NSFW CLASSIFIER

The NSFW Classifier is used to Classify Not safe for work images vs Safe images.

NSFW images includes Porn and Sexy images.

Rest are classified as Safe Images.

This model is hugely inspired by https://www.freecodecamp.org/news/how-to-set-up-nsfw-content-detection-with-machine-learning-229a9725829c/), https://github.com/GantMan/nsfw model (https://github.com/GantMan/nsfw model (https://github.com/GantMan/nsfw model (https://github.com/GantMan/nsfw model (https://github.com/GantMan/nsfw model architecture is changed.

The Data for this project is collected with help of scripts at https://github.com/alex000kim/nsfw data scraper (https://github.com/alex000kim/nsfw data scraper)

Mobile Net Architecture is used for the classification as it is very fast and has less params to train.

An ios app is also made and its screen recording is uploaded at https://github.com/lakshaychhabra/NSFW-ios-ML (https://github.com/lakshaychhabra/NSFW-ios-ML)

```
In [1]:
           import numpy as np
         2 import matplotlib.pyplot as plt
         3 import seaborn as sns
           %matplotlib inline
           from keras.preprocessing.image import ImageDataGenerator
           from keras.backend import clear session
         7 from keras.optimizers import SGD, Adam
           from pathlib import Path
           from keras.applications.mobilenet v2 import MobileNetV2
        10 from keras.models import Sequential, Model, load model
        11 from keras.layers import Dense, Dropout, Flatten, AveragePooling2D, BatchNormalization
        12 from keras import initializers, regularizers
        13 from pathlib import Path
        14 from keras.callbacks import ModelCheckpoint, TensorBoard, ReduceLROnPlateau, History, LearningRateScheduler
        15 from datetime import datetime
        16 import warnings
        17 warnings.filterwarnings("ignore")
        18 import os
        19 import matplotlib.image as mpimg
        Using TensorFlow backend.
```

```
In [2]: 1 !pwd
```

/home/lakshaychhabralc7/nsfw/data

```
In [3]:
    train_neutral = (len([iq for iq in os.scandir('/home/lakshaychhabralc7/nsfw/data/train/neutral')]))
    test_neutral = (len([iq for iq in os.scandir('/home/lakshaychhabralc7/nsfw/data/test/neutral')]))
    train_porn = (len([iq for iq in os.scandir('/home/lakshaychhabralc7/nsfw/data/train/porn')]))
    train_sexy = (len([iq for iq in os.scandir('/home/lakshaychhabralc7/nsfw/data/train/sexy')]))
    test_porn = (len([iq for iq in os.scandir('/home/lakshaychhabralc7/nsfw/data/test/porn')]))
    test_sexy = (len([iq for iq in os.scandir('/home/lakshaychhabralc7/nsfw/data/test/sexy')]))
```

```
In [4]: 1 train_data = [train_neutral, train_porn, train_sexy]
2 test_data = [test_neutral, test_porn, test_sexy]
```

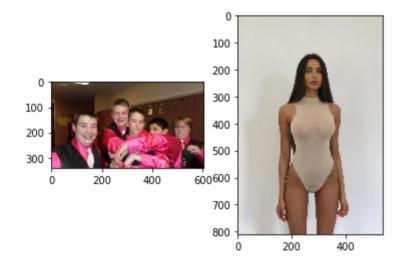
```
In [5]: 1 print("Total number of train data is : ", train_data[0], "+", train_data[1], "+", train_data[2],"=", sum(tr
2 print("Total number of test data is : ", test_data[0], "+", test_data[1], "+", test_data[2],"=", sum(test_d

Total number of train data is : 34387 + 55581 + 17191 = 107159
Total number of test data is : 2000 + 2000 + 2000 = 6000
In [6]: 1 train_path =r"/home/lakshaychhabralc7/nsfw/data/train"
2 test_path = r"/home/lakshaychhabralc7/nsfw/data/test"
```

So we Have 107k images to train and have 6k images to test

Example of the data Neutral and Sexy category

Out[7]: <matplotlib.image.AxesImage at 0x7f79ec883588>



We need a fast model which gives high accuracy and also have less parsms to train. We are choosing MobileNet for it.

We will use Transfer Learning and choose weights which were trained for ImageNet.

```
In [8]:  # As we know the input size in ImageNet was 224 so we have to resize our images accordingly
    size = 224
    epochs = 100
    steps = 500
```

```
In [9]:
         1 # We have to take in account different angle of images and to avoid overfit we will use Data Generator,
         2 | # More the Merrier
           train data generation = ImageDataGenerator(
                rescale=1./255,
                rotation range=30,
         5
                width shift range=0.2,
         6
         7
                height shift range=0.2,
         8
                shear range=0.2,
                zoom range=0.2,
         9
                channel shift range=20,
        10
                horizontal flip=True
        11
        12 )
        validation data generation = ImageDataGenerator(
        14
                rescale=1./255 #need float values
        15 )
```

```
train generator = train data generation.flow from directory(
In [10]:
           2
                     train path,
                     target size=(size, size),
           3
                     class mode='categorical',
                     batch size = 64
           5
           6
           7
             validation generator = validation data generation.flow from directory(
          9
                 test path,
                 target size=(size, size),
         10
                 class mode='categorical',
         11
         12
                 batch size = 64
         13 )
```

Found 107159 images belonging to 3 classes. Found 6000 images belonging to 3 classes.

DL Model

```
1 conv m = MobileNetV2(weights='imagenet', include top=False, input shape=(size, size, 3))
In [12]:
            conv m.trainable = False
          3 conv m.summary()
         block 15 add (Add)
                                          (None, 7, 7, 160)
                                                               0
                                                                           block 14 add[0][0]
                                                                           block 15 project BN[0][0]
         block 16 expand (Conv2D)
                                          (None, 7, 7, 960)
                                                                           block 15_add[0][0]
                                                               153600
         block 16 expand BN (BatchNormal (None, 7, 7, 960)
                                                                           block 16 expand[0][0]
                                                               3840
                                                                           block 16 expand_BN[0][0]
         block 16 expand relu (ReLU)
                                          (None, 7, 7, 960)
                                                               0
         block 16 depthwise (DepthwiseCo (None, 7, 7, 960)
                                                                           block 16 expand relu[0][0]
                                                               8640
         block 16 depthwise BN (BatchNor (None, 7, 7, 960)
                                                               3840
                                                                           block 16 depthwise[0][0]
         block 16 depthwise relu (ReLU)
                                         (None, 7, 7, 960)
                                                                           block 16 depthwise BN[0][0]
         block 16 project (Conv2D)
                                          (None, 7, 7, 320)
                                                               307200
                                                                           block 16 depthwise relu[0][0]
         block 16 project BN (BatchNorma (None, 7, 7, 320)
                                                               1280
                                                                           block 16 project[0][0]
         Conv 1 (Conv2D)
                                          (None, 7, 7, 1280)
                                                               409600
                                                                           block 16 project BN[0][0]
In [ ]:
          1
```

```
In [36]: 1  model = Sequential()
2  model.add(conv_m)
3  model.add(AveragePooling2D(pool_size=(7, 7)))
4  model.add(Flatten())
5  model.add(Dense(32, activation = 'relu'))
6  model.add(BatchNormalization())
7  model.add(Dropout(0.5))
8  model.add(Dense(3, activation='softmax'))
9  model.summary()
```

```
Output Shape
                                                         Param #
Layer (type)
mobilenetv2 1.00 224 (Model) (None, 7, 7, 1280)
                                                         2257984
average pooling2d 5 (Average (None, 1, 1, 1280)
                                                         0
flatten 7 (Flatten)
                              (None, 1280)
                                                         0
dense 19 (Dense)
                                                         40992
                              (None, 32)
batch normalization 7 (Batch (None, 32)
                                                         128
dropout 11 (Dropout)
                              (None, 32)
                                                         0
dense 20 (Dense)
                                                         99
                              (None, 3)
Total params: 2,299,203
Trainable params: 41,155
Non-trainable params: 2,258,048
```

time taken: 2:58:14.032359

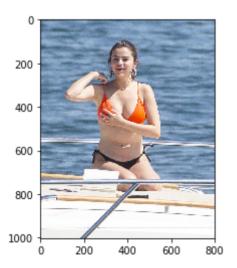
```
In [17]:
      1 start = datetime.now()
       history = model.fit generator(
      3
          train generator,
          callbacks=callbacks,
      4
      5
          epochs=100,
      6
          steps per epoch=10,
      7
          validation data=validation generator,
      8
          validation steps=10,
      9
          initial epoch = 30
     10 )
     acc: 0.9047
     Epoch 00073: val acc did not improve from 0.92344
     Epoch 74/100
     acc: 0.9000
     Epoch 00074: val acc did not improve from 0.92344
     Epoch 75/100
     acc: 0.9172
     Epoch 00075: val acc did not improve from 0.92344
     Epoch 76/100
     acc: 0.9125
     Epoch 00076: val acc did not improve from 0.92344
      1 print("time taken : ", datetime.now() - start)
In [18]:
```

```
In [19]:
      1 history = model.fit generator(
      2
           train generator,
       3
           callbacks=callbacks,
           epochs=100,
       4
       5
           steps per epoch=10,
       6
           validation data=validation generator,
      7
           validation steps=10,
       8
           initial epoch = 78
       9
      Epoch 00093: val acc did not improve from 0.93590
     Epoch 94/100
     acc: 0.9187
     Epoch 00094: val acc did not improve from 0.93590
     Epoch 95/100
     acc: 0.9094
     Epoch 00095: val_acc did not improve from 0.93590
     Epoch 96/100
     acc: 0.9215
     Epoch 00096: val acc did not improve from 0.93590
     Epoch 97/100
     acc: 0.9141
In [43]:
      1 | # model.save("nsfwnsfw mobilenet2 100.h5")
In [ ]:
In [91]:
      1 model = load model("weights00077.h5")
In [93]:
      1 import coremitools
      2 model.author = "Lakshay Chhabra"
      3 model.short description = "NSFW Image Classifier"
```

```
1 output labels = ['Neutral', 'Porn', 'Sexy']
In [941:
          2 ios = coremltools.converters.keras.convert(model, input names=['image'], output names=['output'],
                                                                class labels=output labels, image input names='image', i
          3
           4
         48933320>
         116: mobilenetv2 1.00 224 block 12 add, <keras.layers.merge.Add object at 0x1c48933438>
         117: mobilenetv2 1.00 224 block 13 expand, <keras.layers.convolutional.Conv2D object at 0x1c48933470>
         118: mobilenetv2 1.00 224 block 13 expand BN, <keras.layers.normalization.BatchNormalization object at 0x1c4
         89335f8>
         119: mobilenetv2 1.00 224 block 13 expand relu, <keras.layers.advanced activations.ReLU object at 0x1c489337
         10>
         120 : mobilenetv2 1.00 224 block 13 pad, <keras.layers.convolutional.ZeroPadding2D object at 0x1c48933748>
         121: mobilenetv2 1.00 224 block 13 depthwise, <keras.layers.convolutional.DepthwiseConv2D object at 0x1c4893
         3780>
         122 : mobilenetv2 1.00 224 block 13 depthwise BN, <keras.layers.normalization.BatchNormalization object at 0x
         1c489337f0>
         123 : mobilenetv2 1.00 224 block 13 depthwise relu, <keras.layers.advanced activations.ReLU object at 0x1c489
         33ac8>
         124 : mobilenetv2 1.00 224 block 13 project, <keras.layers.convolutional.Conv2D object at 0x1c48933b00>
         125 : mobilenetv2 1.00 224 block 13 project BN, <keras.layers.normalization.BatchNormalization object at 0x1c
         48933c88>
         126: mobilenetv2 1.00 224 block 14 expand, <keras.layers.convolutional.Conv2D object at 0x1c48933da0>
         127 : mobilenetv2 1.00 224 block 14 expand BN, <keras.layers.normalization.BatchNormalization object at 0x1c4
In [95]:
          1 ios.save('NSFW.mlmodel')
In [ ]:
          1
In [69]:
          1 test file = test path+"/sexy/ffc15b09-10a0-4753-9adf-d38eb53cf8a1.jpg"
```

```
In [105]:
           1 # https://stackoverflow.com/a/53403805/7437264
             from PIL import Image
             import numpy as np
              from skimage import transform
              def load(filename):
                  np image = Image.open(filename)
            6
           7
                  np image = np.array(np image).astype('float32')/255
            8
                  np image = transform.resize(np image, (224, 224, 3))
           9
                  np image = np.expand dims(np image, axis=0)
                  img=mpimg.imread(filename)
          10
                  plt.imshow(img)
          11
          12
                  return np image
          13
          14
              image = load("8.jpg")
          15 ans = model.predict(image)
              maping = {0 : "Neutral", 1 : "Porn", 2 : "Sexy"}
              new ans = np.argmax(ans[0])
          17
          18
          19
              print(maping[new ans], np.round(ans,2))
          20 print("With {} probability".format(ans[0][new ans]))
```

Sexy [[0.02 0.02 0.96]] With 0.9588373303413391 probability



Summary

- 1. This Model unable to classify drawings and Anime as it is not trained for them.
- 2. It fails to classify Male genitalia because images in train data are mostly of females.
- 3. The accuracy can further be improved as I was limited by resources and don't have a GPU, so further training can increase accuracy.
- 4. Future Goals: To add bounded box with help or YOLO or Sliding window or any other object detection algo and classify video data live.