```
In [1]: import numpy as np
        import pandas as pd
        import seaborn as sns
        import matplotlib.pyplot as plt
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.linear model import LogisticRegression
        from sklearn.ensemble import RandomForestClassifier
        import glob
        from sklearn.model selection import train test split
        from keras.layers import Dropout, Dense
        from keras.layers.normalization import BatchNormalization
        from keras.models import Sequential, load model
        from keras.applications import VGG16
        from sklearn.metrics import accuracy score, confusion matrix
        C:\Users\LENOVO\Anaconda3\lib\site-packages\h5py\__init__.py:36: Future
        Warning: Conversion of the second argument of issubdtype from `float` t
        o `np.floating` is deprecated. In future, it will be treated as `np.flo
        at64 == np.dtype(float).type`.
          from . conv import register converters as register converters
        Using TensorFlow backend.
In [2]: def load fer2013():
            data = pd.read csv('fer2013.csv')
            pixels = data['pixels'].tolist()
            width, height = 48,48
            faces = []
            input shape = (48, 48, 1)
            for sequence in pixels:
                face = [int(pixel) for pixel in sequence.split(' ')]
                face = np.asarray(face).reshape(width,height)
                face= cv2.resize(face.astype('uint8'), input shape[:2])
                faces.append(face.astype('float32'))
            faces = np.asarray(faces)
            faces = np.expand dims(faces,-1)
```

```
emotions = pd.get dummies(data['emotion']).as matrix()
            return faces, emotions
In [3]: def split data(x, y, validation split=.2):
            num samples = len(x)
            num train samples = int((1 - validation split)*num samples)
            train x = x[:num train samples]
            train y = y[:num train samples]
            val x = x[num train samples:]
            val y = y[num train samples:]
            train data = (train x, train y)
            val data = (val x, val y)
            return train data, val data
In [4]: from keras.layers import Activation, Convolution2D, Dropout, Conv2D
        from keras.layers import AveragePooling2D, BatchNormalization
        from keras.layers import GlobalAveragePooling2D
        from keras.models import Sequential
        from keras.layers import Flatten
        from keras.models import Model
        from keras.layers import Input
        from keras.layers import MaxPooling2D
        from keras.layers import SeparableConv2D
        from keras import layers
        from keras.regularizers import 12
In [6]: def preprocess input(x, v2=True):
            x = x.astype('float32')
            x = x / 255.0
            if v2:
                x = x - 0.5
                x = x * 2.0
            return x
In [7]: def mini XCEPTION(input shape, num classes, l2 regularization=0.01):
            regularization = l2(l2 regularization)
```

```
# base
   img input = Input(input shape)
   x = Conv2D(8, (3, 3), strides=(1, 1), kernel regularizer=regulariza
tion,
               use bias=False)(img input)
   x = BatchNormalization()(x)
   x = Activation('relu')(x)
   x = Conv2D(8, (3, 3), strides=(1, 1), kernel regularizer=regulariza
tion,
               use bias=False)(x)
   x = BatchNormalization()(x)
   x = Activation('relu')(x)
    # module 1
   residual = Conv2D(16, (1, 1), strides=(2, 2),
                      padding='same', use bias=False)(x)
   residual = BatchNormalization()(residual)
   x = SeparableConv2D(16, (3, 3), padding='same',
                        kernel regularizer=regularization,
                        use bias=False)(x)
   x = BatchNormalization()(x)
   x = Activation('relu')(x)
   x = SeparableConv2D(16, (3, 3), padding='same',
                        kernel regularizer=regularization,
                        use bias=False)(x)
   x = BatchNormalization()(x)
   x = MaxPooling2D((3, 3), strides=(2, 2), padding='same')(x)
   x = layers.add([x, residual])
    # module 2
   residual = Conv2D(32, (1, 1), strides=(2, 2),
                      padding='same', use bias=False)(x)
    residual = BatchNormalization()(residual)
   x = SeparableConv2D(32, (3, 3), padding='same',
                        kernel regularizer=regularization,
                        use bias=False)(x)
```

```
x = BatchNormalization()(x)
x = Activation('relu')(x)
x = SeparableConv2D(32, (3, 3), padding='same',
                    kernel regularizer=regularization,
                    use bias=False)(x)
x = BatchNormalization()(x)
x = MaxPooling2D((3, 3), strides=(2, 2), padding='same')(x)
x = layers.add([x, residual])
# module 3
residual = Conv2D(64, (1, 1), strides=(2, 2),
                  padding='same', use_bias=False)(x)
residual = BatchNormalization()(residual)
x = SeparableConv2D(64, (3, 3), padding='same',
                    kernel regularizer=regularization,
                    use bias=False)(x)
x = BatchNormalization()(x)
x = Activation('relu')(x)
x = SeparableConv2D(64, (3, 3), padding='same',
                    kernel regularizer=regularization,
                    use bias=False)(x)
x = BatchNormalization()(x)
x = MaxPooling2D((3, 3), strides=(2, 2), padding='same')(x)
x = layers.add([x, residual])
# module 4
residual = Conv2D(128, (1, 1), strides=(2, 2),
                  padding='same', use bias=False)(x)
residual = BatchNormalization()(residual)
x = SeparableConv2D(128, (3, 3), padding='same',
                    kernel regularizer=regularization,
                    use bias=False)(x)
x = BatchNormalization()(x)
x = Activation('relu')(x)
x = SeparableConv2D(128, (3, 3), padding='same',
```

```
kernel regularizer=regularization,
                                use bias=False)(x)
            x = BatchNormalization()(x)
            x = MaxPooling2D((3, 3), strides=(2, 2), padding='same')(x)
            x = layers.add([x, residual])
            x = Conv2D(num classes, (3, 3),
                       # kernel regularizer=regularization,
                       padding='same')(x)
            x = GlobalAveragePooling2D()(x)
            output = Activation('softmax', name='predictions')(x)
            model = Model(img input, output)
            return model
In [2]: import cv2
        from keras import applications
        from keras import optimizers
        from keras.preprocessing.image import ImageDataGenerator
        from keras.models import Sequential
        from keras.layers import Dense, Dropout, Flatten , BatchNormalization
        from keras.layers import Conv2D, MaxPooling2D
        from keras import backend as K
        from keras.layers import Input
        batch size = 128
        num classes = 10
        epochs = 12
In [9]: datagen = ImageDataGenerator(featurewise center=False,
                                featurewise std normalization=True,
                                 rotation range=10,
                                width shift range=0.1,
                                height shift range=0.1,
                                zoom range=.1,
                                horizontal flip=True)
        /usr/local/lib/python3.5/dist-packages/keras preprocessing/image/image
        data generator.py:346: UserWarning: This ImageDataGenerator specifies
```

```
featurewise std normalization`, which overrides setting of `featurewise
         center`.
           warnings.warn('This ImageDataGenerator specifies '
In [4]: from keras.regularizers import 12
         from keras.layers import Activation, Convolution2D, Dropout, Conv2D
         from keras.layers import AveragePooling2D, BatchNormalization
         from keras.layers import GlobalAveragePooling2D
         from keras.models import Sequential
         from keras.layers import Flatten
         from keras.models import Model
         from keras.layers import Input
         from keras.layers import MaxPooling2D
         from keras.layers import SeparableConv2D
         from keras import layers
In [11]: input shape =(48,48,1)
         model = mini XCEPTION(input shape,7)
         model.compile(optimizer='adam', loss='categorical crossentropy',
                       metrics=['accuracy'])
         model.summary()
         WARNING:tensorflow:From /usr/local/lib/python3.5/dist-packages/tensorfl
         ow/python/framework/op def library.py:263: colocate with (from tensorfl
         ow.python.framework.ops) is deprecated and will be removed in a future
         version.
         Instructions for updating:
         Colocations handled automatically by placer.
         Layer (type)
                                         Output Shape
                                                             Param #
                                                                          Connec
         ted to
         input 1 (InputLayer)
                                        (None, 48, 48, 1)
         conv2d 1 (Conv2D)
                                         (None, 46, 46, 8)
                                                             72
                                                                         input
```

| (None, | 46, | 46, | 8) | 32 | conv2d |
|--------|--|--|--|--|--|
| (None, | 46, | 46, | 8) | 0 | batch_ |
| (None, | 44, | 44, | 8) | 576 | activa |
| None, | 44, | 44, | 8) | 32 | conv2d |
| (None, | 44, | 44, | 8) | 0 | batch_ |
| (None, | 44, | 44, | 16) | 200 | activa |
| (None, | 44, | 44, | 16) | 64 | separa |
| (None, | 44, | 44, | 16) | 0 | batch_ |
| (None, | 44, | 44, | 16) | 400 | activa |
| | (None, (None, (None, (None, (None, (None, | (None, 46, (None, 44, (None, 44, (None, 44, (None, 44, | (None, 46, 46, (None, 44, 44, (None, 44, 44, (None, 44, 44, (None, 44, 44, (None, 44, 44, | (None, 46, 46, 8) (None, 46, 46, 8) (None, 44, 44, 8) (None, 44, 44, 8) (None, 44, 44, 16) (None, 44, 44, 16) (None, 44, 44, 16) | (None, 46, 46, 8) 0 (None, 44, 44, 8) 576 (None, 44, 44, 8) 32 (None, 44, 44, 8) 0 (None, 44, 44, 16) 200 (None, 44, 44, 16) 64 (None, 44, 44, 16) 0 |

| <pre>batch_normalization_5 (BatchNor ble_conv2d_2[0][0]</pre> | (None, | 44, | 44, | 16) | 64 | separa |
|---|--------|-----|-----|-----|------|--------|
| conv2d_3 (Conv2D) tion_2[0][0] | (None, | 22, | 22, | 16) | 128 | activa |
| <pre>max_pooling2d_1 (MaxPooling2D) normalization_5[0][0]</pre> | (None, | 22, | 22, | 16) | 0 | batch_ |
| <pre>batch_normalization_3 (BatchNor _3[0][0]</pre> | (None, | 22, | 22, | 16) | 64 | conv2d |
| add_1 (Add) oling2d_1[0][0] normalization_3[0][0] | (None, | 22, | 22, | 16) | 0 | max_po |
| <pre>separable_conv2d_3 (SeparableCo [0][0]</pre> | (None, | 22, | 22, | 32) | 656 | add_1 |
| batch_normalization_7 (BatchNorble_conv2d_3[0][0] | (None, | 22, | 22, | 32) | 128 | separa |
| activation_4 (Activation) normalization_7[0][0] | (None, | 22, | 22, | 32) | 0 | batch_ |
| <pre>separable_conv2d_4 (SeparableCo tion_4[0][0]</pre> | (None, | 22, | 22, | 32) | 1312 | activa |
| batch_normalization_8 (BatchNor | (None, | 22, | 22, | 32) | 128 | separa |

| ble_conv2d_4[0][0] | | | | | | |
|---|--------|-----|-----|-----|------|--------|
| conv2d_4 (Conv2D) [0][0] | (None, | 11, | 11, | 32) | 512 | add_1 |
| <pre>max_pooling2d_2 (MaxPooling2D) normalization_8[0][0]</pre> | (None, | 11, | 11, | 32) | 0 | batch_ |
| <pre>batch_normalization_6 (BatchNor _4[0][0]</pre> | (None, | 11, | 11, | 32) | 128 | conv2d |
| add_2 (Add) oling2d_2[0][0] normalization_6[0][0] | (None, | 11, | 11, | 32) | 0 | max_po |
| <pre>separable_conv2d_5 (SeparableCo [0][0]</pre> | (None, | 11, | 11, | 64) | 2336 | add_2 |
| batch_normalization_10 (BatchNo ble_conv2d_5[0][0] | (None, | 11, | 11, | 64) | 256 | separa |
| activation_5 (Activation) normalization_10[0][0] | (None, | 11, | 11, | 64) | 0 | batch_ |
| <pre>separable_conv2d_6 (SeparableCo tion_5[0][0]</pre> | (None, | 11, | 11, | 64) | 4672 | activa |
| batch_normalization_11 (BatchNo ble_conv2d_6[0][0] | (None, | 11, | 11, | 64) | 256 | separa |

| conv2d_5 (Conv2D) [0][0] | (None, | 6, | 6, | 64) | 2048 | add_2 |
|--|--------|----|----|------|-------|--------|
| <pre>max_pooling2d_3 (MaxPooling2D) normalization_11[0][0]</pre> | (None, | 6, | 6, | 64) | 0 | batch_ |
| <pre>batch_normalization_9 (BatchNor _5[0][0]</pre> | (None, | 6, | 6, | 64) | 256 | conv2d |
| add_3 (Add) oling2d_3[0][0] | (None, | 6, | 6, | 64) | 0 | max_po |
| normalization_9[0][0] | | | | | | batch_ |
| <pre>separable_conv2d_7 (SeparableCo [0][0]</pre> | (None, | 6, | 6, | 128) | 8768 | add_3 |
| batch_normalization_13 (BatchNo ble_conv2d_7[0][0] | (None, | 6, | 6, | 128) | 512 | separa |
| activation_6 (Activation) normalization_13[0][0] | (None, | 6, | 6, | 128) | 0 | batch_ |
| <pre>separable_conv2d_8 (SeparableCo tion_6[0][0]</pre> | (None, | 6, | 6, | 128) | 17536 | activa |
| batch_normalization_14 (BatchNo ble_conv2d_8[0][0] | (None, | 6, | 6, | 128) | 512 | separa |
| | | | | | | |

```
conv2d 6 (Conv2D)
                                         (None, 3, 3, 128)
                                                             8192
                                                                          add 3
         [0][0]
         max pooling2d 4 (MaxPooling2D) (None, 3, 3, 128)
                                                                          batch
         normalization_14[0][0]
         batch normalization 12 (BatchNo (None, 3, 3, 128)
                                                              512
                                                                          conv2d
         _6[0][0]
         add 4 (Add)
                                         (None, 3, 3, 128)
                                                                          max po
         oling2d 4[0][0]
                                                                          batch
         normalization 12[0][0]
         conv2d 7 (Conv2D)
                                                              8071
                                         (None, 3, 3, 7)
                                                                          add 4
         [0][0]
         global average_pooling2d_1 (Glo (None, 7)
                                                                          conv2d
         7[0][0]
         predictions (Activation)
                                         (None, 7)
                                                                          global
         average pooling2d 1[0][0]
         Total params: 58,423
         Trainable params: 56,951
         Non-trainable params: 1,472
In [41]: faces , emotions = load_fer2013()
```

Warning: Method .as matrix will be removed in a future version. Use .va lues instead. In [42]: faces = preprocess input(faces) In [43]: num shapes , num classes = emotions.shape In [44]: num shapes Out[44]: 35887 In [45]: num classes Out[45]: 7 In [46]: validation split = 0.2 In [47]: train data, val data = split data(faces, emotions, validation split) In [48]: train faces , train emotion = train data In [62]: batch size = 64 num epochs = 100input shape = (48, 48, 1)validation split = .2 $verbose = \overline{1}$ num classes = 7patience = 50In []: base path = emodetect/ In []: log_file_path = base_path + dataset_name + '_emotion_training.log' csv logger = CSVLogger(log file path, append=False)

/usr/local/lib/python3.5/dist-packages/ipykernel launcher.py:14: Future

```
early_stop = EarlyStopping('val_loss', patience=patience)
       reduce lr = ReduceLROnPlateau('val loss', factor=0.1,
                                  patience=int(patience/4), verbose=1)
       trained_models_path = base_path + dataset_name + '_mini_XCEPTION'
       model names = trained models path + '.{epoch:02d}-{val acc:.2f}.hdf5'
       model checkpoint = ModelCheckpoint(model names, 'val loss', verbose=1,
                                                save best only=True
       callbacks = [model checkpoint, csv logger, early stop, reduce lr]
In [63]: model.fit generator(datagen.flow(train faces, train emotion,
                                          batch size),
                          steps per epoch=100,
                          epochs=num epochs, verbose=1,
                          validation data=val data)
       Epoch 1/100
       100/100 [============ ] - 124s ls/step - loss: 1.1164
       - acc: 0.5858 - val loss: 1.1896 - val_acc: 0.5631
       Epoch 2/100
       100/100 [=========== ] - 130s ls/step - loss: 1.1310
       - acc: 0.5744 - val loss: 1.1983 - val acc: 0.5612
       Epoch 3/100
       100/100 [============== ] - 132s 1s/step - loss: 1.1097
       - acc: 0.5786 - val loss: 1.4524 - val acc: 0.4979
       Epoch 4/100
       - acc: 0.5792 - val loss: 1.2078 - val acc: 0.5598
       Epoch 5/100
       - acc: 0.5770 - val loss: 1.1673 - val acc: 0.5609
       Epoch 6/100
       - acc: 0.5894 - val loss: 1.1634 - val acc: 0.5667
       Epoch 7/100
       100/100 [=========== ] - 134s ls/step - loss: 1.1093
       - acc: 0.5864 - val loss: 1.2080 - val acc: 0.5532
       Epoch 8/100
```

```
- acc: 0.5909 - val loss: 1.2182 - val acc: 0.5481
Epoch 9/100
- acc: 0.5785 - val loss: 1.1362 - val acc: 0.5790
Epoch 10/100
- acc: 0.5933 - val loss: 1.1582 - val acc: 0.5766
Epoch 11/100
- acc: 0.5958 - val loss: 1.1772 - val acc: 0.5734
Epoch 12/100
- acc: 0.5861 - val loss: 1.2263 - val acc: 0.5639
Epoch 13/100
- acc: 0.5841 - val loss: 1.2141 - val acc: 0.5716
Epoch 14/100
- acc: 0.6001 - val loss: 1.1916 - val acc: 0.5549
Epoch 15/100
100/100 [=============== ] - 130s 1s/step - loss: 1.0752
- acc: 0.6041 - val loss: 1.2195 - val acc: 0.5502
Epoch 16/100
- acc: 0.5959 - val loss: 1.1558 - val acc: 0.5677
Epoch 17/100
- acc: 0.5872 - val loss: 1.1293 - val acc: 0.5840
Epoch 18/100
- acc: 0.5878 - val loss: 1.1434 - val acc: 0.5692
Epoch 19/100
- acc: 0.5939 - val loss: 1.1655 - val acc: 0.5678
Epoch 20/100
- acc: 0.5936 - val loss: 1.1189 - val acc: 0.5837
Epoch 21/100
```

```
- acc: 0.5961 - val loss: 1.1045 - val acc: 0.5900
Epoch 22/100
- acc: 0.6030 - val loss: 1.2405 - val acc: 0.5450
Epoch 23/100
- acc: 0.6019 - val loss: 1.1557 - val acc: 0.5768
Epoch 24/100
100/100 [============ ] - 135s 1s/step - loss: 1.0696
- acc: 0.5992 - val loss: 1.1591 - val acc: 0.5692
Epoch 25/100
- acc: 0.5994 - val loss: 1.1061 - val acc: 0.5931
Epoch 26/100
- acc: 0.5925 - val loss: 1.1694 - val acc: 0.5699
Epoch 27/100
- acc: 0.5918 - val loss: 1.1387 - val acc: 0.5821
Epoch 28/100
- acc: 0.6067 - val loss: 1.2104 - val acc: 0.5699
Epoch 29/100
- acc: 0.6031 - val loss: 1.1490 - val acc: 0.5711
Epoch 30/100
- acc: 0.6025 - val loss: 1.1649 - val acc: 0.5740
Epoch 31/100
- acc: 0.6020 - val loss: 1.2368 - val acc: 0.5626
Epoch 32/100
- acc: 0.5894 - val loss: 1.1174 - val acc: 0.5894
Epoch 33/100
100/100 [============== ] - 132s 1s/step - loss: 1.0566
- acc: 0.5947 - val loss: 1.1247 - val acc: 0.5794
Epoch 34/100
```

```
- acc: 0.6053 - val loss: 1.1190 - val acc: 0.5770
Epoch 35/100
- acc: 0.6006 - val loss: 1.0966 - val acc: 0.6004
Epoch 36/100
- acc: 0.6147 - val loss: 1.1326 - val acc: 0.5848
Epoch 37/100
- acc: 0.6058 - val loss: 1.1263 - val acc: 0.5901
Epoch 38/100
- acc: 0.6034 - val loss: 1.1153 - val acc: 0.5843
Epoch 39/100
- acc: 0.5980 - val loss: 1.0999 - val acc: 0.5970
Epoch 40/100
- acc: 0.6078 - val loss: 1.1879 - val acc: 0.5658
Epoch 41/100
- acc: 0.6189 - val loss: 1.1920 - val acc: 0.5724
Epoch 42/100
- acc: 0.6134 - val loss: 1.1571 - val acc: 0.5722
Epoch 43/100
- acc: 0.5972 - val loss: 1.0916 - val acc: 0.5949
Epoch 44/100
- acc: 0.6070 - val loss: 1.1637 - val acc: 0.5772
Epoch 45/100
- acc: 0.6065 - val loss: 1.1859 - val acc: 0.5726
Epoch 46/100
100/100 [============== ] - 139s 1s/step - loss: 1.0322
- acc: 0.6080 - val loss: 1.1544 - val acc: 0.5762
Epoch 47/100
```

```
- acc: 0.5991 - val loss: 1.1277 - val acc: 0.5836
Epoch 48/100
- acc: 0.6164 - val loss: 1.0863 - val acc: 0.6021
Epoch 49/100
- acc: 0.6097 - val loss: 1.1329 - val acc: 0.5780
Epoch 50/100
- acc: 0.5976 - val loss: 1.1328 - val acc: 0.5784
Epoch 51/100
- acc: 0.6155 - val loss: 1.2352 - val acc: 0.5368
Epoch 52/100
- acc: 0.6036 - val loss: 1.0989 - val acc: 0.5942
Epoch 53/100
- acc: 0.6131 - val loss: 1.1137 - val acc: 0.5992
Epoch 54/100
- acc: 0.6164 - val loss: 1.1646 - val acc: 0.5719
Epoch 55/100
- acc: 0.6072 - val loss: 1.1049 - val acc: 0.5961
Epoch 56/100
- acc: 0.6155 - val loss: 1.1591 - val acc: 0.5772
Epoch 57/100
- acc: 0.6259 - val loss: 1.0844 - val acc: 0.6071
Epoch 58/100
- acc: 0.6148 - val loss: 1.0802 - val acc: 0.6042
Epoch 59/100
100/100 [============== ] - 130s 1s/step - loss: 1.0016
- acc: 0.6280 - val loss: 1.1060 - val acc: 0.5970
Epoch 60/100
```

```
- acc: 0.6069 - val loss: 1.0735 - val acc: 0.6082
Epoch 61/100
- acc: 0.6205 - val loss: 1.0858 - val acc: 0.6045
Epoch 62/100
- acc: 0.6186 - val loss: 1.0660 - val acc: 0.6096
Epoch 63/100
100/100 [============ ] - 139s 1s/step - loss: 1.0212
- acc: 0.6182 - val loss: 1.1269 - val acc: 0.5874
Epoch 64/100
- acc: 0.6159 - val loss: 1.1380 - val acc: 0.5864
Epoch 65/100
- acc: 0.6208 - val loss: 1.0897 - val acc: 0.6025
Epoch 66/100
- acc: 0.6120 - val loss: 1.1106 - val acc: 0.5977
Epoch 67/100
100/100 [=============== ] - 135s 1s/step - loss: 1.0180
- acc: 0.6248 - val loss: 1.0779 - val acc: 0.6080
Epoch 68/100
- acc: 0.6235 - val loss: 1.1081 - val acc: 0.5971
Epoch 69/100
- acc: 0.6270 - val loss: 1.0848 - val acc: 0.5942
Epoch 70/100
- acc: 0.6108 - val loss: 1.0931 - val acc: 0.5986
Epoch 71/100
- acc: 0.6194 - val loss: 1.1369 - val acc: 0.5897
Epoch 72/100
100/100 [============== ] - 135s 1s/step - loss: 1.0259
- acc: 0.6196 - val loss: 1.1715 - val acc: 0.5740
Epoch 73/100
```

```
- acc: 0.6300 - val loss: 1.2282 - val acc: 0.5588
Epoch 74/100
- acc: 0.6241 - val loss: 1.1039 - val acc: 0.5876
Epoch 75/100
- acc: 0.6267 - val loss: 1.1469 - val acc: 0.5901
Epoch 76/100
100/100 [============ ] - 130s 1s/step - loss: 1.0175
- acc: 0.6213 - val loss: 1.1153 - val acc: 0.5911
Epoch 77/100
- acc: 0.6279 - val loss: 1.1104 - val acc: 0.6060
Epoch 78/100
100/100 [============ ] - 136s 1s/step - loss: 0.9960
- acc: 0.6314 - val loss: 1.1223 - val acc: 0.5997
Epoch 79/100
- acc: 0.6291 - val loss: 1.0870 - val acc: 0.6018
Epoch 80/100
- acc: 0.6211 - val loss: 1.0927 - val acc: 0.6032
Epoch 81/100
- acc: 0.6300 - val loss: 1.0935 - val acc: 0.6073
Epoch 82/100
- acc: 0.6205 - val loss: 1.1970 - val acc: 0.5690
Epoch 83/100
- acc: 0.6259 - val loss: 1.1039 - val acc: 0.5985
Epoch 84/100
- acc: 0.6286 - val loss: 1.1052 - val acc: 0.5978
Epoch 85/100
100/100 [=============== ] - 130s 1s/step - loss: 1.0241
- acc: 0.6162 - val loss: 1.1084 - val acc: 0.5919
Epoch 86/100
```

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- acc: 0.6239 - val loss: 1.1141 - val acc: 0.5918
Epoch 87/100
- acc: 0.6280 - val loss: 1.2237 - val acc: 0.5627
Epoch 88/100
- acc: 0.6209 - val loss: 1.0957 - val acc: 0.5947
Epoch 89/100
100/100 [============ ] - 134s 1s/step - loss: 0.9928
- acc: 0.6247 - val loss: 1.2730 - val acc: 0.5364
Epoch 90/100
- acc: 0.6219 - val loss: 1.1031 - val acc: 0.6027
Epoch 91/100
100/100 [=========== ] - 132s 1s/step - loss: 0.9739
- acc: 0.6395 - val loss: 1.0707 - val acc: 0.6116
Epoch 92/100
- acc: 0.6208 - val loss: 1.0796 - val acc: 0.5979
Epoch 93/100
- acc: 0.6323 - val loss: 1.0758 - val acc: 0.6042
Epoch 94/100
- acc: 0.6230 - val loss: 1.1292 - val acc: 0.5938
Epoch 95/100
- acc: 0.6347 - val loss: 1.0638 - val acc: 0.6120
Epoch 96/100
- acc: 0.6255 - val loss: 1.0827 - val acc: 0.6070
Epoch 97/100
- acc: 0.6258 - val loss: 1.0969 - val acc: 0.5940
Epoch 98/100
- acc: 0.6325 - val loss: 1.1130 - val acc: 0.5993
Epoch 99/100
```

```
- acc: 0.6257 - val loss: 1.1689 - val acc: 0.5651
        Epoch 100/100
        - acc: 0.6280 - val loss: 1.0550 - val acc: 0.6089
Out[63]: <keras.callbacks.History at 0x7ff52e3f7470>
In [64]: model.save('mini xception model.hdf5')
In [16]: emotion classifier.history
        AttributeError
                                                Traceback (most recent call l
        ast)
        <ipython-input-16-ed003721168f> in <module>()
        ----> 1 emotion classifier.history
        AttributeError: 'Model' object has no attribute 'history'
In [ ]: from keras.preprocessing.image import img to array
        import imutils
        import cv2
        from keras.models import load_model
        import numpy as np
        # parameters for loading data and images
        detection model path = 'haarcascade frontalface default.xml'
        emotion model path = 'mini xception model1.hdf5'
        # hyper-parameters for bounding boxes shape
        # loading models
        face detection = cv2.CascadeClassifier(detection model path)
        emotion classifier = load model(emotion model path, compile=False)
         EMOTIONS = ["angry" ,"disgust", "scared", "happy", "sad", "surprised",
         "neutral"1
        # starting video streaming
```

```
cv2.namedWindow('your face')
camera = cv2.VideoCapture(0)
while True:
    frame = camera.read()[1]
    #reading the frame
   frame = imutils.resize(frame, width=400)
    gray = cv2.cvtColor(frame, cv2.COLOR BGR2GRAY)
   faces = face detection.detectMultiScale(gray,scaleFactor=1.1,minNei
ghbors=5,minSize=(30,30),flags=cv2.CASCADE SCALE IMAGE)
    canvas = np.zeros((250, 300, 3), dtype="uint8")
   frameClone = frame.copy()
   if len(faces) > 0:
        faces = sorted(faces, reverse=True,
        key=lambda x: (x[2] - x[0]) * (x[3] - x[1]))[0]
        (fX, fY, fW, fH) = faces
                    # Extract the ROI of the face from the grayscale im
age, resize it to a fixed 48x48 pixels, and then prepare
            # the ROI for classification via the CNN
        roi = gray[fY:fY + fH, fX:fX + fW]
        roi = cv2.resize(roi, (48, 48))
        roi = roi.astype("float") / 255.0
        roi = img to array(roi)
        roi = np.expand dims(roi, axis=0)
        preds = emotion classifier.predict(roi)[0]
        emotion probability = np.max(preds)
        label = EMOTIONS[preds.argmax()]
   for (i, (emotion, prob)) in enumerate(zip(EMOTIONS, preds)):
                # construct the label text
                text = "{}: {:.2f}%".format(emotion, prob * 100)
                w = int(prob * 300)
                cv2.rectangle(canvas, (7, (i * 35) + 5),
                (w, (i * 35) + 35), (0, 0, 255), -1)
                cv2.putText(canvas, text, (10, (i * 35) + 23),
                cv2.FONT HERSHEY SIMPLEX, 0.45,
```

```
(255, 255, 255), 2)
                        cv2.putText(frameClone, label, (fX, fY - 10),
                        cv2.FONT HERSHEY SIMPLEX, 0.45, (0, 0, 255), 2)
                        cv2.rectangle(frameClone, (fX, fY), (fX + fW, fY + fH),
                                       (0, 0, 255), 2)
            cv2.imshow('your face', frameClone)
            cv2.imshow("Probabilities", canvas)
            if cv2.waitKey(1) \& 0xFF == ord('q'):
                break
        camera.release()
        cv2.destroyAllWindows()
In [5]: !pip install --upgrade imutils
        Collecting imutils
          Downloading https://files.pythonhosted.org/packages/5e/0c/659c2bdae8e
        8ca5ef810b9da02db28feaa29ea448ff36b65a1664ff28142/imutils-0.5.2.tar.gz
        Building wheels for collected packages: imutils
          Running setup.py bdist wheel for imutils: started
          Running setup.py bdist wheel for imutils: finished with status 'done'
          Stored in directory: C:\Users\LENOVO\AppData\Local\pip\Cache\wheels\b
        2\40\59\139d450e68847ef2f27d876d527b13389dac23df0f66526b5d
        Successfully built imutils
        Installing collected packages: imutils
        Successfully installed imutils-0.5.2
        distributed 1.21.8 requires msqpack, which is not installed.
        You are using pip version 10.0.1, however version 19.0.3 is available.
        You should consider upgrading via the 'python -m pip install --upgrade
        pip' command.
```