```
In [1]: from __future__ import print_function
    import keras
    from keras.preprocessing.image import ImageDataGenerator,img_to_array
    from keras.models import Sequential
    from keras.layers import Dense,Dropout,Activation,Flatten,BatchNormalization
    from keras.layers import Conv2D,MaxPooling2D
    import os
    from keras.optimizers import RMSprop,SGD,Adam
    from keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
    import numpy as np
```

```
In [2]: import cv2
from keras.models import load_model
```

```
In [3]: num_classes = 5
  img_rows,img_cols = 48,48
  batch_size = 32

  train_data_dir = 'C:/Users/suman/Downloads/images/train'
  validation_data_dir = 'C:/Users/suman/Downloads/images/validation'
```

```
In [1]: # from google.colab import drive
# drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call d rive.mount("/content/drive", force\_remount=True).

```
In [4]: train datagen = ImageDataGenerator(
                             rescale=1./255,
                             rotation range=30,
                             shear range=0.3,
                             zoom range=0.3,
                             width_shift_range=0.4,
                             height shift range=0.4,
                             horizontal flip=True,
                             fill mode='nearest')
        validation datagen = ImageDataGenerator(rescale=1./255)
        train_generator = train_datagen.flow_from_directory(
                             train data dir,
                             color mode='grayscale',
                             target_size=(img_rows,img_cols),
                             batch size=batch size,
                             class mode='categorical',
                             shuffle=True)
        validation generator = validation datagen.flow from directory(
                                     validation_data_dir,
                                     color mode='grayscale',
                                     target_size=(img_rows,img_cols),
                                     batch_size=batch_size,
                                     class mode='categorical',
                                     shuffle=True)
```

Found 24282 images belonging to 5 classes. Found 5937 images belonging to 5 classes.

```
In [5]: model = Sequential()

In [6]: # Block-1

model.add(Conv2D(32,(3,3),padding='same',kernel_initializer='he_normal',input_shamodel.add(Activation('elu'))
 model.add(BatchNormalization())
 model.add(Conv2D(32,(3,3),padding='same',kernel_initializer='he_normal',input_shamodel.add(Activation('elu'))
 model.add(BatchNormalization())
 model.add(MaxPooling2D(pool_size=(2,2)))
 model.add(Dropout(0.2))
```

```
In [7]: # Block-2
         model.add(Conv2D(64,(3,3),padding='same',kernel initializer='he normal'))
         model.add(Activation('elu'))
         model.add(BatchNormalization())
         model.add(Conv2D(64,(3,3),padding='same',kernel_initializer='he_normal'))
         model.add(Activation('elu'))
         model.add(BatchNormalization())
         model.add(MaxPooling2D(pool size=(2,2)))
         model.add(Dropout(0.2))
 In [8]: # Block-3
         model.add(Conv2D(128,(3,3),padding='same',kernel_initializer='he_normal'))
         model.add(Activation('elu'))
         model.add(BatchNormalization())
         model.add(Conv2D(128,(3,3),padding='same',kernel_initializer='he_normal'))
         model.add(Activation('elu'))
         model.add(BatchNormalization())
         model.add(MaxPooling2D(pool_size=(2,2)))
         model.add(Dropout(0.2))
 In [9]: # Block-4
         model.add(Conv2D(256,(3,3),padding='same',kernel initializer='he normal'))
         model.add(Activation('elu'))
         model.add(BatchNormalization())
         model.add(Conv2D(256,(3,3),padding='same',kernel_initializer='he_normal'))
         model.add(Activation('elu'))
         model.add(BatchNormalization())
         model.add(MaxPooling2D(pool_size=(2,2)))
         model.add(Dropout(0.2))
In [10]: # Block-5
         model.add(Flatten())
         model.add(Dense(64,kernel initializer='he normal'))
         model.add(Activation('elu'))
         model.add(BatchNormalization())
         model.add(Dropout(0.5))
In [11]: # Block-6
         model.add(Dense(64,kernel_initializer='he_normal'))
         model.add(Activation('elu'))
         model.add(BatchNormalization())
         model.add(Dropout(0.5))
```

```
In [12]: # Block-7

model.add(Dense(num_classes,kernel_initializer='he_normal'))
model.add(Activation('softmax'))

print(model.summary())
```

Model: "sequential"

Layer (type)	Output	Shap	oe		Param #
conv2d (Conv2D)	(None,	48,	48,	32)	320
activation (Activation)	(None,	48,	48,	32)	0
batch_normalization (BatchNo	(None,	48,	48,	32)	128
conv2d_1 (Conv2D)	(None,	48,	48,	32)	9248
activation_1 (Activation)	(None,	48,	48,	32)	0
batch_normalization_1 (Batch	(None,	48,	48,	32)	128
max_pooling2d (MaxPooling2D)	(None,	24,	24,	32)	0
dropout (Dropout)	(None,	24,	24,	32)	0
conv2d_2 (Conv2D)	(None,	24,	24,	64)	18496
activation_2 (Activation)	(None,	24,	24,	64)	0
batch_normalization_2 (Batch	(None,	24,	24,	64)	256
conv2d_3 (Conv2D)	(None,	24,	24,	64)	36928
activation_3 (Activation)	(None,	24,	24,	64)	0
batch_normalization_3 (Batch	(None,	24,	24,	64)	256
max_pooling2d_1 (MaxPooling2	(None,	12,	12,	64)	0
dropout_1 (Dropout)	(None,	12,	12,	64)	0
conv2d_4 (Conv2D)	(None,	12,	12,	128)	73856
activation_4 (Activation)	(None,	12,	12,	128)	0
batch_normalization_4 (Batch	(None,	12,	12,	128)	512
conv2d_5 (Conv2D)	(None,	12,	12,	128)	147584
activation_5 (Activation)	(None,	12,	12,	128)	0
batch_normalization_5 (Batch	(None,	12,	12,	128)	512

<pre>max_pooling2d_2 (MaxPooling2</pre>	(None,	6, 6, 128)	0
dropout_2 (Dropout)	(None,	6, 6, 128)	0
conv2d_6 (Conv2D)	(None,	6, 6, 256)	295168
activation_6 (Activation)	(None,	6, 6, 256)	0
batch_normalization_6 (Batch	(None,	6, 6, 256)	1024
conv2d_7 (Conv2D)	(None,	6, 6, 256)	590080
activation_7 (Activation)	(None,	6, 6, 256)	0
batch_normalization_7 (Batch	(None,	6, 6, 256)	1024
max_pooling2d_3 (MaxPooling2	(None,	3, 3, 256)	0
dropout_3 (Dropout)	(None,	3, 3, 256)	0
flatten (Flatten)	(None,	2304)	0
dense (Dense)	(None,	64)	147520
activation_8 (Activation)	(None,	64)	0
batch_normalization_8 (Batch	(None,	64)	256
dropout_4 (Dropout)	(None,	64)	0
dense_1 (Dense)	(None,	64)	4160
activation_9 (Activation)	(None,	64)	0
batch_normalization_9 (Batch	(None,	64)	256
dropout_5 (Dropout)	(None,	64)	0
dense_2 (Dense)	(None,	5)	325
activation_10 (Activation)	(None,	·	0

Total params: 1,328,037 Trainable params: 1,325,861 Non-trainable params: 2,176

None

```
In [13]:
         checkpoint = ModelCheckpoint('Emotion_little_vgg.h5',
                                       monitor='val loss',
                                       mode='min',
                                       save_best_only=True,
                                       verbose=1)
         earlystop = EarlyStopping(monitor='val_loss',
                                    min_delta=0,
                                    patience=3,
                                    verbose=1,
                                    restore_best_weights=True
         reduce_lr = ReduceLROnPlateau(monitor='val_loss',
                                        factor=0.2,
                                        patience=3,
                                        verbose=1,
                                        min_delta=0.0001)
         callbacks = [earlystop,checkpoint,reduce_lr]
```

```
In [14]: | model.compile(loss='categorical crossentropy',
                      optimizer = Adam(lr=0.001),
                      metrics=['accuracy'])
         nb train samples = 24176
         nb_validation_samples = 3006
         epochs=25
         history=model.fit generator(
                        train_generator,
                        steps per epoch=nb train samples//batch size,
                        epochs=epochs,
                        callbacks=callbacks,
                        validation data=validation_generator,
                        validation steps=nb validation samples//batch size)
         E:\Anaconda\lib\site-packages\tensorflow\python\keras\engine\training.py:1844:
         UserWarning: `Model.fit_generator` is deprecated and will be removed in a futur
         e version. Please use `Model.fit`, which supports generators.
          warnings.warn('`Model.fit generator` is deprecated and '
         Epoch 1/25
         755/755 [============= ] - 527s 693ms/step - loss: 2.1226 - a
         ccuracy: 0.2223 - val_loss: 1.5537 - val_accuracy: 0.3004
         Epoch 00001: val loss improved from inf to 1.55368, saving model to Emotion 1
         ittle_vgg.h5
         Epoch 2/25
         755/755 [============= ] - 489s 648ms/step - loss: 1.5976 - a
         ccuracy: 0.2638 - val_loss: 1.5351 - val_accuracy: 0.2957
         Epoch 00002: val loss improved from 1.55368 to 1.53515, saving model to Emoti
         on_little_vgg.h5
         Epoch 3/25
         755/755 [============= ] - 472s 626ms/step - loss: 1.5500 - a
         ccuracy: 0.2992 - val loss: 1.4711 - val accuracy: 0.3663
         Epoch 00003: val loss improved from 1.53515 to 1.47106, saving model to Emoti
         on_little_vgg.h5
         Epoch 4/25
         755/755 [============= ] - 448s 593ms/step - loss: 1.5238 - a
         ccuracy: 0.3159 - val_loss: 1.4299 - val_accuracy: 0.3760
         Epoch 00004: val loss improved from 1.47106 to 1.42988, saving model to Emoti
         on_little_vgg.h5
         Epoch 5/25
         755/755 [============ ] - 425s 563ms/step - loss: 1.4573 - a
         ccuracy: 0.3628 - val loss: 1.4743 - val accuracy: 0.4113
         Epoch 00005: val loss did not improve from 1.42988
         Epoch 6/25
         755/755 [============= ] - 407s 538ms/step - loss: 1.3687 - a
         ccuracy: 0.4204 - val loss: 1.1898 - val accuracy: 0.5134
         Epoch 00006: val loss improved from 1.42988 to 1.18985, saving model to Emoti
         on little vgg.h5
         Epoch 7/25
```

```
755/755 [============= ] - 366s 485ms/step - loss: 1.2857 - a
ccuracy: 0.4619 - val_loss: 1.0016 - val_accuracy: 0.5857
Epoch 00007: val loss improved from 1.18985 to 1.00161, saving model to Emoti
on little vgg.h5
Epoch 8/25
755/755 [============= ] - 399s 529ms/step - loss: 1.2136 - a
ccuracy: 0.4997 - val_loss: 0.9423 - val_accuracy: 0.6179
Epoch 00008: val loss improved from 1.00161 to 0.94225, saving model to Emoti
on little vgg.h5
Epoch 9/25
755/755 [============= ] - 504s 668ms/step - loss: 1.1890 - a
ccuracy: 0.5240 - val_loss: 0.9160 - val_accuracy: 0.6300
Epoch 00009: val loss improved from 0.94225 to 0.91598, saving model to Emoti
on little vgg.h5
Epoch 10/25
755/755 [============= ] - 442s 586ms/step - loss: 1.1616 - a
ccuracy: 0.5296 - val loss: 0.8710 - val accuracy: 0.6546
Epoch 00010: val loss improved from 0.91598 to 0.87101, saving model to Emoti
on little vgg.h5
Epoch 11/25
755/755 [============== ] - 498s 660ms/step - loss: 1.1225 - a
ccuracy: 0.5466 - val_loss: 0.8900 - val_accuracy: 0.6552
Epoch 00011: val_loss did not improve from 0.87101
Epoch 12/25
755/755 [============= ] - 459s 607ms/step - loss: 1.1108 - a
ccuracy: 0.5520 - val_loss: 0.8770 - val_accuracy: 0.6475
Epoch 00012: val loss did not improve from 0.87101
Epoch 13/25
755/755 [============= ] - 465s 616ms/step - loss: 1.1060 - a
ccuracy: 0.5595 - val_loss: 0.8604 - val_accuracy: 0.6663
Epoch 00013: val loss improved from 0.87101 to 0.86040, saving model to Emoti
on little vgg.h5
Epoch 14/25
755/755 [============= ] - 459s 608ms/step - loss: 1.0713 - a
ccuracy: 0.5769 - val_loss: 0.9210 - val_accuracy: 0.6502
Epoch 00014: val loss did not improve from 0.86040
Epoch 15/25
755/755 [============= ] - 463s 613ms/step - loss: 1.0588 - a
ccuracy: 0.5866 - val_loss: 0.8454 - val_accuracy: 0.6801
Epoch 00015: val_loss improved from 0.86040 to 0.84535, saving model to Emoti
on_little_vgg.h5
Epoch 16/25
755/755 [============= ] - 446s 591ms/step - loss: 1.0450 - a
ccuracy: 0.5949 - val_loss: 0.7866 - val_accuracy: 0.6956
Epoch 00016: val_loss improved from 0.84535 to 0.78664, saving model to Emoti
on_little_vgg.h5
Epoch 17/25
```

```
In [3]: face_classifier = cv2.CascadeClassifier(r'C:\Users\suman\Downloads\haarcascade_fr
classifier =load_model(r'C:\Users\suman\Downloads\Emotion_little_vgg.h5')

class_labels = ['Angry','Happy','Neutral','Sad','Surprise']

cap = cv2.VideoCapture(0)
```

```
In [4]: from PIL import *
```

In [5]: |while True:

# Grab a single frame of video

ret, frame = cap.read()

```
labels = []
            gray = cv2.cvtColor(frame,cv2.COLOR BGR2GRAY)
            faces = face_classifier.detectMultiScale(gray,1.3,5)
            for (x,y,w,h) in faces:
                cv2.rectangle(frame,(x,y),(x+w,y+h),(255,0,0),2)
                roi_gray = gray[y:y+h,x:x+w]
                 roi_gray = cv2.resize(roi_gray,(48,48),interpolation=cv2.INTER_AREA)
            # rect,face,image = face_detector(frame)
                if np.sum([roi gray])!=0:
                     roi = roi_gray.astype('float')/255.0
                     roi = img to array(roi)
                     roi = np.expand_dims(roi,axis=0)
                # make a prediction on the ROI, then lookup the class
                     preds = classifier.predict(roi)[0]
                     label=class labels[preds.argmax()]
                     label position = (x,y)
                     cv2.putText(frame, label, label position, cv2.FONT HERSHEY SIMPLEX, 2, (0,
                else:
                     cv2.putText(frame, 'No Face Found', (20,60), cv2.FONT HERSHEY SIMPLEX, 2,
            cv2.imshow('Emotion Detector',frame)
            if cv2.waitKey(1) & 0xFF == ord('q'):
                break
        cap.release()
        cv2.destroyAllWindows()
In [ ]:
In [ ]:
In [ ]:
```