj4["age"]," years old  
",obj4["dominant_race"]," ",obj4["dominant_emotion"]," ",  
obj4["gender"])
```

```
Action: emotion: 100%|██████████| 4/4 [00:01<00:00, 2.27it/s]
```

```
Result for image4 24 years old asian neutral Man
```

3. Building Face Emotion Recognition Model

Link Google Drive

```
from google.colab import drive  
drive.mount('/content/drive')
```

```
Mounted at /content/drive
```

Import Necessary Libraries

```
%matplotlib inline  
import matplotlib.pyplot as plt  
  
import numpy as np  
import pandas as pd  
from tensorflow.keras.utils import to_categorical  
from sklearn.model_selection import train_test_split  
from sklearn.metrics import confusion_matrix  
  
from keras.models import Sequential #Initialise our neural network  
model as a sequential network  
from keras.layers import Conv2D #Convolution operation  
from keras.layers import BatchNormalization  
from keras.regularizers import l2  
from keras.layers import Activation #Applies activation function  
from keras.layers import Dropout #Prevents overfitting by randomly  
converting few outputs to zero  
from keras.layers import MaxPooling2D # Maxpooling function  
from keras.layers import Flatten # Converting 2D arrays into a 1D  
linear vector
```

```

from keras.layers import Dense # Regular fully connected neural network
from tensorflow.keras import optimizers
from keras.callbacks import ReduceLROnPlateau, EarlyStopping, TensorBoard, ModelCheckpoint
from sklearn.metrics import accuracy_score

```

##3.3 Load Dataset

```

def load_data(dataset_path): #Run once

    #classes = ['Angry', 'Disgust', 'Fear', 'Happy', 'Sad', 'Surprsie', 'Neutral'] #We will be dealing with seven different types of emotions.

    data = []
    test_data = []
    test_labels = []
    labels = []

    with open(dataset_path, 'r') as file:
        for line_no, line in enumerate(file.readlines()):
            if 0 < line_no <= 35887:
                curr_class, line, set_type = line.split(',')
                image_data = np.asarray([int(x) for x in line.split()]).reshape(48, 48)#Creating a list out of the string then converting it into a 2-Dimensional numpy array.
                image_data = image_data.astype(np.uint8)/255.0

                if (set_type.strip() == 'PrivateTest'):

                    test_data.append(image_data)
                    test_labels.append(curr_class)
                else:
                    data.append(image_data)
                    labels.append(curr_class)

    test_data = np.expand_dims(test_data, -1)
    test_labels = to_categorical(test_labels, num_classes = 7)
    data = np.expand_dims(data, -1)
    labels = to_categorical(labels, num_classes = 7)

    return np.array(data), np.array(labels), np.array(test_data), np.array(test_labels)

```

Splitting of Data

```

dataset_path = "/content/drive/MyDrive/Capstone Project 5/fer2013.csv"

```

```

train_data, train_labels, test_data, test_labels =
load_data(dataset_path)

print("Number of images in Training set:", len(train_data))
print("Number of images in Test set:", len(test_data))

Number of images in Training set: 32298
Number of images in Test set: 3589

```

Model Training

```

#####HYPERPARAMETERS#####
epochs = 75
batch_size = 64
learning_rate = 0.001
#####

model = Sequential()

model.add(Conv2D(64, (3, 3), activation='relu', input_shape=(48, 48,
1), kernel_regularizer=l2(0.01)))
model.add(Conv2D(64, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2,2), strides=(2, 2)))
model.add(Dropout(0.5))

model.add(Conv2D(128, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(128, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(128, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Dropout(0.5))

model.add(Conv2D(256, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(256, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(256, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Dropout(0.5))

model.add(Conv2D(512, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(512, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(512, (3, 3), padding='same', activation='relu'))

```

```

model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Dropout(0.5))

model.add(Flatten())

model.add(Dense(512, activation='relu'))
model.add(Dropout(0.5))

model.add(Dense(256, activation='relu'))
model.add(Dropout(0.5))

model.add(Dense(128, activation='relu'))
model.add(Dropout(0.5))

model.add(Dense(64, activation='relu'))
model.add(Dropout(0.5))

model.add(Dense(7, activation='softmax'))

adam = optimizers.Adam(lr = learning_rate)
model.compile(optimizer = adam, loss = 'categorical_crossentropy',
metrics = ['accuracy'])

print(model.summary())

lr_reducer = ReduceLROnPlateau(monitor='val_loss', factor=0.9,
patience=3)
early_stopper = EarlyStopping(monitor='val_acc', min_delta=0,
patience=6, mode='auto')
checkpointer = ModelCheckpoint('/content/gdrive/My Drive/Colab
Notebooks/Emotion Recognition/Model/weights.hd5', monitor='val_loss',
verbose=1, save_best_only=True)

history = model.fit(
    train_data,
    train_labels,
    epochs = epochs,
    batch_size = batch_size,
    validation_split = 0.2,
    shuffle = True
)

/usr/local/lib/python3.7/dist-packages/keras/optimizer_v2/
optimizer_v2.py:356: UserWarning: The `lr` argument is deprecated, use
`learning_rate` instead.
    "The `lr` argument is deprecated, use `learning_rate` instead.")

Model: "sequential_1"

```

Layer (type)	Output Shape	Param #
conv2d_11 (Conv2D)	(None, 46, 46, 64)	640
conv2d_12 (Conv2D)	(None, 46, 46, 64)	36928
batch_normalization_10 (Batch Normalization)	(None, 46, 46, 64)	256
max_pooling2d_4 (MaxPooling2D)	(None, 23, 23, 64)	0
dropout_8 (Dropout)	(None, 23, 23, 64)	0
conv2d_13 (Conv2D)	(None, 23, 23, 128)	73856
batch_normalization_11 (Batch Normalization)	(None, 23, 23, 128)	512
conv2d_14 (Conv2D)	(None, 23, 23, 128)	147584
batch_normalization_12 (Batch Normalization)	(None, 23, 23, 128)	512
conv2d_15 (Conv2D)	(None, 23, 23, 128)	147584
batch_normalization_13 (Batch Normalization)	(None, 23, 23, 128)	512
max_pooling2d_5 (MaxPooling2D)	(None, 11, 11, 128)	0
dropout_9 (Dropout)	(None, 11, 11, 128)	0
conv2d_16 (Conv2D)	(None, 11, 11, 256)	295168
batch_normalization_14 (Batch Normalization)	(None, 11, 11, 256)	1024
conv2d_17 (Conv2D)	(None, 11, 11, 256)	590080
batch_normalization_15 (Batch Normalization)	(None, 11, 11, 256)	1024
conv2d_18 (Conv2D)	(None, 11, 11, 256)	590080
batch_normalization_16 (Batch Normalization)	(None, 11, 11, 256)	1024
max_pooling2d_6 (MaxPooling2D)	(None, 5, 5, 256)	0
dropout_10 (Dropout)	(None, 5, 5, 256)	0
conv2d_19 (Conv2D)	(None, 5, 5, 512)	1180160
batch_normalization_17 (Batch Normalization)	(None, 5, 5, 512)	2048
conv2d_20 (Conv2D)	(None, 5, 5, 512)	2359808

batch_normalization_18 (Batch Normalization)	(None, 5, 5, 512)	2048
conv2d_21 (Conv2D)	(None, 5, 5, 512)	2359808
batch_normalization_19 (Batch Normalization)	(None, 5, 5, 512)	2048
max_pooling2d_7 (MaxPooling2D)	(None, 2, 2, 512)	0
dropout_11 (Dropout)	(None, 2, 2, 512)	0
flatten_1 (Flatten)	(None, 2048)	0
dense_5 (Dense)	(None, 512)	1049088
dropout_12 (Dropout)	(None, 512)	0
dense_6 (Dense)	(None, 256)	131328
dropout_13 (Dropout)	(None, 256)	0
dense_7 (Dense)	(None, 128)	32896
dropout_14 (Dropout)	(None, 128)	0
dense_8 (Dense)	(None, 64)	8256
dropout_15 (Dropout)	(None, 64)	0
dense_9 (Dense)	(None, 7)	455

=====
 Total params: 9,014,727
 Trainable params: 9,009,223
 Non-trainable params: 5,504

None

Epoch 1/75

404/404 [=====] - 59s 141ms/step - loss: 2.0908 - accuracy: 0.2009 - val_loss: 1.8732 - val_accuracy: 0.2489

Epoch 2/75

404/404 [=====] - 57s 140ms/step - loss: 1.8643 - accuracy: 0.2362 - val_loss: 1.8446 - val_accuracy: 0.2489

Epoch 3/75

404/404 [=====] - 56s 140ms/step - loss: 1.8404 - accuracy: 0.2485 - val_loss: 1.8268 - val_accuracy: 0.2489

Epoch 4/75

404/404 [=====] - 57s 140ms/step - loss: 1.8323 - accuracy: 0.2509 - val_loss: 1.8246 - val_accuracy: 0.2489

Epoch 5/75

```
404/404 [=====] - 57s 140ms/step - loss:
1.8244 - accuracy: 0.2506 - val_loss: 1.8210 - val_accuracy: 0.2489
Epoch 6/75
298/404 [=====>.....] - ETA: 13s - loss: 1.8191 -
accuracy: 0.2553
```

3.6 Evaluation Metrics

To Check Accuracy

```
predicted_test_labels = np.argmax(model.predict(test_data), axis=1)
test_labels = np.argmax(test_labels, axis=1)
print ("Accuracy score = ", accuracy_score(test_labels,
predicted_test_labels))
print ("Accuracy percentage = ", accuracy_score(test_labels,
predicted_test_labels)*100, "%")
```

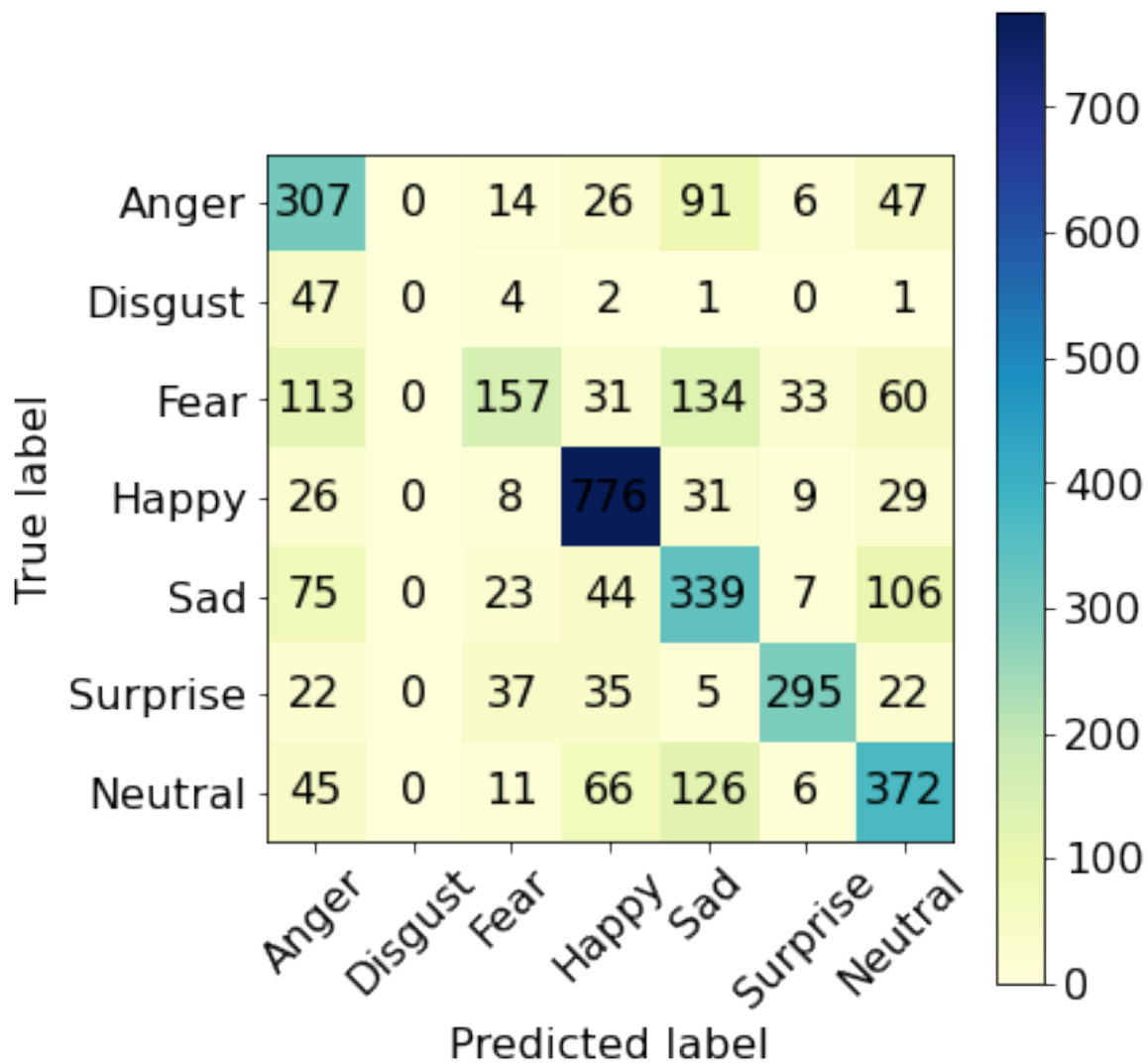
```
Accuracy score = 0.6258010587907495
Accuracy percentage = 62.58010587907496 %
```

Confusion Matrix

```
labels = ['Anger', 'Disgust', 'Fear', 'Happy', 'Sad', 'Surprise',
'Neutral']

def plot_confusion_matrix(y_true, y_pred, cmap=plt.cm.Blues):
    cm = confusion_matrix(y_true, y_pred)
    fig = plt.figure(figsize=(6,6))
    plt.rcParams.update({'font.size': 16})
    ax = fig.add_subplot(111)
    matrix = ax.imshow(cm, interpolation='nearest', cmap=cmap)
    fig.colorbar(matrix)
    for i in range(0,7):
        for j in range(0,7):
            ax.text(j,i,cm[i,j],va='center', ha='center')
    # ax.set_title('Confusion Matrix')
    ticks = np.arange(len(labels))
    ax.set_xticks(ticks)
    ax.set_xticklabels(labels, rotation=45)
    ax.set_yticks(ticks)
    ax.set_yticklabels(labels)
    plt.tight_layout()
    plt.ylabel('True label')
    plt.xlabel('Predicted label')

plot_confusion_matrix(test_labels, predicted_test_labels,
cmap=plt.cm.YlGnBu)
plt.show()
```



Classification report

```
from sklearn.metrics import classification_report
print(classification_report(test_labels, predicted_test_labels,
target_names=labels))
```

	precision	recall	f1-score	support
Anger	0.48	0.63	0.55	491
Disgust	0.00	0.00	0.00	55
Fear	0.62	0.30	0.40	528
Happy	0.79	0.88	0.83	879
Sad	0.47	0.57	0.51	594
Surprise	0.83	0.71	0.76	416
Neutral	0.58	0.59	0.59	626
accuracy			0.63	3589

macro avg	0.54	0.53	0.52	3589
weighted avg	0.63	0.63	0.61	3589

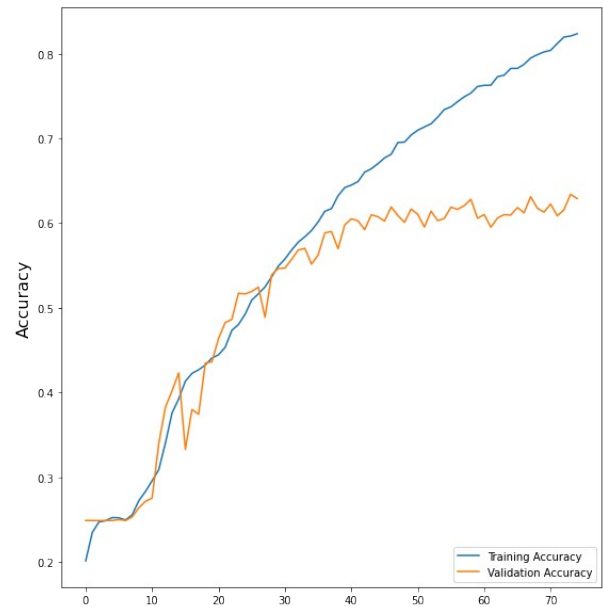
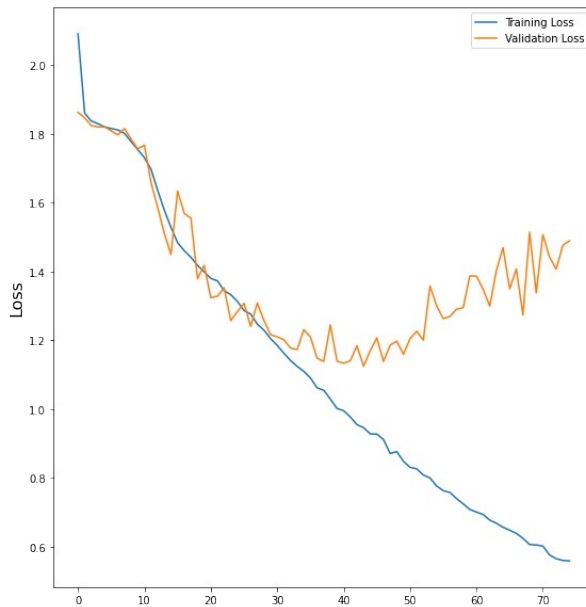
```
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
```

Loss & Accuracy Graph

```
plt.figure(figsize=(20,10))
plt.subplot(1, 2, 1)
plt.suptitle('Optimizer : Adam', fontsize=10)
plt.ylabel('Loss', fontsize=16)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.legend(loc='upper right')

plt.subplot(1, 2, 2)
plt.ylabel('Accuracy', fontsize=16)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend(loc='lower right')
plt.show()
```

Optimizer : Adam



Saving of Model

```
from keras.models import model_from_json
model_json = model.to_json()
with open("/content/drive/MyDrive/Capstone Project 5/model.json", "w")
as json_file:
    json_file.write(model_json)
```

```
model.save_weights("/content/drive/MyDrive/Capstone Project
5/model.hdf5")
print("Saved model to disk")
```

Saved model to disk

```
model.save('/content/drive/MyDrive/Capstone Project 5/FER_model.h5')
print('Model Saved')
```

Model Saved

```
!pip install keras
```

Requirement already satisfied: keras in /usr/local/lib/python3.7/dist-packages (2.6.0)

```
import keras
```

```
saveBestModel = keras.callbacks.ModelCheckpoint('/best_model.hdf5',
monitor='val_acc', verbose=0, save_best_only=True,
save_weights_only=False, mode='auto', period=1, save_freq='epoch')
```

WARNING:tensorflow:`period` argument is deprecated. Please use
`save_freq` to specify the frequency in number of batches seen.

4.To Check our Model on Images

Preparation for emotion recognition using photo

```
def emotion_analysis(emotions):  
    objects = ('angry', 'disgust', 'fear', 'happy', 'sad', 'surprise',  
              'neutral')  
    y_pos = np.arange(len(objects))  
  
    plt.bar(y_pos, emotions, align='center', alpha=0.5)  
    plt.xticks(y_pos, objects)  
    plt.ylabel('percentage')  
    plt.title('emotion')  
  
    plt.show()
```

Model Loading

```
from keras.models import model_from_json  
import numpy as np  
import cv2  
  
def load_model(path):  
    json_file = open(path + 'model.json', 'r')  
    loaded_model_json = json_file.read()  
    json_file.close()  
  
    model = model_from_json(loaded_model_json)  
    model.load_weights(path + "model.h5")  
    print("Loaded model from disk")  
    return model  
  
#model loading  
path = "/content/drive/MyDrive/Capstone Project 5/"  
model = load_model(path)
```

Loaded model from disk

Defining function which will click images

```
##CODE for Capturing an image on Colab from here:  
https://colab.research.google.com/notebook#fileId=10nUy6eFE7XhdfGfAHDCqQxpwueT0j\_N0
```

```

from IPython.display import display, Javascript
from google.colab.output import eval_js
from base64 import b64decode

def take_photo(filename='photo.jpg', quality=0.8):
    js = Javascript('''
        async function takePhoto(quality) {
            const div = document.createElement('div');
            const capture = document.createElement('button');
            capture.textContent = 'Capture';
            div.appendChild(capture);

            const video = document.createElement('video');
            video.style.display = 'block';
            const stream = await navigator.mediaDevices.getUserMedia({video:
true});

            document.body.appendChild(div);
            div.appendChild(video);
            video.srcObject = stream;
            await video.play();

            // Resize the output to fit the video element.

google.colab.output.setIframeHeight(document.documentElement.scrollHei
ght, true);

            // Wait for Capture to be clicked.
            await new Promise((resolve) => capture.onclick = resolve);

            const canvas = document.createElement('canvas');
            canvas.width = video.videoWidth;
            canvas.height = video.videoHeight;
            canvas.getContext('2d').drawImage(video, 0, 0);
            stream.getVideoTracks()[0].stop();
            div.remove();
            return canvas.toDataURL('image/jpeg', quality);
        }
        ''')
    display(js)
    data = eval_js('takePhoto({})'.format(quality))
    binary = b64decode(data.split(',')[1])
    with open(filename, 'wb') as f:
        f.write(binary)
    return filename

```

Use this to click *photo*

```
take_photo()
```

```
<IPython.core.display.Javascript object>
{"type": "string"}
```

Defining Function to crop face and analysing photo taken

```
import cv2

def facecrop(image):
    facedata = '/content/haarcascade_frontalface_alt.xml'
    cascade = cv2.CascadeClassifier(facedata)

    img = cv2.imread(image)

    try:

        minisize = (img.shape[1],img.shape[0])
        miniframe = cv2.resize(img, minisize)

        faces = cascade.detectMultiScale(miniframe)

        for f in faces:
            x, y, w, h = [ v for v in f ]
            cv2.rectangle(img, (x,y), (x+w,y+h), (0,255,0), 2)

            sub_face = img[y:y+h, x:x+w]

            cv2.imwrite('capture.jpg', sub_face)

    except Exception as e:
        print (e)

if __name__ == '__main__':
    facecrop('/content/photo.jpg')

from keras.preprocessing import image
from keras.preprocessing.image import ImageDataGenerator

import numpy as np
import matplotlib.pyplot as plt

file = '/content/photo.jpg'
true_image = image.load_img(file)
img = image.load_img(file, color_mode="grayscale", target_size=(48,
```

```
48))

x = image.img_to_array(img)
x = np.expand_dims(x, axis = 0)

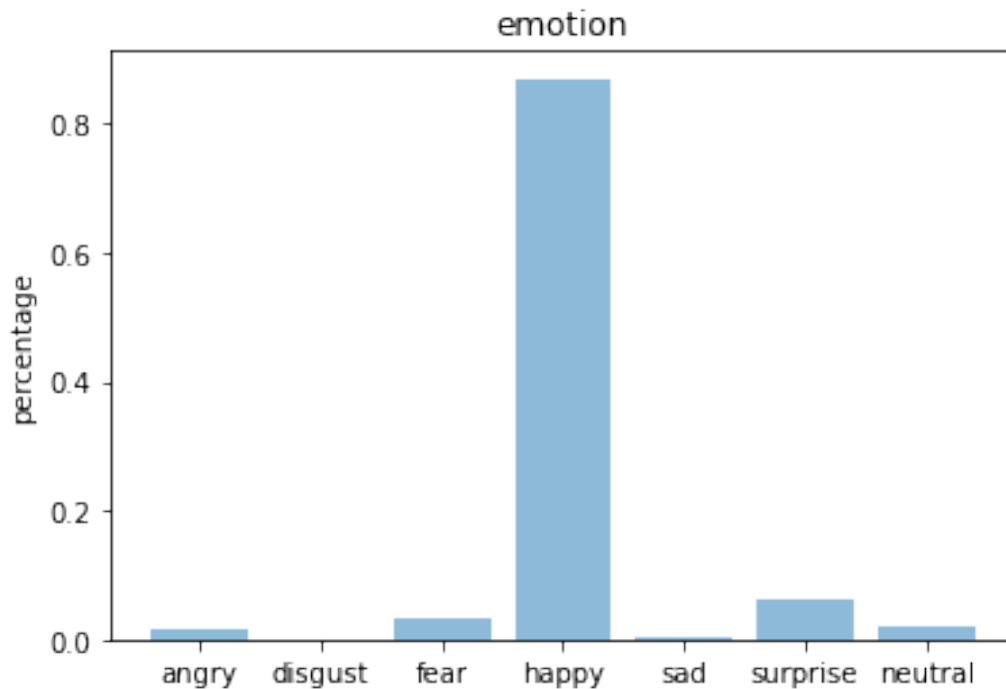
x /= 255

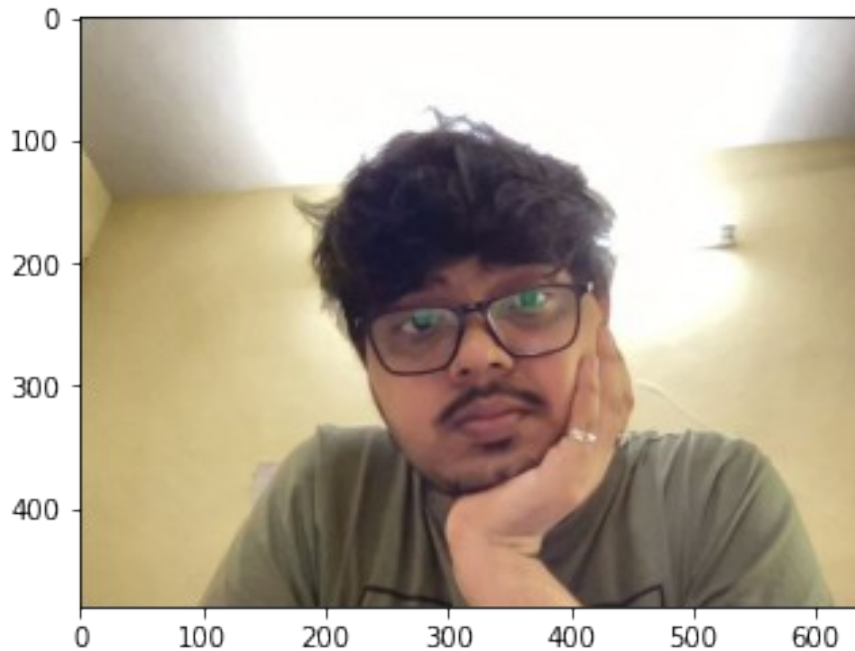
custom = model.predict(x)
emotion_analysis(custom[0])

x = np.array(x, 'float32')
x = x.reshape([48, 48]);

plt.imshow(true_image)
plt.show()

OpenCV(4.1.2) /io/opencv/modules/objdetect/src/cascadedetect.cpp:1689:
error: (-215:Assertion failed) !empty() in function 'detectMultiScale'
```





#5.To run Webcam

```
!wget --no-check-certificate \
  https://raw.githubusercontent.com/computationalcore/introduction-
to-opencv/master/assets/haarcascade_frontalface_default.xml \
  -O haarcascade_frontalface_default.xml
```

```
from IPython.display import display, Javascript, Image
from google.colab.output import eval_js
from base64 import b64decode, b64encode
import cv2
import numpy as np
import PIL
import io
import html
import time
```

```
def js_to_image(js_reply):
    """
    Params:
        js_reply: JavaScript object containing image from webcam
    Returns:
        img: OpenCV BGR image
    """
```

```

image_bytes = b64decode(js_reply.split(',')[1])

jpg_as_np = np.frombuffer(image_bytes, dtype=np.uint8)

img = cv2.imdecode(jpg_as_np, flags=1)

return img

def bbox_to_bytes(bbox_array):
    """
    Params:
        bbox_array: Numpy array (pixels) containing rectangle to
        overlay on video stream.
    Returns:
        bytes: Base64 image byte string
    """

    bbox_PIL = PIL.Image.fromarray(bbox_array, 'RGBA')
    iobuf = io.BytesIO()

    bbox_PIL.save(iobuf, format='png')

    bbox_bytes = 'data:image/png;base64,
{}'.format((str(b64encode(iobuf.getvalue())), 'utf-8'))

    return bbox_bytes

face_cascade =
cv2.CascadeClassifier(cv2.samples.findFile(cv2.data.harcascades +
'haarcascade_frontalface_default.xml'))

def video_stream():
    js = Javascript('''
    var video;
    var div = null;
    var stream;
    var captureCanvas;
    var imgElement;
    var labelElement;

    var pendingResolve = null;
    var shutdown = false;

```



```

function removeDom() {
    stream.getVideoTracks()[0].stop();
    video.remove();
    div.remove();
    video = null;
    div = null;
    stream = null;
    imgElement = null;
    captureCanvas = null;
    labelElement = null;
}

function onAnimationFrame() {
    if (!shutdown) {
        window.requestAnimationFrame(onAnimationFrame);
    }
    if (pendingResolve) {
        var result = "";
        if (!shutdown) {
            captureCanvas.getContext('2d').drawImage(video, 0, 0, 640,
480);
            result = captureCanvas.toDataURL('image/jpeg', 0.8)
        }
        var lp = pendingResolve;
        pendingResolve = null;
        lp(result);
    }
}

async function createDom() {
    if (div !== null) {
        return stream;
    }

    div = document.createElement('div');
    div.style.border = '2px solid black';
    div.style.padding = '3px';
    div.style.width = '100%';
    div.style.maxWidth = '600px';
    document.body.appendChild(div);

    const modelOut = document.createElement('div');
    modelOut.innerHTML = "<span>Status:</span>";
    labelElement = document.createElement('span');
    labelElement.innerText = 'No data';
    labelElement.style.fontWeight = 'bold';
    modelOut.appendChild(labelElement);
    div.appendChild(modelOut);

```

```

    video = document.createElement('video');
    video.style.display = 'block';
    video.width = div.clientWidth - 6;
    video.setAttribute('playsinline', '');
    video.onclick = () => { shutdown = true; };
    stream = await navigator.mediaDevices.getUserMedia(
        {video: { facingMode: "environment"}});
    div.appendChild(video);

    imgElement = document.createElement('img');
    imgElement.style.position = 'absolute';
    imgElement.style.zIndex = 1;
    imgElement.onclick = () => { shutdown = true; };
    div.appendChild(imgElement);

    const instruction = document.createElement('div');
    instruction.innerHTML =
        '<span style="color: red; font-weight: bold;">' +
        'When finished, click here or on the video to stop this
demo</span>';
    div.appendChild(instruction);
    instruction.onclick = () => { shutdown = true; };

    video.srcObject = stream;
    await video.play();

    captureCanvas = document.createElement('canvas');
    captureCanvas.width = 640; //video.videoWidth;
    captureCanvas.height = 480; //video.videoHeight;
    window.requestAnimationFrame(onAnimationFrame);

    return stream;
}
async function stream_frame(label, imgData) {
    if (shutdown) {
        removeDom();
        shutdown = false;
        return '';
    }

    var preCreate = Date.now();
    stream = await createDom();

    var preShow = Date.now();
    if (label != "") {
        labelElement.innerHTML = label;
    }

    if (imgData != "") {
        var videoRect = video.getClientRects()[0];

```

```

        imgElement.style.top = videoRect.top + "px";
        imgElement.style.left = videoRect.left + "px";
        imgElement.style.width = videoRect.width + "px";
        imgElement.style.height = videoRect.height + "px";
        imgElement.src = imgData;
    }

    var preCapture = Date.now();
    var result = await new Promise(function(resolve, reject) {
        pendingResolve = resolve;
    });
    shutdown = false;

    return {'create': preShow - preCreate,
            'show': preCapture - preShow,
            'capture': Date.now() - preCapture,
            'img': result};
    },
    ...),

display(js)

def video_frame(label, bbox):
    data = eval_js('stream_frame("{}","{}").format(label, bbox)')
    return data

colour_cycle = ((255, 0, 0), (0, 255, 0), (0, 0, 255), (230, 230, 250))

```

5.1 To load our model

```

from keras.models import model_from_json
import numpy as np
import cv2

def load_model(path):

    json_file = open(path + 'model.json', 'r')
    loaded_model_json = json_file.read()
    json_file.close()

    model = model_from_json(loaded_model_json)
    model.load_weights(path + "model.h5")
    print("Loaded model from disk")
    return model

def predict_emotion(gray, x, y, w, h):
    face = np.expand_dims(np.expand_dims(np.resize(gray[y:y+w,
x:x+h]/255.0, (48, 48)), -1), 0)

```

```

        prediction = model.predict([face])

        return(int(np.argmax(prediction)), round(max(prediction[0])*100,
2))

path = "/content/drive/MyDrive/Capstone Project 5/"
model = load_model(path)

fcc_path = "Tools/haarcascade_frontalface_alt.xml"
faceCascade = cv2.CascadeClassifier(fcc_path)
emotion_dict = {0: "Angry", 1: "Disgust", 2: "Fear", 3: "Happy", 4:
"Sad", 5: "Surprise", 6: "Neutral"}
colour_cycle = ((255, 0, 0), (0, 255, 0), (0, 0, 255), (230, 230,
250))

Loaded model from disk

```

Start VideoStream and live Recognition

```

from google.colab.patches import cv2_imshow
from IPython.display import clear_output

video_stream()

label_html = 'Capturing...'

bbox = ''
count = 0
counter = 1
while True:
    js_reply = video_frame(label_html, bbox)
    if not js_reply:
        break

    img = js_to_image(js_reply["img"])

    bbox_array = np.zeros([480,640,4], dtype=np.uint8)

    gray = cv2.cvtColor(img, cv2.COLOR_RGB2GRAY)

    faces =
face_cascade.detectMultiScale(gray,scaleFactor=1.1,minNeighbors=5,minS
ize=(30, 30))

    for (count,(x, y, w, h)) in enumerate(faces):

```

```

        colour = colour_cycle[int(count%len(colour_cycle))]
        bbox_array = cv2.rectangle(bbox_array, (x, y), (x+w, y+h),
colour, 2)
        bbox_array = cv2.line(bbox_array, (x+5, y+h+5),(x+100, y+h+5),
colour, 20)
        bbox_array = cv2.putText(bbox_array, "Face #"+str(count+1),
(x+5, y+h+11), cv2.FONT_HERSHEY_SIMPLEX, 0.5, (255, 255, 255),
lineType=cv2.LINE_AA)
        bbox_array = cv2.line(bbox_array, (x+8, y),(x+150, y), colour,
20)
        emotion_id, confidence = predict_emotion(gray, x, y, w, h)
        emotion = emotion_dict[emotion_id]
        bbox_array = cv2.putText(bbox_array, emotion + ": " +
str(confidence) + "%", (x+20, y+5), cv2.FONT_HERSHEY_SIMPLEX, 0.5,
(255, 255, 255), lineType=cv2.LINE_AA)
        bbox_array[:, :, 3] = (bbox_array.max(axis = 2) > 0 ).astype(int)
* 255

        bbox_bytes = bbox_to_bytes(bbox_array)

        bbox = bbox_bytes
        if cv2.waitKey(1) & 0xFF == ord('q'):
            break;

cv2.destroyAllWindows()

```