

DESCRIPTION

Do you have your father's nose?

Blood relatives often share facial features. Now researchers at Northeastern University want to improve their algorithm for facial image classification to bridge the gap between research and other familial markers like DNA results. That will be challenge in this new Kaggle competition.

An automatic kinship classifier has been in the works at Northeastern since 2010. Yet this technology remains largely unseen in practice for a couple of reasons:

1. Existing image databases for kinship recognition tasks aren't large enough to capture and reflect the true data distributions of the families of the world.
2. Many hidden factors affect familial facial relationships, so a more discriminant model is needed than the computer vision algorithms used most often for higher-level categorizations (e.g. facial recognition or object classification).

In this competition, you'll help researchers build a more complex model by determining if two people are blood-related based solely on images of their faces. If you think you can get it "on the nose," this competition is for you.

The SMILE Lab at Northeastern focuses on the frontier research of applied machine learning, social media analytics, human-computer interaction, and high-level image and video understanding. Their research is driven by the explosion of diverse multimedia from the Internet, including both personal and publicly-available photos and videos. They start by treating fundamental theory from learning algorithms as the soul of machine intelligence and arm it with visual perception.

EVALUATION

Evaluated on area under the ROC curve between the predicted probability and the observed target. Not all pairs will be scored.

ABOUT DATA

train-faces.zip - the training set is divided in Families (F0123), then individuals (MIDx). Images in the same MIDx folder belong to the same person. Images in the same F0123 folder belong to the same family. train.csv - training labels. Remember, not every individual in a family shares a kinship relationship. For example, a mother and father are kin to their children, but not to each other.

test-faces.zip - the test set contains face images of unknown individuals

In [1]:

```
# importing the tensorflow package
import tensorflow as tf
```

In [2]:

```
tf.__version__
```

Out[2]:

```
'2.0.0'
```

In [3]:

```
tf.test.is_built_with_cuda()
```

Out[3]:

```
True
```

In [4]:

```
print("Num GPUs Available: ", len(tf.config.experimental.list_physical_devices('GPU')))
```

```
Num GPUs Available: 0
```

GOOGLE CLOUD

In [1]:

```
from collections import defaultdict
from glob import glob
from random import choice, sample
import matplotlib.pyplot as plt
from tqdm import tqdm
import numpy as np
import pandas as pd
import pickle
import gc, psutil, random, base64, h5py, os,shutil
#import cv2
pd.options.mode.chained_assignment = None
pd.options.display.max_columns = 9999
pd.options.display.float_format = '{:20, .2f}'.format
from keras import backend as K

from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Concatenate, Multiply
, Dropout, Subtract, Add
from keras_vggface.vggface import VGGFace
from pathlib import Path

from PIL import Image
from io import BytesIO
from IPython.display import HTML
import plotly.offline as py
py.init_notebook_mode(connected=True)
import plotly.graph_objs as go
import plotly.tools as tls

import warnings
warnings.filterwarnings('ignore')
from keras.layers import Conv2D, ZeroPadding2D, Activation, concatenate
from keras.layers.pooling import MaxPooling2D, AveragePooling2D
from keras.layers.core import Lambda
import keras
from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau,EarlyStopping
from keras.layers import Conv1D, BatchNormalization,Reshape, LeakyReLU
from keras.preprocessing import image
from keras.optimizers import Adam,SGD,rmsprop
from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Conv2D,MaxPooling1D,G
lobalAveragePooling1D,LSTM
from keras.models import Model
from keras.losses import binary_crossentropy
import tensorflow as tf
from sklearn.metrics import roc_auc_score
from keras import regularizers
%pylab inline
import matplotlib.image as mpimg
```

Using TensorFlow backend.

Populating the interactive namespace from numpy and matplotlib

In [6]:

```
!sudo pip install kaggle
```

```
Requirement already satisfied: kaggle in /usr/local/lib/python3.5/dist-packages (1.5.6)
Requirement already satisfied: certifi in /usr/local/lib/python3.5/dist-packages (from kaggle) (2019.11.28)
Requirement already satisfied: python-dateutil in /usr/local/lib/python3.5/dist-packages (from kaggle) (2.8.1)
Requirement already satisfied: requests in /usr/local/lib/python3.5/dist-packages (from kaggle) (2.22.0)
Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.5/dist-packages (from kaggle) (1.13.0)
Requirement already satisfied: urllib3<1.25,>=1.21.1 in /usr/local/lib/python3.5/dist-packages (from kaggle) (1.24.2)
```

```
Requirement already satisfied: tqdm in /usr/local/lib/python3.5/dist-packages (from kaggle) (4.40.2)
Requirement already satisfied: python-slugify in /usr/local/lib/python3.5/dist-packages (from kaggle) (4.0.0)
Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3.5/dist-packages (from requests->kaggle) (3.0.4)
Requirement already satisfied: idna<2.9,>=2.5 in /usr/local/lib/python3.5/dist-packages (from requests->kaggle) (2.8)
Requirement already satisfied: text-unidecode>=1.3 in /usr/local/lib/python3.5/dist-packages (from python-slugify->kaggle) (1.3)
WARNING: You are using pip version 19.3.1; however, version 20.0.2 is available.
You should consider upgrading via the 'pip install --upgrade pip' command.
```

In [7]:

```
!sudo mkdir .kaggle
```

mkdir: cannot create directory '.kaggle': File exists

In [8]:

```
!sudo mkdir ~/.kaggle
```

mkdir: cannot create directory '/root/.kaggle': File exists

In [9]:

```
!pwd
```

In [10]:

```
!ls
```

```
driver_installer.run      SAVE_MODEL
Northeastern_GCP_submit2_1_final_1.ipynb  TENSORBOARD
Northeastern_GCP_submit2_1_test.ipynb    test
recognizing-faces-in-the-wild            train
recognizing-faces-in-the-wild.zip
```

In [11]:

```
import json
token = {
with open(
    json.dump(token, file)
```

In [12]:

```
!cp /home/.kaggle/kaggle.json ~/.kaggle/kaggle.json
```

In [13]:

```
!sudo chmod 600 ~/.kaggle/kaggle.json
```

In [14]:

```
!kaggle datasets list
```

ref	size	lastUpdated	downloadCount	voteCount	title	usabilityRating
chrisfilo/to-bee-or-no-to-bee	3GB	2020-02-04 22:55:31	47	9	To bee or not to bee	0.8125
peijenlin/msl-m-rems-2-edr-v1.0	11GB	2020-02-05 00:55:52	4	4	MSL-M-REMS-2-EDR-V1.0	0.8125
timoboz/superbowl-history-1967-2020	2KB	2020-02-03 23:41:14	279	31	Superbowl History 1967 - 2020	1.0
timoboz/google-trends-data	1MB	2020-02-04 17:31:10	244	20	Google Trends Data	0.85294116
fmejia21/demographics-of-academy-awards-oscars-winners	20KB	2020-02-04 17:38:26	153	12	Demographics of Academy Awards (Oscars) Winners	0.88235295
timoboz/big-list-of-naughty-strings	66KB	2020-02-04 18:11:53	14	10	Big List of Naughty Strings	0.8125
fmejia21/2020-democratic-primary-endorsements	22KB	2020-02-04 18:05:34	43	10	2020 Democratic Primary Endorsements	0.88235295
prondeau/diceimages	48MB	2020-02-04 18:24:40	14	10	Images of Board Game Dice	0.8125
timoboz/python-data-science-handbook	15MB	2020-02-04 18:27:14	82	12	Python Data Science Handbook	0.8235294
timoboz/clevr-dataset	18GB	2020-02-04 18:36:16	4	9	CLEVR Dataset	0.9375
peijenlin/surviving-mars-maps	4MB	2020-01-27 20:30:13	21	12	Surviving Mars Maps	1.0
chrisfilo/urbansound8k	6GB	2020-02-04 18:37:24	7	10	UrbanSound8K	0.7647059
prondeau/350000-jeopardy-questions	19MB	2020-02-04 18:53:30	50	11	350,000+ Jeopardy Questions	0.9411765
prondeau/superbowlads	36KB	2020-02-04 18:04:07	99	11	Super Bowl Ads	1.0
timoboz/can-james-bond-be-black	345B	2020-02-04 19:08:57	18	9	Can James Bond Be Black?	1.0
devvret/farm-ads-binary-classification	5MB	2020-02-04 19:12:44	14	10	Farm Ads Binary Classification	0.8125
fmejia21/nba-all-star-game-20002016	35KB	2020-02-04 19:21:13	102	13	NBA All Star Game 2000-2016	0.88235295
timoboz/wikidata-jsons	858MB	2020-02-04 19:19:08	4	8	Wikidata jsons	0.8125
chrisfilo/firesense	784MB	2020-02-04 19:20:54	16	10	FIRESENSE	0.8125
prondeau/the-car-connection-picture-dataset	682MB	2020-02-04 19:19:07	24	12	60,000+ Images of Cars	0.9375

```
In [0]:
```

```
!kaggle competitions download -c recognizing-faces-in-the-wild -p [REDACTED]
```

```
Downloading recognizing-faces-in-the-wild.zip to [REDACTED]  
98% | [REDACTED] | 373M/381M [00:02<00:00] 142MB/s
```

[illegible]

381M/381M [00:02<00:00, 151MB/s]

Northeastern GCP.ipynb recognizing-faces-in-the-wild.zip

```
!unzip recognizing-faces-in-the-wild.zip -d ./recognizing-faces-in-the-wild
```

```
Archive:  recognizing-faces-in-the-wild.zip
inflating: ./recognizing-faces-in-the-wild/README.md
inflating: ./recognizing-faces-in-the-wild/sample_submission.csv
inflating: ./recognizing-faces-in-the-wild/test-faces.zip
inflating: ./recognizing-faces-in-the-wild/test-private-faces.zip
inflating: ./recognizing-faces-in-the-wild/test-private-labels.zip
inflating: ./recognizing-faces-in-the-wild/test-private-lists.zip
inflating: ./recognizing-faces-in-the-wild/test-public-faces.zip
inflating: ./recognizing-faces-in-the-wild/test-public-lists.zip
inflating: ./recognizing-faces-in-the-wild/test.zip
inflating: ./recognizing-faces-in-the-wild/train-faces.zip
inflating: ./recognizing-faces-in-the-wild/train.zip
inflating: ./recognizing-faces-in-the-wild/train_relationships.csv
```

Northeastern GCP.ipynb recognizing-faces-in-the-wild.zip

recognizing-faces-in-the-wild

```
!sudo mkdir test
!sudo chmod 777 test
```

```
!sudo mkdir train
!sudo chmod 777 train
```

```
!unzip ./recognizing-faces-in-the-wild/test.zip -d ./test
```

```
!unzip ./recognizing-faces-in-the-wild/train.zip -d ./train
```

Northeastern GCP.ipynb recognizing-faces-in-the-wild.zip train

recognizing-faces-in-the-wild test

GOOGLE COLAB

In [6]:

```
!ls
```

sample_data

In [7]:

```
!pwd
```

/content

In [8]:

```
# we used VGGFace file from the below github which has some models works better for Face Identification
!git clone https://github.com/rcmalli/keras-vggface.git
```

```
Cloning into 'keras-vggface'...
remote: Enumerating objects: 26, done.
remote: Counting objects: 100% (26/26), done.
remote: Compressing objects: 100% (24/24), done.
remote: Total 277 (delta 10), reused 6 (delta 2), pack-reused 251
Receiving objects: 100% (277/277), 6.03 MiB | 9.04 MiB/s, done.
Resolving deltas: 100% (129/129), done.
```

In [0]:

```
import os,shutil
for file_ in os.listdir('keras-vggface'):
    if '.'* not in file_:
        try:
            shutil.move('keras-vggface/'+file_,'.')
        except:
            print(9)
```

In [10]:

```
from collections import defaultdict
from glob import glob
from random import choice, sample
import matplotlib.pyplot as plt
from tqdm import tqdm
import numpy as np
import pandas as pd
import pickle
import gc, psutil, cv2, random, base64, h5py, os,shutil
pd.options.mode.chained_assignment = None
pd.options.display.max_columns = 9999
pd.options.display.float_format = '{:20, .2f}'.format
from keras import backend as K

from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Concatenate, Multiply, Dropout, Subtract, Add
from keras_vggface.vggface import VGGFace
from pathlib import Path

from PIL import Image
from io import BytesIO
from IPython.display import HTML
import plotly.offline as py
pv.init notebook mode(connected=True)
```

```

import plotly.graph_objs as go
import plotly.tools as tls

import warnings
warnings.filterwarnings('ignore')
from keras.layers import Conv2D, ZeroPadding2D, Activation, concatenate
from keras.layers.pooling import MaxPooling2D, AveragePooling2D
from keras.layers.core import Lambda
import keras
from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau, EarlyStopping
from keras.layers import Conv1D, BatchNormalization, Reshape, LeakyReLU
from keras.preprocessing import image
from keras.optimizers import Adam, SGD, rmsprop
from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Conv2D, MaxPooling1D, GlobalAveragePooling1D, LSTM
from keras.models import Model
from keras.losses import binary_crossentropy
import tensorflow as tf
from sklearn.metrics import roc_auc_score
from keras import regularizers
%pylab inline
import matplotlib.image as mpimg

```

Using TensorFlow backend.

Populating the interactive namespace from numpy and matplotlib

In [0]:

In [0]:

In [11]:

```

import shutil
from google.colab import drive
drive.mount('/content/drive', force_remount=True)
# path='drive/My Drive/AAIC/'

```

In [0]:

```

# UPLOADING model1.h5 from local PC to colab

#from google.colab import files
#uploaded = files.upload()

#import io
#df2 = pd.read_csv(io.BytesIO(uploaded['Filename.csv']))
# Dataset is now stored in a Pandas Dataframe

```

In [12]:

```
!ls
```

```
drive keras_vggface LICENSE.txt README.md setup.py tools
image keras-vggface MANIFEST.in sample_data test.py visualize
```

In [13]:

```
!pwd
```

```
/content
```

In [0]:

In [14]:

```
!pip install kaggle
```

```
Requirement already satisfied: kaggle in /usr/local/lib/python3.6/dist-packages (1.5.6)
Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.6/dist-packages (from kaggle) (1.12.0)
Requirement already satisfied: tqdm in /usr/local/lib/python3.6/dist-packages (from kaggle) (4.28.1)
Requirement already satisfied: certifi in /usr/local/lib/python3.6/dist-packages (from kaggle) (2019.11.28)
Requirement already satisfied: requests in /usr/local/lib/python3.6/dist-packages (from kaggle) (2.21.0)
Requirement already satisfied: python-dateutil in /usr/local/lib/python3.6/dist-packages (from kaggle) (2.6.1)
Requirement already satisfied: python-slugify in /usr/local/lib/python3.6/dist-packages (from kaggle) (4.0.0)
Requirement already satisfied: urllib3<1.25,>=1.21.1 in /usr/local/lib/python3.6/dist-packages (from kaggle) (1.24.3)
Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist-packages (from requests->kaggle) (3.0.4)
Requirement already satisfied: idna<2.9,>=2.5 in /usr/local/lib/python3.6/dist-packages (from requests->kaggle) (2.8)
Requirement already satisfied: text-unidecode>=1.3 in /usr/local/lib/python3.6/dist-packages (from python-slugify->kaggle) (1.3)
```

In [0]:

```
!mkdir .kaggle
```

In [0]:

```
!mkdir ~/.kaggle
```

In [0]:

In [0]:

```
import json
token = {
with open('kaggle.json', 'w') as file:
    json.dump(token, file)
```

In [0]:

```
!cp /content/.kaggle/kaggle.json ~/.kaggle/kaggle.json
```

In [19]:

```
!kaggle config set -n path -v{/content}
```


Warning: Your Kaggle API key is readable by other users on this system! To fix this, you can run 'chmod 600 /root/.kaggle/kaggle.json'
- path is now set to: {/content}

In [0]:

```
!chmod 600 /root/.kaggle/kaggle.json
```

In [21]:

```
!kaggle datasets list
```

Warning: Looks like you're using an outdated API Version, please consider updating (server 1.5.6 / client 1.5.4)

ref		title
size	lastUpdated	downloadCount
-----	-----	-----
timoboz/google-trends-data		Google Trends Data
1MB	2020-02-04 17:31:10	204
fmejia21/demographics-of-academy-awards-oscar-winners		Demographics of Academy Awards (Oscars) Winners
20KB	2020-02-04 17:38:26	135
timoboz/big-list-of-naughty-strings		Big List of Naughty Strings
66KB	2020-02-04 18:11:53	13
fmejia21/2020-democratic-primary-endorsements		2020 Democratic Primary Endorsements
22KB	2020-02-04 18:05:34	38
prondeau/diceimages		Images of Board Game Dice
48MB	2020-02-04 18:24:40	8
timoboz/python-data-science-handbook		Python Data Science Handbook
15MB	2020-02-04 18:27:14	67
timoboz/clevr-dataset		CLEVR Dataset
18GB	2020-02-04 18:36:16	4
peijenlin/surviving-mars-maps		Surviving Mars Maps
4MB	2020-01-27 20:30:13	19
chrisfilo/urbansound8k		UrbanSound8K
6GB	2020-02-04 18:37:24	7
prondeau/350000-jeopardy-questions		350,000+ Jeopardy Questions
19MB	2020-02-04 18:53:30	45
prondeau/superbowlads		Super Bowl Ads
36KB	2020-02-04 18:04:07	90
timoboz/can-james-bond-be-black		Can James Bond Be Black?
345B	2020-02-04 19:08:57	9
devvret/farm-ads-binary-classification		Farm Ads Binary Classification
5MB	2020-02-04 19:12:44	11
fmejia21/nba-all-star-game-2000-2016		NBA All Star Game 2000-2016
35KB	2020-02-04 19:21:13	78
timoboz/wikidata-jsons		Wikidata jsons
858MB	2020-02-04 19:19:08	4
chrisfilo/firesense		FIRESENSE
784MB	2020-02-04 19:20:54	13
prondeau/the-car-connection-picture-dataset		60,000+ Images of Cars
682MB	2020-02-04 19:19:07	19
fmejia21/trump-impeachment-polls		Trump Impeachment Polls
110KB	2020-02-04 19:29:05	48
chrisfilo/demand		DEMAND
7GB	2020-02-04 19:41:36	4
timoboz/data-science-cheat-sheets		Data Science Cheat Sheets
596MB	2020-02-04 19:42:27	150

In [22]:

```
!kaggle datasets list -s Northeastern
```

Warning: Looks like you're using an outdated API Version, please consider updating (server 1.5.6 / client 1.5.4)

ref	title	size	lastUpdated	downloadCount
-----	-----	-----	-----	-----
marcodena/mobile-phone-activity	Mobile phone activity in a city	242MB	2019-11-14 06:11:32	8084
theriley106/college-common-data-sets	College Common Data Sets	210MB	2018-01-20 22:14:19	

In [23]:

```
#!kaggle competitions download -c recognizing-faces-in-the-wild -p /content  
!kaggle competitions download -c recognizing-faces-in-the-wild -p /content
```

Warning: Looks like you're using an outdated API Version, please consider updating (server 1.5.6 / client 1.5.4)

Downloading test-private-faces.zip to /content

71% 21.0M/29.7M [00:00<00:00, 20.8MB/s]

100% 29.7M/29.7M [00:00<00:00, 60.6MB/s]

README.md: Skipping, found more recently modified local copy (use --force to force download)

Downloading test-public-faces.zip to /content

89% 116M/130M [00:01<00:00, 63.5MB/s]

100% 130M/130M [00:01<00:00, 81.6MB/s]

Downloading test-public-lists.zip to /content

0% 0.00/33.7k [00:00<?, ?B/s]

100% 33.7k/33.7k [00:00<00:00, 69.5MB/s]

Downloading test.zip to /content

56% 19.0M/34.1M [00:00<00:00, 18.7MB/s]

100% 34.1M/34.1M [00:00<00:00, 58.0MB/s]

Downloading test-private-lists.zip to /content

0% 0.00/354k [00:00<?, ?B/s]

100% 354k/354k [00:00<00:00, 111MB/s]

Downloading test-faces.zip to /content

33% 9.00M/27.1M [00:00<00:01, 17.9MB/s]

100% 27.1M/27.1M [00:00<00:00, 46.3MB/s]

Downloading test-private-labels.zip to /content

0% 0.00/8.67k [00:00<?, ?B/s]

100% 8.67k/8.67k [00:00<00:00, 8.88MB/s]

Downloading train_relationships.csv to /content

0% 0.00/77.6k [00:00<?, ?B/s]

100% 77.6k/77.6k [00:00<00:00, 68.4MB/s]

Downloading train.zip to /content

95% 65.0M/68.6M [00:00<00:00, 45.4MB/s]

100% 68.6M/68.6M [00:00<00:00, 78.5MB/s]

Downloading sample_submission.csv to /content

0% 0.00/156k [00:00<?, ?B/s]

100% 156k/156k [00:00<00:00, 133MB/s]

Downloading train-faces.zip to /content

83% 92.0M/111M [00:01<00:00, 38.3MB/s]

100% 111M/111M [00:01<00:00, 61.5MB/s]

In [24]:

```
!ls
```

```
drive          sample_submission.csv test.py  
image          setup.py      test.zip  
keras_vggface  test-faces.zip tools  
keras_vggface  test-private-faces.zip train-faces.zip  
LICENSE.txt    test-private-labels.zip train_relationships.csv  
MANIFEST.in    test-private-lists.zip train.zip  
README.md      test-public-faces.zip visualize  
sample_data    test-public-lists.zip
```

In [0]:

```
!mkdir test  
!chmod 777 test
```

In [0]:

```
!mkdir train  
!chmod 777 train
```

In [27]:

```
%%time
%%capture
!unzip test.zip -d test
```

CPU times: user 553 ms, sys: 53.2 ms, total: 606 ms
Wall time: 4.33 s

In [28]:

```
%%time
%%capture
!unzip train.zip -d train
```

CPU times: user 1.41 s, sys: 184 ms, total: 1.59 s
Wall time: 6.56 s

In [29]:

```
!ls
```

```
drive          setup.py  test.zip
image          test     tools
keras_vggface  test-faces.zip  train
keras-vggface  test-private-faces.zip train-faces.zip
LICENSE.txt    test-private-labels.zip train_relationships.csv
MANIFEST.in    test-private-lists.zip train.zip
README.md      test-public-faces.zip visualize
sample_data    test-public-lists.zip
sample_submission.csv test.py
```

In [30]:

```
print("The no of train images:",len(os.listdir('train')))
# here it contains both the folders and pics
```

The no of train images: 470

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [4]:

```
!pip install git+https://github.com/rcmalli/keras-vggface.git
```

```
Collecting git+https://github.com/rcmalli/keras-vggface.git
  Cloning https://github.com/rcmalli/keras-vggface.git to /tmp/pip-req-build-awgjtqb8
  Running command git clone -q https://github.com/rcmalli/keras-vggface.git /tmp/pip-req-build-awgjtqb8
Requirement already satisfied (use --upgrade to upgrade): keras-vggface==0.6 from git+https://github.com/rcmalli/keras-vggface.git in /home/deeplearning/anaconda3/lib/python3.7/site-packages
Requirement already satisfied: numpy>=1.9.1 in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras-vggface==0.6) (1.17.2)
Requirement already satisfied: scipy>=0.14 in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras-vggface==0.6) (1.3.1)
Requirement already satisfied: h5py in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras-vggface==0.6) (2.9.0)
Requirement already satisfied: pillow in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras-vggface==0.6) (6.2.0)
Requirement already satisfied: keras in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras-vggface==0.6) (2.3.1)
Requirement already satisfied: six>=1.9.0 in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras-vggface==0.6) (1.12.0)
Requirement already satisfied: pyyaml in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras-vggface==0.6) (5.1.2)
Requirement already satisfied: keras-preprocessing>=1.0.5 in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras->keras-vggface==0.6) (1.1.0)
Requirement already satisfied: keras-applications>=1.0.6 in /home/deeplearning/anaconda3/lib/python3.7/site-packages (from keras->keras-vggface==0.6) (1.0.8)
Building wheels for collected packages: keras-vggface
  Building wheel for keras-vggface (setup.py) ... done
  Created wheel for keras-vggface: filename=keras_vggface-0.6-cp37-none-any.whl size=8312 sha256=827030c2d77efc6ce7c2d83f73db137135a7ecale6628a005d99f48b2193e2f5
  Stored in directory: /tmp/pip-ephem-wheel-cache-cqplb64/wheels/36/07/46/06c25ce8e9cd396dabe151ea1d8a2bc28dafcb11321c1f3a6d
Successfully built keras-vggface
WARNING: You are using pip version 19.3.1; however, version 20.0.2 is available.
You should consider upgrading via the 'pip install --upgrade pip' command.
```

In [5]:

```
from collections import defaultdict
from glob import glob
from random import choice, sample

#import cv2
import numpy as np
import pandas as pd
from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau
```

```

from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau
from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Concatenate, Multiply
, Dropout, Subtract, Add, Conv2D
from keras.models import Model
from keras.preprocessing import image
from keras.optimizers import Adam
from keras_vggface.utils import preprocess_input
from keras_vggface.vggface import VGGFace

# import regularizer
from keras.regularizers import l1
from keras.regularizers import l2

import h5py
import itertools

```

EDA

In [6]:

```

import matplotlib.pyplot as plt
from PIL import Image
from pathlib import Path
import os

import plotly.offline as py
py.init_notebook_mode(connected=True)
import plotly.graph_objs as go
import plotly.tools as tls

pd.options.mode.chained_assignment = None
pd.options.display.max_columns = 9999
pd.options.display.float_format = '{:20, .2f}'.format

```

In [9]:

```

files = [os.path.join(dp, f) for dp, dn, fn in os.walk(os.path.expanduser("./train")) for f in fn]
images_df = pd.DataFrame({
    'files': files,
    'familyId': [file.split('/')[2] for file in files],
    'kinId': [file.split('/')[3] for file in files],
    'uniqueId': [file.split('/')[2] + '/' + file.split('/')[3] for file in files]
})
images_df.head()

```

Out[9]:

	familyId	files	kinId	uniqueId
0	F0435	./train/F0435/MD5/P04588_face5.jpg	MD5	F0435/MD5
1	F0435	./train/F0435/MD5/P04592_face3.jpg	MD5	F0435/MD5
2	F0435	./train/F0435/MD5/P04585_face2.jpg	MD5	F0435/MD5
3	F0435	./train/F0435/MD5/P04584_face4.jpg	MD5	F0435/MD5
4	F0435	./train/F0435/MD5/P04583_face2.jpg	MD5	F0435/MD5

In [10]:

```

print("Total number of members in the dataset: {}".format(images_df["uniqueId"].nunique()))
print("Total number of families in the dataset: {}".format(images_df["familyId"].nunique()))

```

Total number of members in the dataset: 2316
Total number of families in the dataset: 470

In [11]:

```
family_with_most_pic = images_df["familyId"].value_counts()
kin_with_most_pic = images_df["uniqueId"].value_counts()
print("Family with maximum number of images: {0}, Image Count: {1}".format(family_with_most_pic.index[0], family_with_most_pic[0]))
print("Member with maximum number of images: {0}, Image Count: {1}".format(kin_with_most_pic.index[0], kin_with_most_pic[0]))
```

Family with maximum number of images: F0601, Image Count: 776
 Member with maximum number of images: F0601/MID6, Image Count: 95

In [12]:

```
family_series = family_with_most_pic[:25]
labels = (np.array(family_series.index))
sizes = (np.array((family_series / family_with_most_pic.sum()) * 100))

trace = go.Pie(labels=labels, values=sizes)
layout = go.Layout(title='Pic Count by Families')
data = [trace]
fig = go.Figure(data=data, layout=layout)
py.iplot(fig, filename='Families')
```

PREPARING DATA

In []:

In [13]:

```
#train_file_path = "./recognizing-faces-in-the-wild/train_relationships.csv"
train_file_path = "./train_relationships.csv"
train_folders_path = "./train/"
test_path = "./test/"
sample_sub_df=pd.read_csv('sample_submission.csv')
# Considering F09 for validation
val_families = "F09"
```

In [14]:

In [14]:

```
# all_images will have path of all jpg with
all_images = glob(train_folders_path + "**/*/*.jpg")
print(all_images[0])

# For windows 10 issue using following comment above
#from wcmatch import glob
#all_images = glob.glob(train_folders_path + "**\\*\\*\\*\\.jpg")
#all_images = glob.glob(train_folders_path + "**/*/*.jpg")
#print(all_images[0])
```

./train/F0435/MID5/P04588_face5.jpg

In [15]:

```
# In train_images collecting path of all training data
# Path consist of Family name and person name as well
train_images = [x for x in all_images if val_families not in x]

# In val_images collecting path of all validation data
# Path consist of Family name and person name as well
val_images = [x for x in all_images if val_families in x]
```

In [16]:

```
# ppl contains Family name and person name
ppl = [x.split("/")[-3] + "/" + x.split("/")[-2] for x in all_images]
# For windows 10 comment above use following
#ppl = [x.split("\\")[-3] + "/" + x.split("\\")[-2] for x in all_images]
print(ppl[0])
```

F0435/MID5

In [17]:

```
# There are 2 for loops to create
# --> key as Family and person id
# --> value as image complete path of that person
train_person_to_images_map = defaultdict(list)
for x in train_images:
    train_person_to_images_map[x.split("/")[-3] + "/" + x.split("/")[-2]].append(x)
    #train_person_to_images_map[x.split("\\")[-3] + "/" + x.split("\\")[-2]].append(x)

val_person_to_images_map = defaultdict(list)
for x in val_images:
    val_person_to_images_map[x.split("/")[-3] + "/" + x.split("/")[-2]].append(x)
    #val_person_to_images_map[x.split("\\")[-3] + "/" + x.split("\\")[-2]].append(x)
```

In [18]:

```
#print(train_person_to_images_map.keys())
print(train_person_to_images_map['F0002/MID1'])
print(val_person_to_images_map['F0900/MID1'])
```

```
['./train/F0002/MID1/P00017_face3.jpg', './train/F0002/MID1/P00009_face3.jpg', './train/F0002/MID1/P00010_face4.jpg', './train/F0002/MID1/P00011_face1.jpg', './train/F0002/MID1/P00016_face2.jpg', './train/F0002/MID1/P00012_face2.jpg', './train/F0002/MID1/P00018_face1.jpg', './train/F0002/MID1/P00014_face2.jpg', './train/F0002/MID1/P00015_face2.jpg', './train/F0002/MID1/P00013_face2.jpg']
['./train/F0900/MID1/P09509_face1.jpg', './train/F0900/MID1/P09513_face1.jpg', './train/F0900/MID1/P09505_face1.jpg', './train/F0900/MID1/P09506_face1.jpg', './train/F0900/MID1/P09508_face1.jpg']
```

In [19]:

```
# Creating relationship dataframe from train_relationships.csv
# Then considering only those values which are present in ppl
relationships = pd.read_csv(train_file_path)
print('#'* 50)
print(relationships.head())
relationships = list(zip(relationships.p1.values, relationships.p2.values))
```

```

print('#'* 50)
print(relationships[0])
print(len(relationships))
relationships = [x for x in relationships if x[0] in ppl and x[1] in ppl]
print('#'* 50)
print(relationships[0])
print(len(relationships))
print('#'* 50)

```

```

#####
          p1          p2
0  F0002/MID1  F0002/MID3
1  F0002/MID2  F0002/MID3
2  F0005/MID1  F0005/MID2
3  F0005/MID3  F0005/MID2
4  F0009/MID1  F0009/MID4
#####
('F0002/MID1', 'F0002/MID3')
3598
#####
('F0002/MID1', 'F0002/MID3')
3362
#####

```

In [20]:

```

# Now Creating train and val
train = [x for x in relationships if val_families not in x[0]]
print('train data sample', train[0])
val = [x for x in relationships if val_families in x[0]]
print('validation data sample', val[0])

```

```

train data sample ('F0002/MID1', 'F0002/MID3')
validation data sample ('F0900/MID2', 'F0900/MID1')

```

In []:

In []:

In []:

In [21]:

```

batch_tuples_test = sample(train, 8)
print(batch_tuples_test)
labels_test = [1] * len(batch_tuples_test)
print(labels_test)

```

```

[('F0748/MID1', 'F0748/MID3'), ('F0829/MID6', 'F0829/MID1'), ('F0693/MID2', 'F0693/MID4'), ('F0376/MID8', 'F0376/MID4'), ('F0421/MID3', 'F0421/MID7'), ('F0568/MID1', 'F0568/MID4'), ('F0174/MID5', 'F0174/MID7'), ('F0287/MID3', 'F0287/MID4')]
[1, 1, 1, 1, 1, 1, 1, 1]

```

In [22]:

```

ppl_test = list(train_person_to_images_map.keys())
p1 = choice(ppl_test)
p2 = choice(ppl_test)

print(p1)
print(p2)

```


F0387/MID2
F0209/MID4

In [23]:

```
print((train_person_to_images_map['F0488/MID1']))
print(len(train_person_to_images_map['F0488/MID1']))
print(choice(train_person_to_images_map['F0488/MID1']))

print('*' * 50)
for x in batch_tuples_test:
    print(x[0])
    print((train_person_to_images_map[x[0]]))
    print(len(train_person_to_images_map[x[0]]))
    print(choice(train_person_to_images_map[x[0]]))
```

```
['./train/F0488/MID1/P05139_face3.jpg', './train/F0488/MID1/P05138_face2.jpg', './train/F0488/MID1/P05132_face1.jpg', './train/F0488/MID1/P05134_face1.jpg', './train/F0488/MID1/P05133_face1.jpg']
5
./train/F0488/MID1/P05138_face2.jpg
*****
F0748/MID1
['./train/F0748/MID1/P07842_face2.jpg', './train/F0748/MID1/P07840_face3.jpg', './train/F0748/MID1/P07834_face2.jpg', './train/F0748/MID1/P07835_face2.jpg', './train/F0748/MID1/P07836_face2.jpg', './train/F0748/MID1/P07841_face1.jpg', './train/F0748/MID1/P07837_face1.jpg', './train/F0748/MID1/P07838_face2.jpg', './train/F0748/MID1/P07839_face2.jpg']
9
./train/F0748/MID1/P07842_face2.jpg
F0829/MID6
['./train/F0829/MID6/P08762_face2.jpg', './train/F0829/MID6/P08763_face2.jpg', './train/F0829/MID6/P08770_face5.jpg', './train/F0829/MID6/P08766_face1.jpg']
4
./train/F0829/MID6/P08762_face2.jpg
F0693/MID2
['./train/F0693/MID2/P07225_face2.jpg', './train/F0693/MID2/P07220_face2.jpg', './train/F0693/MID2/P07222_face2.jpg']
3
./train/F0693/MID2/P07220_face2.jpg
F0376/MID8
['./train/F0376/MID8/P10643_face1.jpg', './train/F0376/MID8/P10625_face4.jpg', './train/F0376/MID8/P10629_face1.jpg', './train/F0376/MID8/P10640_face1.jpg', './train/F0376/MID8/P10630_face2.jpg', './train/F0376/MID8/P10626_face5.jpg', './train/F0376/MID8/P10627_face1.jpg', './train/F0376/MID8/P10636_face1.jpg', './train/F0376/MID8/P10642_face1.jpg', './train/F0376/MID8/P10637_face3.jpg', './train/F0376/MID8/P10628_face1.jpg', './train/F0376/MID8/P10641_face4.jpg']
12
./train/F0376/MID8/P10627_face1.jpg
F0421/MID3
['./train/F0421/MID3/P04422_face7.jpg', './train/F0421/MID3/P04425_face3.jpg', './train/F0421/MID3/P04428_face4.jpg', './train/F0421/MID3/P04429_face1.jpg', './train/F0421/MID3/P04423_face5.jpg', './train/F0421/MID3/P04430_face3.jpg']
6
./train/F0421/MID3/P04425_face3.jpg
F0568/MID1
['./train/F0568/MID1/P05975_face1.jpg', './train/F0568/MID1/P05980_face2.jpg', './train/F0568/MID1/P05971_face2.jpg', './train/F0568/MID1/P05977_face1.jpg', './train/F0568/MID1/P05972_face1.jpg', './train/F0568/MID1/P05982_face2.jpg', './train/F0568/MID1/P05981_face1.jpg', './train/F0568/MID1/P05978_face1.jpg', './train/F0568/MID1/P05979_face1.jpg', './train/F0568/MID1/P05973_face2.jpg', './train/F0568/MID1/P05983_face3.jpg', './train/F0568/MID1/P05974_face1.jpg']
12
./train/F0568/MID1/P05977_face1.jpg
F0174/MID5
['./train/F0174/MID5/P01869_face4.jpg', './train/F0174/MID5/P01874_face1.jpg', './train/F0174/MID5/P01873_face2.jpg']
3
./train/F0174/MID5/P01869_face4.jpg
F0287/MID3
['./train/F0287/MID3/P03069_face3.jpg', './train/F0287/MID3/P03063_face1.jpg', './train/F0287/MID3/P03067_face2.jpg', './train/F0287/MID3/P03064_face1.jpg']
4
./train/F0287/MID3/P03064_face1.jpg
```

In [24]:

```
df = pd.read_csv("./train_relationships.csv")
```

```
df = pd.read_csv("../data/relationships.csv")
df.head()
```

Out[24]:

	p1	p2
0	F0002/MD1	F0002/MD3
1	F0002/MD2	F0002/MD3
2	F0005/MD1	F0005/MD2
3	F0005/MD3	F0005/MD2
4	F0009/MD1	F0009/MD4

In [25]:

```
new = df["p1"].str.split("/", n = 1, expand = True)

# making separate first name column from new data frame
df["Family"] = new[0]
# making separate last name column from new data frame
df["Person1"] = new[1]

# Dropping old Name columns
df.drop(columns = ["p1"], inplace = True)

new = df["p2"].str.split("/", n = 1, expand = True)

# making separate first name column from new data frame
df["Family2"] = new[0]
# making separate last name column from new data frame
df["Person2"] = new[1]

# Dropping old Name columns
df.drop(columns = ["p2"], inplace = True)
df.head()
```

Out[25]:

	Family	Person1	Family2	Person2
0	F0002	MD1	F0002	MD3
1	F0002	MD2	F0002	MD3
2	F0005	MD1	F0005	MD2
3	F0005	MD3	F0005	MD2
4	F0009	MD1	F0009	MD4

In [26]:

```
del df['Family2']
```

In [27]:

```
df['Related'] = 1

#Creating a dictionary, and storing members of each family
df_dict = {}
for index, row in df.iterrows():
    if row['Family'] in df_dict:
        df_dict[row['Family']].append(row['Person1'])
    else:
        df_dict[row['Family']] = [row['Person1']]

#For each family in this dictionary, we'll first make pairs of people
#For each pair, we'll check if they're related in our existing Dataset
#If they're not in the dataframe, means we'll create a row with both persons and related value 0
i=1
for key in df_dict:
```

```

pair = list(itertools.combinations(df_dict[key], 2))
for item in pair:
    if len(df[(df['Family']==key) & (df['Person1']==item[0]) & (df['Person2']==item[1])]) == 0 \
    and len(df[(df['Family']==key) & (df['Person1']==item[1]) & (df['Person2']==item[0])]) == 0:
        new = {'Family':key, 'Person1':item[0], 'Person2':item[1], 'Related':0}
        df=df.append(new, ignore_index=True)

#Storing rows only where Person1 and Person2 are not same
df = df[(df['Person1']!=df['Person2'])]

#len(df[(df['Related']==1)])

print(df['Related'].value_counts())

```

```

1    3598
0    1720
Name: Related, dtype: int64

```

In [28]:

```
df.dtypes
```

Out[28]:

```

Family      object
Person1     object
Person2     object
Related     int64
dtype: object

```

In [29]:

```
df.describe
```

Out[29]:

```

<bound method NDFrame.describe of      Family Person1 Person2  Related
0      F0002    MID1    MID3        1
1      F0002    MID2    MID3        1
2      F0005    MID1    MID2        1
3      F0005    MID3    MID2        1
4      F0009    MID1    MID4        1
...      ...      ...      ...      ...
6285   F0203    MID4    MID8        0
6286   F0203    MID4    MID9        0
6289   F0203    MID7    MID8        0
6290   F0203    MID7    MID9        0
6292   F0203    MID8    MID9        0

[5318 rows x 4 columns]>

```

In [30]:

```
df[df['Related']==0].head()
```

Out[30]:

	Family	Person1	Person2	Related
3598	F0656	MD1	MD2	0
3599	F0656	MD1	MD4	0
3601	F0464	MD1	MD2	0
3603	F0821	MD1	MD3	0
3605	F0821	MD2	MD4	0

CHECKING FEW IMAGES

In [0]:

```
img_path = Path('./train/')
```

In [57]:

```
dir = os.path.join(img_path / df.Family[0] / df.Person1[0])
img_list = os.listdir(dir)
print(img_list)
```

```
['P00016_face2.jpg', 'P00015_face2.jpg', 'P00014_face2.jpg', 'P00011_face1.jpg', 'P00018_face1.jpg', 'P00009_face3.jpg', 'P00013_face2.jpg', 'P00017_face3.jpg', 'P00010_face4.jpg', 'P00012_face2.jpg']
```

In [58]:

```
fig, ax = plt.subplots(2, 5, figsize=(50, 20))

for i in range(len(img_list)):
    with open(img_path / df.Family[0] / df.Person1[0] / img_list[i], 'rb') as f:
        img = Image.open(f)
        ax[i%2][i//2].imshow(img)
fig.show()
```



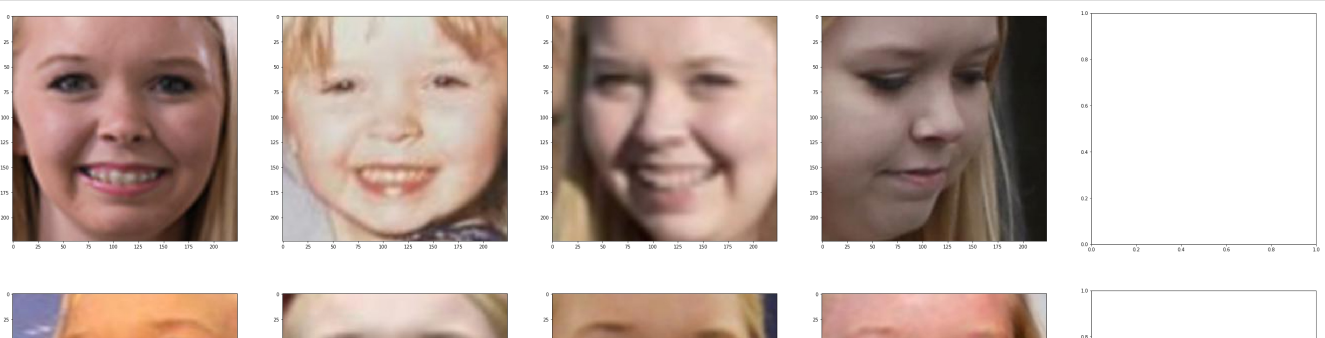
In [0]:

```
img_list = os.listdir(img_path / df.Family[0] / df.Person2[0])
```

In [60]:

```
fig, ax = plt.subplots(2, 5, figsize=(50, 20))

for i in range(len(img_list)):
    with open(img_path / df.Family[0] / df.Person2[0] / img_list[i], 'rb') as f:
        img = Image.open(f)
        ax[i%2][i//2].imshow(img)
fig.show()
```





In [0]:

In [0]:

Defining functions

In [31]:

```
# Function for reading images
def read_img(path, IMG_SIZE):
    img = image.load_img(path, target_size=(IMG_SIZE, IMG_SIZE))
    img = np.array(img).astype(np.float)
    return preprocess_input(img, version=2)
```

In [32]:

```
def auc(y_true, y_pred):
    return tf.py_function(roc_auc_score, (y_true, y_pred), tf.double)
```

In [33]:

```
def get_thumbnail(path):
    if path and os.path.exists(path):
        i = Image.open(path)
        i.thumbnail((150, 150), Image.LANCZOS)
        return i

def image_base64(im):
    if isinstance(im, str):
        im = get_thumbnail(im)
    with BytesIO() as buffer:
        im.save(buffer, 'jpeg')
        return base64.b64encode(buffer.getvalue()).decode()

def add_image_path(x):
    image_path = 'train/' + x
    if os.path.exists(image_path):
        path = os.path.join(image_path, os.listdir(image_path)[0])
        return path
```

In [34]:

```
# https://www.kaggle.com/janpreets/just-another-feature-extractor-0-824-1b
def outer_product(x):
    """
    calculate outer-products of 2 tensors

    args
        x
            list of 2 tensors
            , assuming each of which has shape = (size_minibatch, total_pixels, size_filter)
    """
    return keras.backend.batch_dot(
        x[0]
        , x[1]
        , axes=[1,1]
```

```

) / x[0].get_shape().as_list()[1]

def signed_sqrt(x):
    """
    calculate element-wise signed square root

    args
        x
            a tensor
    """
    return keras.backend.sign(x) * keras.backend.sqrt(keras.backend.abs(x) + 1e-9)

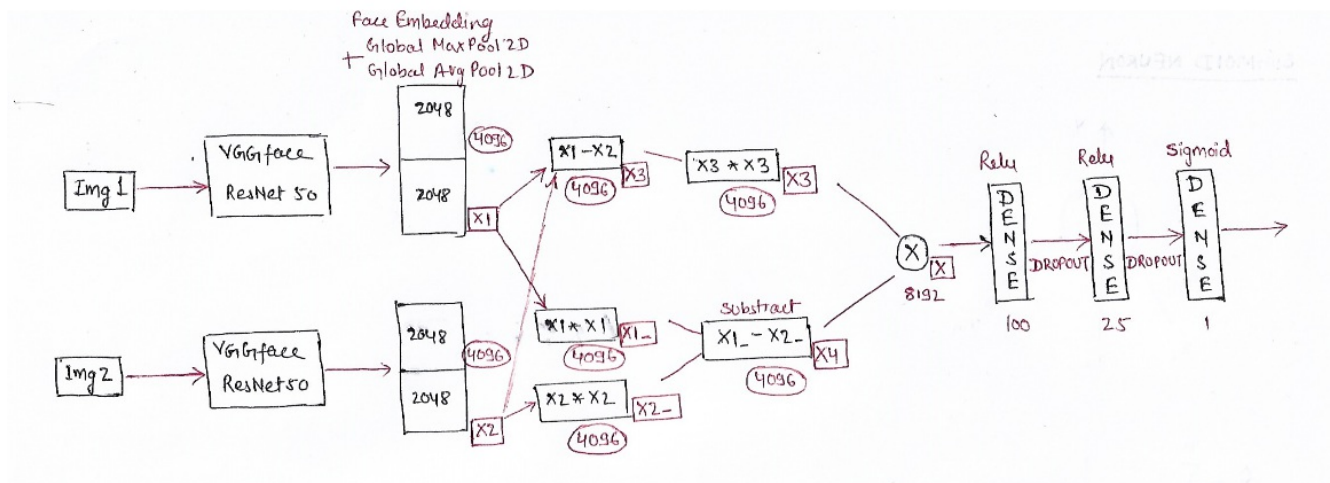
def L2_norm(x, axis=-1):
    """
    calculate L2-norm

    args
        x
            a tensor
    """
    return keras.backend.l2_normalize(x, axis=axis)

```

In [0]:

MODEL 1



In [0]:

```

# Function gen is to generate data for label 0
# example --> gen(train_person_to_images_map, batch_size=16)
# train_person_to_images_map is dictionary with family and person as key and values are image path
#
def gen(list_tuples, person_to_images_map, batch_size=16):
    # All keys of dictionary stored in
    ppl = list(person_to_images_map.keys())
    while True:
        # list_tuples is train data from relationship csv file
        # batch_tuples contains sample from train data (from relationship.csv)
        batch_tuples = sample(list_tuples, batch_size // 2)
        # If batch_size is of size 16 then batch_tuples will have 8 samples
        labels = [1] * len(batch_tuples)
        while len(batch_tuples) < batch_size:
            # choice randomly picking data from ppl
            p1 = choice(ppl)
            p2 = choice(ppl)
            # Randomly Creating data with labels 0
            if p1 != p2 and (p1, p2) not in list_tuples and (p2, p1) not in list_tuples:
                batch_tuples.append((p1, p2))
                labels.append(0)
        # for each family person combination from batch_tuples

```

```

# x[0] means considering first entry for ex --> ('F0488/MID1', 'F0488/MID4') --> consider 'F04
88/MID1'
# Check in dictioanry train_person_to_images_map
for x in batch_tuples:
    if not len(person_to_images_map[x[0]]):
        print(x[0])

# Considering randomly images using choice function
X1 = [choice(person_to_images_map[x[0]]) for x in batch_tuples]
X1 = np.array([read_img(x, 197) for x in X1])

X2 = [choice(person_to_images_map[x[1]]) for x in batch_tuples]
X2 = np.array([read_img(x, 197) for x in X2])

yield [X1, X2], labels

```

In [0]:

```

# Preparing baseline_model1
def baseline_model1():
    input_1 = Input(shape=(197, 197, 3))
    input_2 = Input(shape=(197, 197, 3))

    base_model1 = VGGFace(model='resnet50', include_top=False)

    for x in base_model1.layers[:-3]:
        x.trainable = True
    # for x in base_model1.layers[-3:]:
    #     x.trainable=False

    x1 = base_model1(input_1)
    x2 = base_model1(input_2)

    x1 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x1), GlobalAvgPool2D() (x1)])
    x2 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x2), GlobalAvgPool2D() (x2)])

    x3 = Subtract() ([x1, x2])
    x3 = Multiply() ([x3, x3])

    x1_ = Multiply() ([x1, x1])
    x2_ = Multiply() ([x2, x2])
    x4 = Subtract() ([x1_, x2_])
    x = Concatenate(axis=-1) ([x4, x3])

    x = Dense(100, activation="relu", activity_regularizer=l1(0.001)) (x)
    x = Dropout(0.3) (x)
    x = Dense(25, activation="relu", activity_regularizer=l1(0.001)) (x)
    x = Dropout(0.01) (x)
    out = Dense(1, activation="sigmoid") (x)

    model1 = Model([input_1, input_2], out)

    model1.compile(loss="binary_crossentropy", metrics=['acc'], optimizer=Adam(0.00001))

    model1.summary()

    return model1

```

In [0]:

```

file_path1 = "./model1.h5"

checkpoint1 = ModelCheckpoint(file_path1, monitor='val_acc', verbose=1, save_best_only=True, mode='max'
)

reduce_on_plateau1 = ReduceLROnPlateau(monitor="val_acc", mode="max", factor=0.1, patience=20, verbose=
1)

callbacks_list1 = [checkpoint1, reduce_on_plateau1]

model1 = baseline_model1()

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:66:
The name tf.get_default_graph is deprecated. Please use tf.compat.v1.get_default_graph instead

The name tf.get_default_graph is deprecated. Please use tf.compat.v1.get_default_graph instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:541:
The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4432
: The name tf.random_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:190:
The name tf.get_default_session is deprecated. Please use tf.compat.v1.get_default_session instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:197:
The name tf.ConfigProto is deprecated. Please use tf.compat.v1.ConfigProto instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:203:
The name tf.Session is deprecated. Please use tf.compat.v1.Session instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:207:
The name tf.global_variables is deprecated. Please use tf.compat.v1.global_variables instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:216:
The name tf.is_variable_initialized is deprecated. Please use tf.compat.v1.is_variable_initialized instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:223:
The name tf.variables_initializer is deprecated. Please use tf.compat.v1.variables_initializer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:2041
: The name tf.nn.fused_batch_norm is deprecated. Please use tf.compat.v1.nn.fused_batch_norm instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:148:
The name tf.placeholder_with_default is deprecated. Please use tf.compat.v1.placeholder_with_default instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4267
: The name tf.nn.max_pool is deprecated. Please use tf.nn.max_pool2d instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4271
: The name tf.nn.avg_pool is deprecated. Please use tf.nn.avg_pool2d instead.

Downloading data from https://github.com/rcmalli/keras-vggface/releases/download/v2.0/rcmalli_vggface_tf_notop_resnet50.h5

94699520/94694792 [=====] - 1s 0us/step

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3733
: calling dropout (from tensorflow.python.ops.nn_ops) with keep_prob is deprecated and will be removed
in a future version.

Instructions for updating:

Please use `rate` instead of `keep_prob`. Rate should be set to `rate = 1 - keep_prob`.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3657
: The name tf.log is deprecated. Please use tf.math.log instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow_core/python/ops/nn_impl.py:183:
where (from tensorflow.python.ops.array_ops) is deprecated and will be removed in a future version.
Instructions for updating:

Use tf.where in 2.0, which has the same broadcast rule as np.where

Model: "model_1"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 197, 197, 3)	0	
input_2 (InputLayer)	(None, 197, 197, 3)	0	
vggface_resnet50 (Model)	multiple	23561152	input_1[0][0] input_2[0][0]
global_max_pooling2d_1 (GlobalM	(None, 2048)	0	vggface_resnet50[1][0]
global_average_pooling2d_1 (Glo	(None, 2048)	0	vggface_resnet50[1][0]
global_max_pooling2d_2 (GlobalM	(None, 2048)	0	vggface_resnet50[2][0]
global_average_pooling2d_2 (Glo	(None, 2048)	0	vggface_resnet50[2][0]
concatenate_1 (Concatenate)	(None, 4096)	0	global_max_pooling2d_1[0][0]

concatenate_1 (Concatenate)	(None, 4096)	0	global_max_pooling2d_1[0][0] global_average_pooling2d_1[0][0]
concatenate_2 (Concatenate)	(None, 4096)	0	global_max_pooling2d_2[0][0] global_average_pooling2d_2[0][0]
multiply_2 (Multiply)	(None, 4096)	0	concatenate_1[0][0] concatenate_1[0][0]
multiply_3 (Multiply)	(None, 4096)	0	concatenate_2[0][0] concatenate_2[0][0]
subtract_1 (Subtract)	(None, 4096)	0	concatenate_1[0][0] concatenate_2[0][0]
subtract_2 (Subtract)	(None, 4096)	0	multiply_2[0][0] multiply_3[0][0]
multiply_1 (Multiply)	(None, 4096)	0	subtract_1[0][0] subtract_1[0][0]
concatenate_3 (Concatenate)	(None, 8192)	0	subtract_2[0][0] multiply_1[0][0]
dense_1 (Dense)	(None, 100)	819300	concatenate_3[0][0]
dropout_1 (Dropout)	(None, 100)	0	dense_1[0][0]
dense_2 (Dense)	(None, 25)	2525	dropout_1[0][0]
dropout_2 (Dropout)	(None, 25)	0	dense_2[0][0]
dense_3 (Dense)	(None, 1)	26	dropout_2[0][0]
=====			
Total params: 24,383,003			
Trainable params: 24,329,883			
Non-trainable params: 53,120			
=====			

In [0]:

```
model1.fit_generator(gen(train, train_person_to_images_map, batch_size=16), use_multiprocessing=True,
                      validation_data=gen(val, val_person_to_images_map, batch_size=16), epochs=100, verb
ose=1,
                      workers = 4, callbacks=callbacks_list1, steps_per_epoch=300, validation_steps=100)
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:1033 : The name tf.assign_add is deprecated. Please use tf.compat.v1.assign_add instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:1020 : The name tf.assign is deprecated. Please use tf.compat.v1.assign instead.

Epoch 1/100
300/300 [=====] - 155s 517ms/step - loss: 4.2482 - acc: 0.5408 - val_loss: 4.5607 - val_acc: 0.5850

Epoch 00001: val_acc improved from -inf to 0.58500, saving model to ./model1.h5
Epoch 2/100
300/300 [=====] - 129s 430ms/step - loss: 0.9521 - acc: 0.5585 - val_loss: 1.6322 - val_acc: 0.5944

Epoch 00002: val_acc improved from 0.58500 to 0.59437, saving model to ./model1.h5
Epoch 3/100
300/300 [=====] - 129s 429ms/step - loss: 0.7854 - acc: 0.5708 - val_loss: 1.3397 - val_acc: 0.5719

Epoch 00003: val_acc did not improve from 0.59437
Epoch 4/100
300/300 [=====] - 129s 429ms/step - loss: 0.7376 - acc: 0.5721 - val_loss: 1.3048 - val_acc: 0.5600

Epoch 00004: val_acc did not improve from 0.59437
Epoch 5/100
300/300 [=====] - 129s 430ms/step - loss: 0.7180 - acc: 0.5767 - val_loss: 0.9126 - val_acc: 0.5787

Epoch 00005: val_acc did not improve from 0.59437
Epoch 6/100
300/300 [=====] - 129s 430ms/step - loss: 0.7020 - acc: 0.5985 - val_loss: 0.9738 - val_acc: 0.5837

Epoch 00006: val_acc did not improve from 0.59437
Epoch 7/100
300/300 [=====] - 129s 429ms/step - loss: 0.6957 - acc: 0.6127 - val_loss: 0.8859 - val_acc: 0.5925

Epoch 00007: val_acc did not improve from 0.59437
Epoch 8/100
300/300 [=====] - 128s 426ms/step - loss: 0.6853 - acc: 0.6231 - val_loss: 0.7978 - val_acc: 0.6319

Epoch 00008: val_acc improved from 0.59437 to 0.63187, saving model to ./model1.h5
Epoch 9/100
300/300 [=====] - 128s 425ms/step - loss: 0.6802 - acc: 0.6450 - val_loss: 0.8370 - val_acc: 0.6269

Epoch 00009: val_acc did not improve from 0.63187
Epoch 10/100
300/300 [=====] - 128s 426ms/step - loss: 0.6785 - acc: 0.6415 - val_loss: 0.7451 - val_acc: 0.6362

Epoch 00010: val_acc improved from 0.63187 to 0.63625, saving model to ./model1.h5
Epoch 11/100
300/300 [=====] - 128s 427ms/step - loss: 0.6742 - acc: 0.6473 - val_loss: 0.7798 - val_acc: 0.6469

Epoch 00011: val_acc improved from 0.63625 to 0.64687, saving model to ./model1.h5
Epoch 12/100
300/300 [=====] - 128s 427ms/step - loss: 0.6692 - acc: 0.6571 - val_loss: 0.7349 - val_acc: 0.6425

Epoch 00012: val_acc did not improve from 0.64687
Epoch 13/100
300/300 [=====] - 129s 429ms/step - loss: 0.6644 - acc: 0.6727 - val_loss: 0.6964 - val_acc: 0.6438

Epoch 00013: val_acc did not improve from 0.64687
Epoch 14/100
300/300 [=====] - 129s 430ms/step - loss: 0.6614 - acc: 0.6792 - val_loss: 0.7064 - val_acc: 0.6488

Epoch 00014: val_acc improved from 0.64687 to 0.64875, saving model to ./model1.h5
Epoch 15/100
300/300 [=====] - 129s 429ms/step - loss: 0.6586 - acc: 0.6931 - val_loss: 0.6794 - val_acc: 0.6456

Epoch 00015: val_acc did not improve from 0.64875
Epoch 16/100
300/300 [=====] - 128s 427ms/step - loss: 0.6553 - acc: 0.6977 - val_loss: 0.6722 - val_acc: 0.6750

Epoch 00016: val_acc improved from 0.64875 to 0.67500, saving model to ./model1.h5
Epoch 17/100
300/300 [=====] - 129s 430ms/step - loss: 0.6505 - acc: 0.7106 - val_loss: 0.6761 - val_acc: 0.6919

Epoch 00017: val_acc improved from 0.67500 to 0.69188, saving model to ./model1.h5
Epoch 18/100
300/300 [=====] - 128s 428ms/step - loss: 0.6471 - acc: 0.7127 - val_loss: 0.6755 - val_acc: 0.7050

Epoch 00018: val_acc improved from 0.69188 to 0.70500, saving model to ./model1.h5
Epoch 19/100
300/300 [=====] - 128s 428ms/step - loss: 0.6463 - acc: 0.7160 - val_loss: 0.6808 - val_acc: 0.6863

Epoch 00019: val_acc did not improve from 0.70500
Epoch 20/100
300/300 [=====] - 128s 427ms/step - loss: 0.6377 - acc: 0.7296 - val_loss: 0.6669 - val_acc: 0.7244

Epoch 00020: val_acc improved from 0.70500 to 0.72437, saving model to ./model1.h5
Epoch 21/100

```
Epoch 21/100
300/300 [=====] - 128s 427ms/step - loss: 0.6307 - acc: 0.7498 - val_loss: 0.6530 - val_acc: 0.7181

Epoch 00021: val_acc did not improve from 0.72437
Epoch 22/100
300/300 [=====] - 129s 428ms/step - loss: 0.6288 - acc: 0.7587 - val_loss: 0.6589 - val_acc: 0.7050

Epoch 00022: val_acc did not improve from 0.72437
Epoch 23/100
300/300 [=====] - 128s 427ms/step - loss: 0.6242 - acc: 0.7500 - val_loss: 0.6598 - val_acc: 0.7300

Epoch 00023: val_acc improved from 0.72437 to 0.73000, saving model to ./model1.h5
Epoch 24/100
300/300 [=====] - 129s 429ms/step - loss: 0.6244 - acc: 0.7533 - val_loss: 0.6483 - val_acc: 0.7206

Epoch 00024: val_acc did not improve from 0.73000
Epoch 25/100
300/300 [=====] - 128s 428ms/step - loss: 0.6176 - acc: 0.7663 - val_loss: 0.6410 - val_acc: 0.7581

Epoch 00025: val_acc improved from 0.73000 to 0.75813, saving model to ./model1.h5
Epoch 26/100
300/300 [=====] - 129s 429ms/step - loss: 0.6158 - acc: 0.7733 - val_loss: 0.6311 - val_acc: 0.7638

Epoch 00026: val_acc improved from 0.75813 to 0.76375, saving model to ./model1.h5
Epoch 27/100
300/300 [=====] - 128s 427ms/step - loss: 0.6122 - acc: 0.7758 - val_loss: 0.6455 - val_acc: 0.7331

Epoch 00027: val_acc did not improve from 0.76375
Epoch 28/100
300/300 [=====] - 129s 429ms/step - loss: 0.6057 - acc: 0.7906 - val_loss: 0.6285 - val_acc: 0.7538

Epoch 00028: val_acc did not improve from 0.76375
Epoch 29/100
300/300 [=====] - 129s 431ms/step - loss: 0.6094 - acc: 0.7748 - val_loss: 0.6349 - val_acc: 0.7512

Epoch 00029: val_acc did not improve from 0.76375
Epoch 30/100
300/300 [=====] - 129s 429ms/step - loss: 0.5974 - acc: 0.7967 - val_loss: 0.6207 - val_acc: 0.7638

Epoch 00030: val_acc did not improve from 0.76375
Epoch 31/100
300/300 [=====] - 129s 429ms/step - loss: 0.5969 - acc: 0.7944 - val_loss: 0.6296 - val_acc: 0.7550

Epoch 00031: val_acc did not improve from 0.76375
Epoch 32/100
300/300 [=====] - 129s 428ms/step - loss: 0.5907 - acc: 0.8096 - val_loss: 0.6224 - val_acc: 0.7638

Epoch 00032: val_acc did not improve from 0.76375
Epoch 33/100
300/300 [=====] - 128s 427ms/step - loss: 0.5872 - acc: 0.8081 - val_loss: 0.6253 - val_acc: 0.7719

Epoch 00033: val_acc improved from 0.76375 to 0.77187, saving model to ./model1.h5
Epoch 34/100
300/300 [=====] - 128s 426ms/step - loss: 0.5854 - acc: 0.8123 - val_loss: 0.6266 - val_acc: 0.7550

Epoch 00034: val_acc did not improve from 0.77187
Epoch 35/100
300/300 [=====] - 128s 425ms/step - loss: 0.5852 - acc: 0.8104 - val_loss: 0.5975 - val_acc: 0.7925

Epoch 00035: val_acc improved from 0.77187 to 0.79250, saving model to ./model1.h5
Epoch 36/100
300/300 [=====] - 128s 426ms/step - loss: 0.5848 - acc: 0.8115 - val_loss: 0.6091 - val_acc: 0.7619
```

0.791 val_acc: 0.7913

Epoch 00036: val_acc did not improve from 0.79250

Epoch 37/100

300/300 [=====] - 128s 426ms/step - loss: 0.5727 - acc: 0.8242 - val_loss: 0.6007 - val_acc: 0.7913

Epoch 00037: val_acc did not improve from 0.79250

Epoch 38/100

300/300 [=====] - 128s 426ms/step - loss: 0.5665 - acc: 0.8262 - val_loss: 0.6043 - val_acc: 0.7819

Epoch 00038: val_acc did not improve from 0.79250

Epoch 39/100

300/300 [=====] - 128s 427ms/step - loss: 0.5618 - acc: 0.8242 - val_loss: 0.5861 - val_acc: 0.7744

Epoch 00039: val_acc did not improve from 0.79250

Epoch 40/100

300/300 [=====] - 128s 425ms/step - loss: 0.5561 - acc: 0.8346 - val_loss: 0.5872 - val_acc: 0.7725

Epoch 00040: val_acc did not improve from 0.79250

Epoch 41/100

300/300 [=====] - 128s 427ms/step - loss: 0.5505 - acc: 0.8333 - val_loss: 0.6160 - val_acc: 0.7438

Epoch 00041: val_acc did not improve from 0.79250

Epoch 42/100

300/300 [=====] - 128s 427ms/step - loss: 0.5491 - acc: 0.8377 - val_loss: 0.6024 - val_acc: 0.7656

Epoch 00042: val_acc did not improve from 0.79250

Epoch 43/100

300/300 [=====] - 128s 427ms/step - loss: 0.5395 - acc: 0.8485 - val_loss: 0.5810 - val_acc: 0.7856

Epoch 00043: val_acc did not improve from 0.79250

Epoch 44/100

300/300 [=====] - 128s 426ms/step - loss: 0.5415 - acc: 0.8465 - val_loss: 0.6042 - val_acc: 0.7669

Epoch 00044: val_acc did not improve from 0.79250

Epoch 45/100

300/300 [=====] - 128s 427ms/step - loss: 0.5343 - acc: 0.8421 - val_loss: 0.5973 - val_acc: 0.7731

Epoch 00045: val_acc did not improve from 0.79250

Epoch 46/100

300/300 [=====] - 128s 427ms/step - loss: 0.5278 - acc: 0.8525 - val_loss: 0.5805 - val_acc: 0.7819

Epoch 00046: val_acc did not improve from 0.79250

Epoch 47/100

300/300 [=====] - 128s 427ms/step - loss: 0.5235 - acc: 0.8508 - val_loss: 0.5891 - val_acc: 0.7738

Epoch 00047: val_acc did not improve from 0.79250

Epoch 48/100

300/300 [=====] - 128s 427ms/step - loss: 0.5158 - acc: 0.8615 - val_loss: 0.5995 - val_acc: 0.7781

Epoch 00048: val_acc did not improve from 0.79250

Epoch 49/100

300/300 [=====] - 128s 426ms/step - loss: 0.5201 - acc: 0.8544 - val_loss: 0.5905 - val_acc: 0.7919

Epoch 00049: val_acc did not improve from 0.79250

Epoch 50/100

300/300 [=====] - 128s 426ms/step - loss: 0.5112 - acc: 0.8608 - val_loss: 0.5837 - val_acc: 0.7750

Epoch 00050: val_acc did not improve from 0.79250

Epoch 51/100

300/300 [=====] - 128s 426ms/step - loss: 0.5040 - acc: 0.8677 - val_loss: 0.6004 - val_acc: 0.7600

Epoch 00051: val_acc did not improve from 0.79250

```
Epoch 00051: val_acc did not improve from 0.79250
Epoch 52/100
300/300 [=====] - 128s 426ms/step - loss: 0.4935 - acc: 0.8746 - val_loss: 0.5
881 - val_acc: 0.7688

Epoch 00052: val_acc did not improve from 0.79250
Epoch 53/100
300/300 [=====] - 128s 426ms/step - loss: 0.5065 - acc: 0.8560 - val_loss: 0.5
871 - val_acc: 0.7719

Epoch 00053: val_acc did not improve from 0.79250
Epoch 54/100
300/300 [=====] - 128s 427ms/step - loss: 0.4937 - acc: 0.8702 - val_loss: 0.5
628 - val_acc: 0.7950

Epoch 00054: val_acc improved from 0.79250 to 0.79500, saving model to ./model1.h5
Epoch 55/100
300/300 [=====] - 128s 426ms/step - loss: 0.4875 - acc: 0.8765 - val_loss: 0.5
982 - val_acc: 0.7756

Epoch 00055: val_acc did not improve from 0.79500
Epoch 56/100
300/300 [=====] - 128s 426ms/step - loss: 0.4818 - acc: 0.8744 - val_loss: 0.5
798 - val_acc: 0.7612

Epoch 00056: val_acc did not improve from 0.79500
Epoch 57/100
300/300 [=====] - 128s 426ms/step - loss: 0.4845 - acc: 0.8700 - val_loss: 0.5
742 - val_acc: 0.7969

Epoch 00057: val_acc improved from 0.79500 to 0.79688, saving model to ./model1.h5
Epoch 58/100
300/300 [=====] - 128s 428ms/step - loss: 0.4745 - acc: 0.8860 - val_loss: 0.6
051 - val_acc: 0.7669

Epoch 00058: val_acc did not improve from 0.79688
Epoch 59/100
300/300 [=====] - 127s 425ms/step - loss: 0.4660 - acc: 0.8881 - val_loss: 0.5
793 - val_acc: 0.7812

Epoch 00059: val_acc did not improve from 0.79688
Epoch 60/100
300/300 [=====] - 128s 426ms/step - loss: 0.4691 - acc: 0.8835 - val_loss: 0.5
726 - val_acc: 0.7850

Epoch 00060: val_acc did not improve from 0.79688
Epoch 61/100
300/300 [=====] - 128s 425ms/step - loss: 0.4658 - acc: 0.8810 - val_loss: 0.5
958 - val_acc: 0.7506

Epoch 00061: val_acc did not improve from 0.79688
Epoch 62/100
300/300 [=====] - 128s 425ms/step - loss: 0.4594 - acc: 0.8881 - val_loss: 0.5
752 - val_acc: 0.7856

Epoch 00062: val_acc did not improve from 0.79688
Epoch 63/100
300/300 [=====] - 127s 425ms/step - loss: 0.4547 - acc: 0.8917 - val_loss: 0.5
800 - val_acc: 0.7844

Epoch 00063: val_acc did not improve from 0.79688
Epoch 64/100
300/300 [=====] - 128s 426ms/step - loss: 0.4544 - acc: 0.8900 - val_loss: 0.5
865 - val_acc: 0.7706

Epoch 00064: val_acc did not improve from 0.79688
Epoch 65/100
300/300 [=====] - 128s 426ms/step - loss: 0.4543 - acc: 0.8896 - val_loss: 0.5
769 - val_acc: 0.7788

Epoch 00065: val_acc did not improve from 0.79688
Epoch 66/100
300/300 [=====] - 128s 425ms/step - loss: 0.4478 - acc: 0.8958 - val_loss: 0.5
975 - val_acc: 0.7556

Epoch 00066: val_acc did not improve from 0.79688
Epoch 67/100
300/300 [=====] - 127s 425ms/step - loss: 0.4417 - acc: 0.8988 - val_loss: 0.6
```

```
300/300 [=====] - 127s 425ms/step - loss: 0.4417 - acc: 0.8988 - val_loss: 0.6224 - val_acc: 0.7394

Epoch 00067: val_acc did not improve from 0.79688
Epoch 68/100
300/300 [=====] - 127s 424ms/step - loss: 0.4381 - acc: 0.8994 - val_loss: 0.6112 - val_acc: 0.7450

Epoch 00068: val_acc did not improve from 0.79688
Epoch 69/100
300/300 [=====] - 127s 424ms/step - loss: 0.4322 - acc: 0.9010 - val_loss: 0.6082 - val_acc: 0.7562

Epoch 00069: val_acc did not improve from 0.79688
Epoch 70/100
300/300 [=====] - 127s 425ms/step - loss: 0.4372 - acc: 0.9025 - val_loss: 0.5923 - val_acc: 0.7619

Epoch 00070: val_acc did not improve from 0.79688
Epoch 71/100
300/300 [=====] - 128s 426ms/step - loss: 0.4243 - acc: 0.9017 - val_loss: 0.5851 - val_acc: 0.7700

Epoch 00071: val_acc did not improve from 0.79688
Epoch 72/100
300/300 [=====] - 127s 425ms/step - loss: 0.4233 - acc: 0.9083 - val_loss: 0.5632 - val_acc: 0.7869

Epoch 00072: val_acc did not improve from 0.79688
Epoch 73/100
300/300 [=====] - 127s 425ms/step - loss: 0.4200 - acc: 0.9133 - val_loss: 0.5976 - val_acc: 0.7675

Epoch 00073: val_acc did not improve from 0.79688
Epoch 74/100
300/300 [=====] - 128s 425ms/step - loss: 0.4189 - acc: 0.9102 - val_loss: 0.6240 - val_acc: 0.7400

Epoch 00074: val_acc did not improve from 0.79688
Epoch 75/100
300/300 [=====] - 128s 425ms/step - loss: 0.4147 - acc: 0.9115 - val_loss: 0.6113 - val_acc: 0.7450

Epoch 00075: val_acc did not improve from 0.79688
Epoch 76/100
300/300 [=====] - 127s 425ms/step - loss: 0.4096 - acc: 0.9125 - val_loss: 0.5819 - val_acc: 0.7731

Epoch 00076: val_acc did not improve from 0.79688
Epoch 77/100
300/300 [=====] - 128s 425ms/step - loss: 0.4022 - acc: 0.9160 - val_loss: 0.5917 - val_acc: 0.7475

Epoch 00077: val_acc did not improve from 0.79688

Epoch 00077: ReduceLROnPlateau reducing learning rate to 9.999999747378752e-07.
Epoch 78/100
300/300 [=====] - 128s 425ms/step - loss: 0.4029 - acc: 0.9154 - val_loss: 0.5991 - val_acc: 0.7588

Epoch 00078: val_acc did not improve from 0.79688
Epoch 79/100
300/300 [=====] - 127s 425ms/step - loss: 0.3964 - acc: 0.9231 - val_loss: 0.6198 - val_acc: 0.7300

Epoch 00079: val_acc did not improve from 0.79688
Epoch 80/100
300/300 [=====] - 128s 426ms/step - loss: 0.3930 - acc: 0.9269 - val_loss: 0.6234 - val_acc: 0.7331

Epoch 00080: val_acc did not improve from 0.79688
Epoch 81/100
300/300 [=====] - 127s 425ms/step - loss: 0.3876 - acc: 0.9210 - val_loss: 0.6119 - val_acc: 0.7362

Epoch 00081: val_acc did not improve from 0.79688
Epoch 82/100
300/300 [=====] - 127s 426ms/step - loss: 0.3857 - acc: 0.9250 - val_loss: 0.6119 - val_acc: 0.7362
```

300/300 [=====] - 128s 426ms/step - loss: 0.3857 - acc: 0.9250 - val_loss: 0.6117 - val_acc: 0.7400

Epoch 00082: val_acc did not improve from 0.79688

Epoch 83/100

300/300 [=====] - 128s 426ms/step - loss: 0.3857 - acc: 0.9325 - val_loss: 0.5927 - val_acc: 0.7588

Epoch 00083: val_acc did not improve from 0.79688

Epoch 84/100

300/300 [=====] - 127s 425ms/step - loss: 0.3806 - acc: 0.9310 - val_loss: 0.6024 - val_acc: 0.7388

Epoch 00084: val_acc did not improve from 0.79688

Epoch 85/100

300/300 [=====] - 127s 425ms/step - loss: 0.3845 - acc: 0.9290 - val_loss: 0.6155 - val_acc: 0.7544

Epoch 00085: val_acc did not improve from 0.79688

Epoch 86/100

300/300 [=====] - 127s 424ms/step - loss: 0.3814 - acc: 0.9304 - val_loss: 0.5907 - val_acc: 0.7631

Epoch 00086: val_acc did not improve from 0.79688

Epoch 87/100

300/300 [=====] - 128s 425ms/step - loss: 0.3812 - acc: 0.9354 - val_loss: 0.5776 - val_acc: 0.7669

Epoch 00087: val_acc did not improve from 0.79688

Epoch 88/100

300/300 [=====] - 127s 425ms/step - loss: 0.3763 - acc: 0.9350 - val_loss: 0.6244 - val_acc: 0.7412

Epoch 00088: val_acc did not improve from 0.79688

Epoch 89/100

300/300 [=====] - 128s 425ms/step - loss: 0.3823 - acc: 0.9279 - val_loss: 0.6221 - val_acc: 0.7288

Epoch 00089: val_acc did not improve from 0.79688

Epoch 90/100

300/300 [=====] - 127s 425ms/step - loss: 0.3678 - acc: 0.9425 - val_loss: 0.6074 - val_acc: 0.7356

Epoch 00090: val_acc did not improve from 0.79688

Epoch 91/100

300/300 [=====] - 128s 428ms/step - loss: 0.3742 - acc: 0.9373 - val_loss: 0.6088 - val_acc: 0.7525

Epoch 00091: val_acc did not improve from 0.79688

Epoch 92/100

300/300 [=====] - 128s 428ms/step - loss: 0.3798 - acc: 0.9342 - val_loss: 0.5930 - val_acc: 0.7494

Epoch 00092: val_acc did not improve from 0.79688

Epoch 93/100

300/300 [=====] - 128s 427ms/step - loss: 0.3753 - acc: 0.9329 - val_loss: 0.5945 - val_acc: 0.7538

Epoch 00093: val_acc did not improve from 0.79688

Epoch 94/100

300/300 [=====] - 128s 427ms/step - loss: 0.3709 - acc: 0.9377 - val_loss: 0.5990 - val_acc: 0.7575

Epoch 00094: val_acc did not improve from 0.79688

Epoch 95/100

300/300 [=====] - 128s 426ms/step - loss: 0.3709 - acc: 0.9367 - val_loss: 0.6093 - val_acc: 0.7525

Epoch 00095: val_acc did not improve from 0.79688

Epoch 96/100

300/300 [=====] - 128s 428ms/step - loss: 0.3767 - acc: 0.9340 - val_loss: 0.5994 - val_acc: 0.7575

Epoch 00096: val_acc did not improve from 0.79688

Epoch 97/100

300/300 [=====] - 128s 428ms/step - loss: 0.3719 - acc: 0.9348 - val_loss: 0.6103 - val_acc: 0.7550

Epoch 00097: val_acc did not improve from 0.79688

Epoch 00097: ReduceLROnPlateau reducing learning rate to 9.999999974752428e-08.

Epoch 98/100

300/300 [=====] - 128s 427ms/step - loss: 0.3693 - acc: 0.9427 - val_loss: 0.6068 - val_acc: 0.7544

Epoch 00098: val_acc did not improve from 0.79688

Epoch 99/100

300/300 [=====] - 128s 427ms/step - loss: 0.3610 - acc: 0.9417 - val_loss: 0.6356 - val_acc: 0.7362

Epoch 00099: val_acc did not improve from 0.79688

Epoch 100/100

300/300 [=====] - 129s 429ms/step - loss: 0.3630 - acc: 0.9413 - val_loss: 0.6205 - val_acc: 0.7469

Epoch 00100: val_acc did not improve from 0.79688

Out[0]:

<keras.callbacks.History at 0x7fb86d83de10>

At epoch 57 we can see Train accuracy is 87 and validation accuracy is approx 80 so our model was not overfit that time so considering epoch 57 for trained model

Epoch 57/100 => acc: 0.8700 - val_acc: 0.7969 --> saving model to ./model1.h5

In [0]:

In [0]:

!ls

```
drive      sample_submission.csv test.py
image      setup.py      test.zip
keras_vggface  test      tools
keras_vggface  test-faces.zip  train
LICENSE.txt  test-private-faces.zip  train-faces.zip
MANIFEST.in  test-private-labels.zip  train_relationships.csv
modell1.h5    test-private-lists.zip  train.zip
README.md    test-public-faces.zip  visualize
sample_data  test-public-lists.zip
```

In [0]:

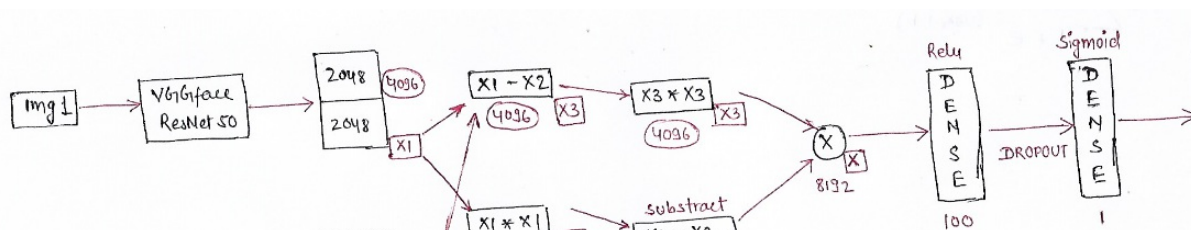
!pwd

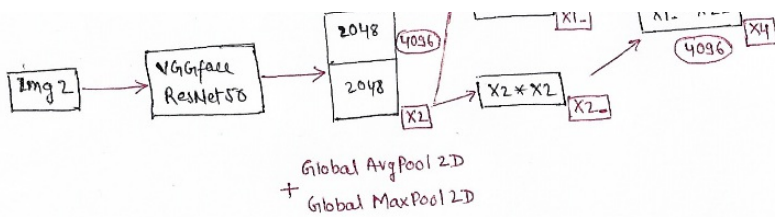
/content

In [0]:

!mv modell1.h5 ./drive/My\ Drive/COLAB_FILES/

MODEL 2





In [0]:

```
def gen(list_tuples, person_to_images_map, batch_size=16):
    ppl = list(person_to_images_map.keys())
    while True:
        batch_tuples = sample(list_tuples, batch_size // 2)
        labels = [1] * len(batch_tuples)
        while len(batch_tuples) < batch_size:
            p1 = choice(ppl)
            p2 = choice(ppl)

            if p1 != p2 and (p1, p2) not in list_tuples and (p2, p1) not in list_tuples:
                batch_tuples.append((p1, p2))
                labels.append(0)

        for x in batch_tuples:
            if not len(person_to_images_map[x[0]]):
                print(x[0])

        X1 = [choice(person_to_images_map[x[0]]) for x in batch_tuples]
        X1 = np.array([read_img(x, 197) for x in X1])

        X2 = [choice(person_to_images_map[x[1]]) for x in batch_tuples]
        X2 = np.array([read_img(x, 197) for x in X2])

        yield [X1, X2], labels
```

In [0]:

```
def baseline_model2():
    input_1 = Input(shape=(197, 197, 3))
    input_2 = Input(shape=(197, 197, 3))

    base_model2 = VGGFace(model='resnet50', include_top=False)

    for x in base_model2.layers[:-3]:
        x.trainable = True

    x1 = base_model2(input_1)
    x2 = base_model2(input_2)

    x1 = Concatenate(axis=-1) ([GlobalMaxPool2D()(x1), GlobalAvgPool2D()(x1)])
    x2 = Concatenate(axis=-1) ([GlobalMaxPool2D()(x2), GlobalAvgPool2D()(x2)])

    x3 = Subtract() ([x1, x2])
    x3 = Multiply() ([x3, x3])

    x1_ = Multiply() ([x1, x1])
    x2_ = Multiply() ([x2, x2])
    x4 = Subtract() ([x1_, x2_])
    x = Concatenate(axis=-1) ([x4, x3])
    x = Dropout(0.3)(x)
    x = Dense(100, activation="relu", activity_regularizer=l1(0.001))(x)
    x = Dropout(0.1)(x)
    out = Dense(1, activation="sigmoid")(x)

    model2 = Model([input_1, input_2], out)

    model2.compile(loss="binary_crossentropy", metrics=['acc', 'auc'], optimizer=Adam(0.00001))

    model2.summary()

    return model2
```

In [0]:

```
file_path2 = "./model2.h5"

checkpoint2 = ModelCheckpoint(file_path2, monitor='val_acc', verbose=1, save_best_only=True, mode='max'
)

reduce_on_plateau2 = ReduceLROnPlateau(monitor="val_acc", mode="max", factor=0.1, patience=20, verbose=
1)

callbacks_list2 = [checkpoint2, reduce_on_plateau2]

model2 = baseline_model2()
#model.load_weights(file_path)
```

Model: "model_4"

Layer (type)	Output Shape	Param #	Connected to
input_10 (InputLayer)	(None, 197, 197, 3)	0	
input_11 (InputLayer)	(None, 197, 197, 3)	0	
vggface_resnet50 (Model)	multiple	23561152	input_10[0][0] input_11[0][0]
global_max_pooling2d_7 (GlobalM	(None, 2048)	0	vggface_resnet50[1][0]
global_average_pooling2d_7 (Glo	(None, 2048)	0	vggface_resnet50[1][0]
global_max_pooling2d_8 (GlobalM	(None, 2048)	0	vggface_resnet50[2][0]
global_average_pooling2d_8 (Glo	(None, 2048)	0	vggface_resnet50[2][0]
concatenate_10 (Concatenate)	(None, 4096)	0	global_max_pooling2d_7[0][0] global_average_pooling2d_7[0][0]
concatenate_11 (Concatenate)	(None, 4096)	0	global_max_pooling2d_8[0][0] global_average_pooling2d_8[0][0]
multiply_11 (Multiply)	(None, 4096)	0	concatenate_10[0][0] concatenate_10[0][0]
multiply_12 (Multiply)	(None, 4096)	0	concatenate_11[0][0] concatenate_11[0][0]
subtract_7 (Subtract)	(None, 4096)	0	concatenate_10[0][0] concatenate_11[0][0]
subtract_8 (Subtract)	(None, 4096)	0	multiply_11[0][0] multiply_12[0][0]
multiply_10 (Multiply)	(None, 4096)	0	subtract_7[0][0] subtract_7[0][0]
concatenate_12 (Concatenate)	(None, 8192)	0	subtract_8[0][0] multiply_10[0][0]
dropout_7 (Dropout)	(None, 8192)	0	concatenate_12[0][0]
dense_8 (Dense)	(None, 100)	819300	dropout_7[0][0]
dropout_8 (Dropout)	(None, 100)	0	dense_8[0][0]
dense_9 (Dense)	(None, 1)	101	dropout_8[0][0]

Total params: 24,380,553

Trainable params: 24,327,433

Non-trainable params: 53,120

In [0]:

```
model2.fit_generator(gen(train, train_person_to_images_map, batch_size=16), use_multiprocessing=True,
validation_data=gen(val, val_person_to_images_map, batch_size=16), epochs=100, work
```

```
validation_data=gen(val, val_person_to_images_map, batch_size=16), epochs=100, verbose=2,  
workers=4, callbacks=callbacks_list2, steps_per_epoch=300, validation_steps=100)
```

Epoch 1/100

- 131s - loss: 0.5935 - acc: 0.7660 - auc: 0.8304 - val_loss: 0.6011 - val_acc: 0.7388 - val_auc: 0.8120

Epoch 00001: val_acc did not improve from 0.75750

Epoch 2/100

- 127s - loss: 0.5863 - acc: 0.7702 - auc: 0.8344 - val_loss: 0.5661 - val_acc: 0.7969 - val_auc: 0.8619

Epoch 00002: val_acc improved from 0.75750 to 0.79688, saving model to ./model2.h5

Epoch 3/100

- 128s - loss: 0.5790 - acc: 0.7765 - auc: 0.8443 - val_loss: 0.6017 - val_acc: 0.7425 - val_auc: 0.8169

Epoch 00003: val_acc did not improve from 0.79688

Epoch 4/100

- 127s - loss: 0.5678 - acc: 0.7865 - auc: 0.8539 - val_loss: 0.5902 - val_acc: 0.7625 - val_auc: 0.8267

Epoch 00004: val_acc did not improve from 0.79688

Epoch 5/100

- 128s - loss: 0.5635 - acc: 0.7831 - auc: 0.8561 - val_loss: 0.5648 - val_acc: 0.7750 - val_auc: 0.8541

Epoch 00005: val_acc did not improve from 0.79688

Epoch 6/100

- 128s - loss: 0.5681 - acc: 0.7850 - auc: 0.8531 - val_loss: 0.5731 - val_acc: 0.7825 - val_auc: 0.8464

Epoch 00006: val_acc did not improve from 0.79688

Epoch 7/100

- 128s - loss: 0.5463 - acc: 0.7937 - auc: 0.8680 - val_loss: 0.5827 - val_acc: 0.7625 - val_auc: 0.8336

Epoch 00007: val_acc did not improve from 0.79688

Epoch 8/100

- 128s - loss: 0.5502 - acc: 0.7929 - auc: 0.8708 - val_loss: 0.5677 - val_acc: 0.7775 - val_auc: 0.8494

Epoch 00008: val_acc did not improve from 0.79688

Epoch 9/100

- 128s - loss: 0.5383 - acc: 0.8048 - auc: 0.8799 - val_loss: 0.5770 - val_acc: 0.7700 - val_auc: 0.8402

Epoch 00009: val_acc did not improve from 0.79688

Epoch 10/100

- 128s - loss: 0.5397 - acc: 0.8079 - auc: 0.8742 - val_loss: 0.5908 - val_acc: 0.7494 - val_auc: 0.8234

Epoch 00010: val_acc did not improve from 0.79688

Epoch 11/100

- 128s - loss: 0.5333 - acc: 0.8058 - auc: 0.8818 - val_loss: 0.5615 - val_acc: 0.7712 - val_auc: 0.8525

Epoch 00011: val_acc did not improve from 0.79688

Epoch 12/100

- 128s - loss: 0.5182 - acc: 0.8204 - auc: 0.8970 - val_loss: 0.5701 - val_acc: 0.7762 - val_auc: 0.8397

Epoch 00012: val_acc did not improve from 0.79688

Epoch 13/100

- 128s - loss: 0.5251 - acc: 0.8181 - auc: 0.8863 - val_loss: 0.5816 - val_acc: 0.7688 - val_auc: 0.8353

Epoch 00013: val_acc did not improve from 0.79688

Epoch 14/100

- 128s - loss: 0.5172 - acc: 0.8269 - auc: 0.8915 - val_loss: 0.5458 - val_acc: 0.7919 - val_auc: 0.8738

Epoch 00014: val_acc did not improve from 0.79688

Epoch 15/100

- 128s - loss: 0.5041 - acc: 0.8329 - auc: 0.9060 - val_loss: 0.5787 - val_acc: 0.7550 - val_auc: 0.8286

Epoch 00015: val_acc did not improve from 0.79688
Epoch 16/100
- 128s - loss: 0.5076 - acc: 0.8252 - auc: 0.8968 - val_loss: 0.5635 - val_acc: 0.7750 - val_auc: 0.8577

Epoch 00016: val_acc did not improve from 0.79688
Epoch 17/100
- 128s - loss: 0.5139 - acc: 0.8229 - auc: 0.8924 - val_loss: 0.5559 - val_acc: 0.7775 - val_auc: 0.8480

Epoch 00017: val_acc did not improve from 0.79688
Epoch 18/100
- 128s - loss: 0.5077 - acc: 0.8304 - auc: 0.8991 - val_loss: 0.5657 - val_acc: 0.7762 - val_auc: 0.8470

Epoch 00018: val_acc did not improve from 0.79688
Epoch 19/100
- 128s - loss: 0.4957 - acc: 0.8402 - auc: 0.9071 - val_loss: 0.5454 - val_acc: 0.7819 - val_auc: 0.8548

Epoch 00019: val_acc did not improve from 0.79688
Epoch 20/100
- 128s - loss: 0.4887 - acc: 0.8415 - auc: 0.9151 - val_loss: 0.5543 - val_acc: 0.7800 - val_auc: 0.8600

Epoch 00020: val_acc did not improve from 0.79688
Epoch 21/100
- 127s - loss: 0.4811 - acc: 0.8508 - auc: 0.9170 - val_loss: 0.5523 - val_acc: 0.7863 - val_auc: 0.8594

Epoch 00021: val_acc did not improve from 0.79688
Epoch 22/100
- 127s - loss: 0.4799 - acc: 0.8515 - auc: 0.9171 - val_loss: 0.5669 - val_acc: 0.7744 - val_auc: 0.8444

Epoch 00022: val_acc did not improve from 0.79688
Epoch 00022: ReduceLROnPlateau reducing learning rate to 9.99999747378752e-07.
Epoch 23/100
- 128s - loss: 0.4779 - acc: 0.8483 - auc: 0.9149 - val_loss: 0.5610 - val_acc: 0.7812 - val_auc: 0.8509

Epoch 00023: val_acc did not improve from 0.79688
Epoch 24/100
- 128s - loss: 0.4679 - acc: 0.8590 - auc: 0.9250 - val_loss: 0.5595 - val_acc: 0.7794 - val_auc: 0.8547

Epoch 00024: val_acc did not improve from 0.79688
Epoch 25/100
- 128s - loss: 0.4695 - acc: 0.8546 - auc: 0.9229 - val_loss: 0.5638 - val_acc: 0.7762 - val_auc: 0.8550

Epoch 00025: val_acc did not improve from 0.79688
Epoch 26/100
- 128s - loss: 0.4639 - acc: 0.8612 - auc: 0.9263 - val_loss: 0.5393 - val_acc: 0.8019 - val_auc: 0.8703

Epoch 00026: val_acc improved from 0.79688 to 0.80188, saving model to ./model2.h5
Epoch 27/100
- 128s - loss: 0.4583 - acc: 0.8633 - auc: 0.9292 - val_loss: 0.5679 - val_acc: 0.7769 - val_auc: 0.8503

Epoch 00027: val_acc did not improve from 0.80188
Epoch 28/100
- 128s - loss: 0.4534 - acc: 0.8696 - auc: 0.9351 - val_loss: 0.5590 - val_acc: 0.7850 - val_auc: 0.8558

Epoch 00028: val_acc did not improve from 0.80188
Epoch 29/100
- 128s - loss: 0.4571 - acc: 0.8610 - auc: 0.9319 - val_loss: 0.5296 - val_acc: 0.8031 - val_auc: 0.8777

Epoch 00029: val_acc improved from 0.80188 to 0.80312, saving model to ./model2.h5
Epoch 30/100
- 128s - loss: 0.4555 - acc: 0.8698 - auc: 0.9312 - val_loss: 0.5384 - val_acc: 0.8056 - val_auc: 0.8742

Epoch 00030: val_acc improved from 0.80312 to 0.80563, saving model to ./model2.h5

Epoch 31/100

- 128s - loss: 0.4538 - acc: 0.8675 - auc: 0.9333 - val_loss: 0.5303 - val_acc: 0.7994 - val_auc: 0.8769

Epoch 00031: val_acc did not improve from 0.80563

Epoch 32/100

- 128s - loss: 0.4534 - acc: 0.8706 - auc: 0.9331 - val_loss: 0.5415 - val_acc: 0.7925 - val_auc: 0.8723

Epoch 00032: val_acc did not improve from 0.80563

Epoch 33/100

- 128s - loss: 0.4434 - acc: 0.8813 - auc: 0.9397 - val_loss: 0.5376 - val_acc: 0.7963 - val_auc: 0.8738

Epoch 00033: val_acc did not improve from 0.80563

Epoch 34/100

- 128s - loss: 0.4462 - acc: 0.8771 - auc: 0.9389 - val_loss: 0.5681 - val_acc: 0.7756 - val_auc: 0.8570

Epoch 00034: val_acc did not improve from 0.80563

Epoch 35/100

- 128s - loss: 0.4433 - acc: 0.8794 - auc: 0.9388 - val_loss: 0.5457 - val_acc: 0.7931 - val_auc: 0.8666

Epoch 00035: val_acc did not improve from 0.80563

Epoch 36/100

- 128s - loss: 0.4452 - acc: 0.8723 - auc: 0.9400 - val_loss: 0.5154 - val_acc: 0.8175 - val_auc: 0.8908

Epoch 00036: val_acc improved from 0.80563 to 0.81750, saving model to ./model2.h5

Epoch 37/100

- 128s - loss: 0.4395 - acc: 0.8860 - auc: 0.9417 - val_loss: 0.5590 - val_acc: 0.7806 - val_auc: 0.8600

Epoch 00037: val_acc did not improve from 0.81750

Epoch 38/100

- 128s - loss: 0.4410 - acc: 0.8758 - auc: 0.9421 - val_loss: 0.5590 - val_acc: 0.7781 - val_auc: 0.8555

Epoch 00038: val_acc did not improve from 0.81750

Epoch 39/100

- 128s - loss: 0.4498 - acc: 0.8673 - auc: 0.9369 - val_loss: 0.5588 - val_acc: 0.7812 - val_auc: 0.8617

Epoch 00039: val_acc did not improve from 0.81750

Epoch 40/100

- 128s - loss: 0.4413 - acc: 0.8792 - auc: 0.9404 - val_loss: 0.5627 - val_acc: 0.7656 - val_auc: 0.8556

Epoch 00040: val_acc did not improve from 0.81750

Epoch 41/100

- 128s - loss: 0.4442 - acc: 0.8765 - auc: 0.9399 - val_loss: 0.5553 - val_acc: 0.7831 - val_auc: 0.8586

Epoch 00041: val_acc did not improve from 0.81750

Epoch 42/100

- 128s - loss: 0.4325 - acc: 0.8802 - auc: 0.9485 - val_loss: 0.5674 - val_acc: 0.7731 - val_auc: 0.8548

Epoch 00042: val_acc did not improve from 0.81750

Epoch 43/100

- 128s - loss: 0.4365 - acc: 0.8825 - auc: 0.9454 - val_loss: 0.5504 - val_acc: 0.7769 - val_auc: 0.8656

Epoch 00043: val_acc did not improve from 0.81750

Epoch 44/100

- 128s - loss: 0.4298 - acc: 0.8883 - auc: 0.9468 - val_loss: 0.5667 - val_acc: 0.7762 - val_auc: 0.8628

Epoch 00044: val_acc did not improve from 0.81750

Epoch 45/100

- 128s - loss: 0.4373 - acc: 0.8798 - auc: 0.9416 - val_loss: 0.5355 - val_acc: 0.8044 - val_auc: 0.8769

Epoch 00045: val_acc did not improve from 0.81750

Epoch 00045: val_acc did not improve from 0.81750
Epoch 46/100
- 128s - loss: 0.4371 - acc: 0.8808 - auc: 0.9424 - val_loss: 0.5636 - val_acc: 0.7775 - val_auc: 0.8641

Epoch 00046: val_acc did not improve from 0.81750
Epoch 47/100
- 128s - loss: 0.4344 - acc: 0.8823 - auc: 0.9432 - val_loss: 0.5785 - val_acc: 0.7562 - val_auc: 0.8416

Epoch 00047: val_acc did not improve from 0.81750
Epoch 48/100
- 128s - loss: 0.4270 - acc: 0.8915 - auc: 0.9486 - val_loss: 0.5528 - val_acc: 0.7825 - val_auc: 0.8636

Epoch 00048: val_acc did not improve from 0.81750
Epoch 49/100
- 128s - loss: 0.4401 - acc: 0.8767 - auc: 0.9394 - val_loss: 0.5397 - val_acc: 0.7944 - val_auc: 0.8714

Epoch 00049: val_acc did not improve from 0.81750
Epoch 50/100
- 128s - loss: 0.4367 - acc: 0.8804 - auc: 0.9443 - val_loss: 0.5570 - val_acc: 0.7806 - val_auc: 0.8633

Epoch 00050: val_acc did not improve from 0.81750
Epoch 51/100
- 128s - loss: 0.4272 - acc: 0.8850 - auc: 0.9469 - val_loss: 0.5656 - val_acc: 0.7750 - val_auc: 0.8505

Epoch 00051: val_acc did not improve from 0.81750
Epoch 52/100
- 128s - loss: 0.4259 - acc: 0.8883 - auc: 0.9466 - val_loss: 0.5543 - val_acc: 0.7806 - val_auc: 0.8628

Epoch 00052: val_acc did not improve from 0.81750
Epoch 53/100
- 128s - loss: 0.4374 - acc: 0.8817 - auc: 0.9431 - val_loss: 0.5646 - val_acc: 0.7719 - val_auc: 0.8606

Epoch 00053: val_acc did not improve from 0.81750
Epoch 54/100
- 128s - loss: 0.4363 - acc: 0.8792 - auc: 0.9426 - val_loss: 0.5645 - val_acc: 0.7831 - val_auc: 0.8538

Epoch 00054: val_acc did not improve from 0.81750
Epoch 55/100
- 128s - loss: 0.4275 - acc: 0.8840 - auc: 0.9452 - val_loss: 0.5621 - val_acc: 0.7750 - val_auc: 0.8611

Epoch 00055: val_acc did not improve from 0.81750
Epoch 56/100
- 128s - loss: 0.4262 - acc: 0.8915 - auc: 0.9491 - val_loss: 0.5851 - val_acc: 0.7662 - val_auc: 0.8473

Epoch 00056: val_acc did not improve from 0.81750
Epoch 57/100
Epoch 00056: ReduceLROnPlateau reducing learning rate to 9.999999974752428e-08.
- 128s - loss: 0.4261 - acc: 0.8917 - auc: 0.9456 - val_loss: 0.5582 - val_acc: 0.7831 - val_auc: 0.8583

Epoch 00057: val_acc did not improve from 0.81750
Epoch 58/100
- 128s - loss: 0.4240 - acc: 0.8867 - auc: 0.9497 - val_loss: 0.5648 - val_acc: 0.7906 - val_auc: 0.8598

Epoch 00058: val_acc did not improve from 0.81750
Epoch 59/100
- 128s - loss: 0.4344 - acc: 0.8829 - auc: 0.9419 - val_loss: 0.5409 - val_acc: 0.7969 - val_auc: 0.8745

Epoch 00059: val_acc did not improve from 0.81750
Epoch 60/100
- 128s - loss: 0.4274 - acc: 0.8844 - auc: 0.9485 - val_loss: 0.5798 - val_acc: 0.7681 - val_auc: 0.8569

Epoch 00060: val_acc did not improve from 0.81750

Epoch 00060: val_acc did not improve from 0.81750
Epoch 61/100
- 128s - loss: 0.4307 - acc: 0.8821 - auc: 0.9429 - val_loss: 0.5783 - val_acc: 0.7700 - val_auc: 0.8483

Epoch 00061: val_acc did not improve from 0.81750
Epoch 62/100
- 128s - loss: 0.4271 - acc: 0.8844 - auc: 0.9445 - val_loss: 0.5727 - val_acc: 0.7681 - val_auc: 0.8456

Epoch 00062: val_acc did not improve from 0.81750
Epoch 63/100
- 127s - loss: 0.4310 - acc: 0.8883 - auc: 0.9463 - val_loss: 0.5416 - val_acc: 0.8019 - val_auc: 0.8786

Epoch 00063: val_acc did not improve from 0.81750
Epoch 64/100
- 128s - loss: 0.4259 - acc: 0.8888 - auc: 0.9477 - val_loss: 0.5508 - val_acc: 0.7931 - val_auc: 0.8642

Epoch 00064: val_acc did not improve from 0.81750
Epoch 65/100
- 128s - loss: 0.4239 - acc: 0.8856 - auc: 0.9504 - val_loss: 0.5676 - val_acc: 0.7881 - val_auc: 0.8552

Epoch 00065: val_acc did not improve from 0.81750
Epoch 66/100
- 128s - loss: 0.4257 - acc: 0.8879 - auc: 0.9461 - val_loss: 0.5324 - val_acc: 0.7950 - val_auc: 0.8811

Epoch 00066: val_acc did not improve from 0.81750
Epoch 67/100
- 128s - loss: 0.4180 - acc: 0.8952 - auc: 0.9560 - val_loss: 0.5912 - val_acc: 0.7588 - val_auc: 0.8438

Epoch 00067: val_acc did not improve from 0.81750
Epoch 68/100
- 128s - loss: 0.4140 - acc: 0.8967 - auc: 0.9528 - val_loss: 0.5730 - val_acc: 0.7562 - val_auc: 0.8470

Epoch 00068: val_acc did not improve from 0.81750
Epoch 69/100
- 128s - loss: 0.4222 - acc: 0.8915 - auc: 0.9493 - val_loss: 0.5612 - val_acc: 0.7794 - val_auc: 0.8681

Epoch 00069: val_acc did not improve from 0.81750
Epoch 70/100
- 128s - loss: 0.4263 - acc: 0.8865 - auc: 0.9474 - val_loss: 0.5597 - val_acc: 0.7694 - val_auc: 0.8622

Epoch 00070: val_acc did not improve from 0.81750
Epoch 71/100
- 127s - loss: 0.4174 - acc: 0.8975 - auc: 0.9543 - val_loss: 0.5582 - val_acc: 0.7800 - val_auc: 0.8645

Epoch 00071: val_acc did not improve from 0.81750
Epoch 72/100
- 128s - loss: 0.4329 - acc: 0.8848 - auc: 0.9459 - val_loss: 0.5692 - val_acc: 0.7788 - val_auc: 0.8592

Epoch 00072: val_acc did not improve from 0.81750
Epoch 73/100
- 127s - loss: 0.4312 - acc: 0.8894 - auc: 0.9425 - val_loss: 0.5708 - val_acc: 0.7750 - val_auc: 0.8477

Epoch 00073: val_acc did not improve from 0.81750
Epoch 74/100
- 128s - loss: 0.4319 - acc: 0.8835 - auc: 0.9423 - val_loss: 0.5387 - val_acc: 0.8037 - val_auc: 0.8852

Epoch 00074: val_acc did not improve from 0.81750
Epoch 75/100
- 128s - loss: 0.4180 - acc: 0.8938 - auc: 0.9499 - val_loss: 0.5596 - val_acc: 0.7869 - val_auc: 0.8658

Epoch 00075: val_acc did not improve from 0.81750
Epoch 76/100
- 128s - loss: 0.4101 - acc: 0.8940 - auc: 0.9520 - val_loss: 0.5600 - val_acc: 0.7800 - val_auc: 0.87

- 128s - loss: 0.4191 - acc: 0.8948 - auc: 0.9526 - val_loss: 0.5600 - val_acc: 0.7800 - val_auc: 0.8703

Epoch 00076: val_acc did not improve from 0.81750

Epoch 00076: ReduceLROnPlateau reducing learning rate to 1.0000000116860975e-08.

Epoch 77/100

- 128s - loss: 0.4243 - acc: 0.8919 - auc: 0.9497 - val_loss: 0.5819 - val_acc: 0.7569 - val_auc: 0.8427

Epoch 00077: val_acc did not improve from 0.81750

Epoch 78/100

- 128s - loss: 0.4219 - acc: 0.8881 - auc: 0.9498 - val_loss: 0.5611 - val_acc: 0.7662 - val_auc: 0.8641

Epoch 00078: val_acc did not improve from 0.81750

Epoch 79/100

- 127s - loss: 0.4185 - acc: 0.8904 - auc: 0.9533 - val_loss: 0.5641 - val_acc: 0.7744 - val_auc: 0.8609

Epoch 00079: val_acc did not improve from 0.81750

Epoch 80/100

- 128s - loss: 0.4230 - acc: 0.8867 - auc: 0.9501 - val_loss: 0.5682 - val_acc: 0.7762 - val_auc: 0.8656

Epoch 00080: val_acc did not improve from 0.81750

Epoch 81/100

- 128s - loss: 0.4331 - acc: 0.8835 - auc: 0.9420 - val_loss: 0.5668 - val_acc: 0.7738 - val_auc: 0.8519

Epoch 00081: val_acc did not improve from 0.81750

Epoch 82/100

- 128s - loss: 0.4207 - acc: 0.8942 - auc: 0.9494 - val_loss: 0.5576 - val_acc: 0.7837 - val_auc: 0.8661

Epoch 00082: val_acc did not improve from 0.81750

Epoch 83/100

- 127s - loss: 0.4220 - acc: 0.8910 - auc: 0.9513 - val_loss: 0.5615 - val_acc: 0.7750 - val_auc: 0.8569

Epoch 00083: val_acc did not improve from 0.81750

Epoch 84/100

- 128s - loss: 0.4176 - acc: 0.8948 - auc: 0.9513 - val_loss: 0.5678 - val_acc: 0.7594 - val_auc: 0.8605

Epoch 00084: val_acc did not improve from 0.81750

Epoch 85/100

- 127s - loss: 0.4196 - acc: 0.8923 - auc: 0.9497 - val_loss: 0.5681 - val_acc: 0.7662 - val_auc: 0.8531

Epoch 00085: val_acc did not improve from 0.81750

Epoch 86/100

- 127s - loss: 0.4268 - acc: 0.8894 - auc: 0.9470 - val_loss: 0.5839 - val_acc: 0.7669 - val_auc: 0.8414

Epoch 00086: val_acc did not improve from 0.81750

Epoch 87/100

- 127s - loss: 0.4263 - acc: 0.8904 - auc: 0.9483 - val_loss: 0.5694 - val_acc: 0.7731 - val_auc: 0.8623

Epoch 00087: val_acc did not improve from 0.81750

Epoch 88/100

- 127s - loss: 0.4236 - acc: 0.8890 - auc: 0.9483 - val_loss: 0.5586 - val_acc: 0.7881 - val_auc: 0.8628

Epoch 00088: val_acc did not improve from 0.81750

Epoch 89/100

- 127s - loss: 0.4260 - acc: 0.8896 - auc: 0.9476 - val_loss: 0.5649 - val_acc: 0.7762 - val_auc: 0.8530

Epoch 00089: val_acc did not improve from 0.81750

Epoch 90/100

- 127s - loss: 0.4279 - acc: 0.8856 - auc: 0.9436 - val_loss: 0.5719 - val_acc: 0.7694 - val_auc: 0.8595

Epoch 00090: val_acc did not improve from 0.81750

Epoch 91/100

- 128s - loss: 0.4188 - acc: 0.8875 - auc: 0.9511 - val_loss: 0.5727 - val_acc: 0.7805 - val_auc: 0.86


```

- 128s - loss: 0.4189 - acc: 0.8975 - auc: 0.9511 - val_loss: 0.5727 - val_acc: 0.7825 - val_auc: 0.86
23

Epoch 00091: val_acc did not improve from 0.81750
Epoch 92/100
- 128s - loss: 0.4243 - acc: 0.8929 - auc: 0.9481 - val_loss: 0.5695 - val_acc: 0.7725 - val_auc: 0.85
34

Epoch 00092: val_acc did not improve from 0.81750
Epoch 93/100
- 127s - loss: 0.4200 - acc: 0.8921 - auc: 0.9485 - val_loss: 0.5695 - val_acc: 0.7725 - val_auc: 0.85
66

Epoch 00093: val_acc did not improve from 0.81750
Epoch 94/100
- 127s - loss: 0.4186 - acc: 0.8950 - auc: 0.9498 - val_loss: 0.5583 - val_acc: 0.7775 - val_auc: 0.86
56

Epoch 00094: val_acc did not improve from 0.81750
Epoch 95/100
- 128s - loss: 0.4164 - acc: 0.8983 - auc: 0.9526 - val_loss: 0.5486 - val_acc: 0.7825 - val_auc: 0.87
55

Epoch 00095: val_acc did not improve from 0.81750
Epoch 96/100
- 128s - loss: 0.4275 - acc: 0.8906 - auc: 0.9456 - val_loss: 0.5781 - val_acc: 0.7675 - val_auc: 0.84
52

Epoch 00096: val_acc did not improve from 0.81750

Epoch 00096: ReduceLROnPlateau reducing learning rate to 9.999999939225292e-10.
Epoch 97/100
- 127s - loss: 0.4313 - acc: 0.8896 - auc: 0.9435 - val_loss: 0.5556 - val_acc: 0.7738 - val_auc: 0.86
41

Epoch 00097: val_acc did not improve from 0.81750
Epoch 98/100
- 128s - loss: 0.4251 - acc: 0.8917 - auc: 0.9481 - val_loss: 0.5690 - val_acc: 0.7769 - val_auc: 0.85
42

Epoch 00098: val_acc did not improve from 0.81750
Epoch 99/100
- 127s - loss: 0.4215 - acc: 0.8944 - auc: 0.9489 - val_loss: 0.5546 - val_acc: 0.7700 - val_auc: 0.86
50

Epoch 00099: val_acc did not improve from 0.81750
Epoch 100/100
- 128s - loss: 0.4252 - acc: 0.8910 - auc: 0.9471 - val_loss: 0.5749 - val_acc: 0.7844 - val_auc: 0.85
08

Epoch 00100: val_acc did not improve from 0.81750

```

Out[0]:

```
<keras.callbacks.History at 0x7fb6193dc518>
```

At epoch 36 we can see Train accuracy is 87 and validation accuracy is approx 82 so considering epoch 36 for trained model

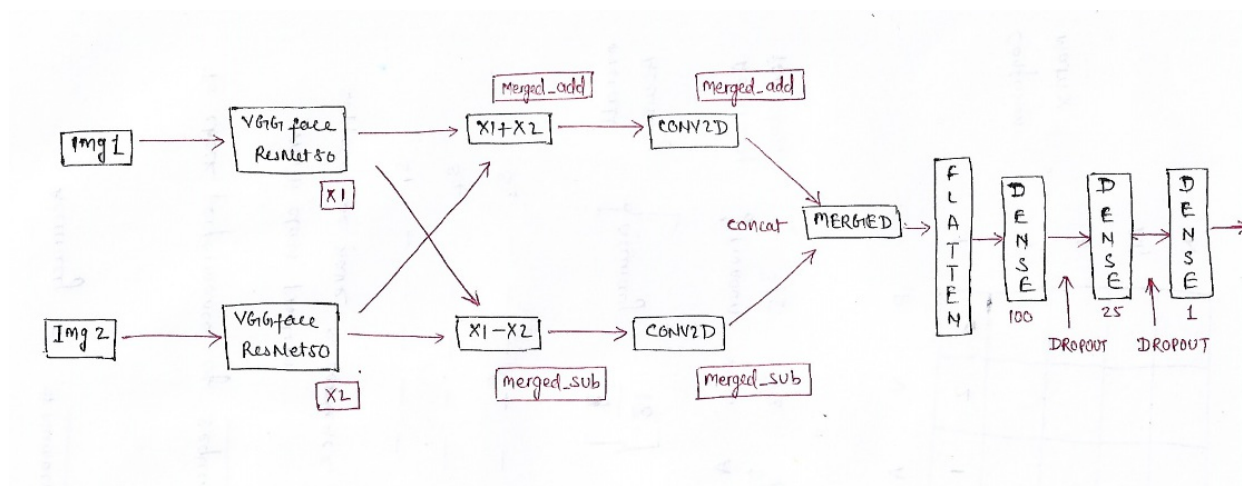
Epoch 36/100 => acc: 0.8723 - val_acc: 0.8175 --> saving model to ./model2.h5

In [0]:

```
!mv model2.h5 ./drive/My\ Drive/COLAB_FILES/
```

In [0]:

MODEL 3



In [0]:

```
def read_img_cv(path):
    img = cv2.imread(path)
    img = np.array(img).astype(np.float)
    return preprocess_input(img, version=2)

def gen(list_tuples, person_to_images_map, batch_size=16):
    ppl = list(person_to_images_map.keys())
    while True:
        batch_tuples = sample(list_tuples, batch_size // 2)
        labels = [1] * len(batch_tuples)
        while len(batch_tuples) < batch_size:
            p1 = choice(ppl)
            p2 = choice(ppl)

            if p1 != p2 and (p1, p2) not in list_tuples and (p2, p1) not in list_tuples:
                batch_tuples.append((p1, p2))
                labels.append(0)

        for x in batch_tuples:
            if not len(person_to_images_map[x[0]]):
                print(x[0])

            X1 = [choice(person_to_images_map[x[0]]) for x in batch_tuples]
            X1 = np.array([read_img_cv(x) for x in X1])

            X2 = [choice(person_to_images_map[x[1]]) for x in batch_tuples]
            X2 = np.array([read_img_cv(x) for x in X2])

        yield [X1, X2], labels
```

In [0]:

In [0]:

```
def baseline_model3_1():
    input_1 = Input(shape=(224, 224, 3))
    input_2 = Input(shape=(224, 224, 3))

    base_model3 = VGGFace(model='resnet50', include_top=False)

    for layer in base_model3.layers[:-3]:
        layer.trainable = True

    x1 = base_model3(input_1)
    x2 = base_model3(input_2)

    merged_add = Add()([x1, x2])
    merged_sub = Subtract()([x1, x2])
```

```

merged_add = Conv2D(100 , [1,1] )(merged_add)
merged_sub = Conv2D(100 , [1,1] )(merged_sub)

merged = Concatenate(axis=-1) ([merged_add, merged_sub])

merged = Flatten() (merged)
merged = Dropout(0.2) (merged)
merged = Dense(100, activation="relu", activity_regularizer=l1(0.001)) (merged)
merged = Dropout(0.2) (merged)
merged = Dense(25, activation="relu", activity_regularizer=l1(0.001)) (merged)
merged = Dropout(0.2) (merged)
out = Dense(1, activation="sigmoid") (merged)
model3 = Model([input_1, input_2], out)
model3.compile(loss="binary_crossentropy", metrics=['acc', auc], optimizer=Adam(0.00001))
model3.summary()
return model3

```

In [0]:

```

file_path3 = "./model3_1.h5"

checkpoint3 = ModelCheckpoint(file_path3, monitor='val_acc', verbose=1, save_best_only=True, mode='max'
)

reduce_on_plateau3 = ReduceLROnPlateau(monitor="val_acc", mode="max", factor=0.1, patience=20, verbose=
1)

callbacks_list3 = [checkpoint3, reduce_on_plateau3]

model3_1 = baseline_model3_1()

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:66:
The name tf.get_default_graph is deprecated. Please use tf.compat.v1.get_default_graph instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:541:
The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4432:
: The name tf.random_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:190:
The name tf.get_default_session is deprecated. Please use tf.compat.v1.get_default_session instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:197:
The name tf.ConfigProto is deprecated. Please use tf.compat.v1.ConfigProto instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:203:
The name tf.Session is deprecated. Please use tf.compat.v1.Session instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:207:
The name tf.global_variables is deprecated. Please use tf.compat.v1.global_variables instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:216:
The name tf.is_variable_initialized is deprecated. Please use tf.compat.v1.is_variable_initialized instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:223:
The name tf.variables_initializer is deprecated. Please use tf.compat.v1.variables_initializer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:2041:
: The name tf.nn.fused_batch_norm is deprecated. Please use tf.compat.v1.nn.fused_batch_norm instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:148:
The name tf.placeholder_with_default is deprecated. Please use tf.compat.v1.placeholder_with_default instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4267:
: The name tf.nn.max_pool is deprecated. Please use tf.nn.max_pool2d instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4271:
: The name tf.nn.avg_pool is deprecated. Please use tf.nn.avg_pool2d instead.

Downloading data from https://github.com/rcmalli/keras-vggface/releases/download/v2.0/rcmalli_vggface_tf_notop_resnet50.h5
94699520/94694792 [=====] - 1s 0us/step

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3733 : calling dropout (from tensorflow.python.ops.nn_ops) with keep_prob is deprecated and will be removed in a future version.
Instructions for updating:
Please use `rate` instead of `keep_prob`. Rate should be set to `rate = 1 - keep_prob`.
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3657 : The name tf.log is deprecated. Please use tf.math.log instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow_core/python/ops/nn_impl.py:183: where (from tensorflow.python.ops.array_ops) is deprecated and will be removed in a future version.
Instructions for updating:
Use tf.where in 2.0, which has the same broadcast rule as np.where
WARNING:tensorflow:From <ipython-input-56-fb283a009ca9>:2: py_func (from tensorflow.python.ops.script_ops) is deprecated and will be removed in a future version.
Instructions for updating:

tf.py_func is deprecated in TF V2. Instead, there are two options available in V2.
- tf.py_function takes a python function which manipulates tf eager tensors instead of numpy arrays. It's easy to convert a tf eager tensor to an ndarray (just call tensor.numpy()) but having access to eager tensors means `tf.py_function`s can use accelerators such as GPUs as well as being differentiable using a gradient tape.
- tf.numpy_function maintains the semantics of the deprecated tf.py_func (it is not differentiable, and manipulates numpy arrays). It drops the stateful argument making all functions stateful.

Model: "model_1"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 224, 224, 3)	0	
input_2 (InputLayer)	(None, 224, 224, 3)	0	
vggface_resnet50 (Model)	multiple	23561152	input_1[0][0] input_2[0][0]
add_17 (Add)	(None, 1, 1, 2048)	0	vggface_resnet50[1][0] vggface_resnet50[2][0]
subtract_1 (Subtract)	(None, 1, 1, 2048)	0	vggface_resnet50[1][0] vggface_resnet50[2][0]
conv2d_1 (Conv2D)	(None, 1, 1, 100)	204900	add_17[0][0]
conv2d_2 (Conv2D)	(None, 1, 1, 100)	204900	subtract_1[0][0]
concatenate_1 (Concatenate)	(None, 1, 1, 200)	0	conv2d_1[0][0] conv2d_2[0][0]
flatten_1 (Flatten)	(None, 200)	0	concatenate_1[0][0]
dropout_1 (Dropout)	(None, 200)	0	flatten_1[0][0]
dense_1 (Dense)	(None, 100)	20100	dropout_1[0][0]
dropout_2 (Dropout)	(None, 100)	0	dense_1[0][0]
dense_2 (Dense)	(None, 25)	2525	dropout_2[0][0]
dropout_3 (Dropout)	(None, 25)	0	dense_2[0][0]
dense_3 (Dense)	(None, 1)	26	dropout_3[0][0]

Total params: 23,993,603
Trainable params: 23,940,483
Non-trainable params: 53,120

In [0]:

```
model3_1.fit_generator(gen(train, train_person_to_images_map, batch_size=16), use_multiprocessing=True,
                      validation_data=gen(val, val_person_to_images_map, batch_size=16), epochs=130, verb
```

```
ose=1,  
workers = 4, callbacks=callbacks_list3, steps_per_epoch=300, validation_steps=100)
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:1033
: The name tf.assign_add is deprecated. Please use tf.compat.v1.assign_add instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:1020
: The name tf.assign is deprecated. Please use tf.compat.v1.assign instead.

Epoch 1/130

300/300 [=====] - 167s 557ms/step - loss: 2.7493 - acc: 0.5581 - auc: 0.5769 -
val_loss: 0.9779 - val_acc: 0.5856 - val_auc: 0.6389

Epoch 00001: val_acc improved from -inf to 0.58562, saving model to ./model3_1.h5

Epoch 2/130

300/300 [=====] - 141s 471ms/step - loss: 0.9386 - acc: 0.5667 - auc: 0.6057 -
val_loss: 0.7455 - val_acc: 0.6406 - val_auc: 0.6764

Epoch 00002: val_acc improved from 0.58562 to 0.64062, saving model to ./model3_1.h5

Epoch 3/130

300/300 [=====] - 141s 471ms/step - loss: 0.8034 - acc: 0.5658 - auc: 0.6184 -
val_loss: 0.7196 - val_acc: 0.6388 - val_auc: 0.6834

Epoch 00003: val_acc did not improve from 0.64062

Epoch 4/130

300/300 [=====] - 141s 470ms/step - loss: 0.7548 - acc: 0.6019 - auc: 0.6596 -
val_loss: 0.7005 - val_acc: 0.6331 - val_auc: 0.6786

Epoch 00004: val_acc did not improve from 0.64062

Epoch 5/130

300/300 [=====] - 140s 468ms/step - loss: 0.7354 - acc: 0.5975 - auc: 0.6608 -
val_loss: 0.6916 - val_acc: 0.6569 - val_auc: 0.7083

Epoch 00005: val_acc improved from 0.64062 to 0.65687, saving model to ./model3_1.h5

Epoch 6/130

300/300 [=====] - 141s 469ms/step - loss: 0.7224 - acc: 0.6042 - auc: 0.6709 -
val_loss: 0.6980 - val_acc: 0.6319 - val_auc: 0.6741

Epoch 00006: val_acc did not improve from 0.65687

Epoch 7/130

300/300 [=====] - 141s 469ms/step - loss: 0.7099 - acc: 0.6135 - auc: 0.6919 -
val_loss: 0.6787 - val_acc: 0.6694 - val_auc: 0.7093

Epoch 00007: val_acc improved from 0.65687 to 0.66938, saving model to ./model3_1.h5

Epoch 8/130

300/300 [=====] - 141s 469ms/step - loss: 0.7008 - acc: 0.6417 - auc: 0.7135 -
val_loss: 0.6776 - val_acc: 0.6531 - val_auc: 0.7084

Epoch 00008: val_acc did not improve from 0.66938

Epoch 9/130

300/300 [=====] - 141s 469ms/step - loss: 0.6939 - acc: 0.6219 - auc: 0.6960 -
val_loss: 0.6759 - val_acc: 0.6400 - val_auc: 0.7149

Epoch 00009: val_acc did not improve from 0.66938

Epoch 10/130

300/300 [=====] - 141s 469ms/step - loss: 0.6892 - acc: 0.6315 - auc: 0.7015 -
val_loss: 0.6709 - val_acc: 0.6519 - val_auc: 0.7117

Epoch 00010: val_acc did not improve from 0.66938

Epoch 11/130

300/300 [=====] - 141s 469ms/step - loss: 0.6812 - acc: 0.6485 - auc: 0.7156 -
val_loss: 0.6600 - val_acc: 0.6875 - val_auc: 0.7231

Epoch 00011: val_acc improved from 0.66938 to 0.68750, saving model to ./model3_1.h5

Epoch 12/130

300/300 [=====] - 140s 468ms/step - loss: 0.6748 - acc: 0.6425 - auc: 0.7003 -
val_loss: 0.6660 - val_acc: 0.6438 - val_auc: 0.6925

Epoch 00012: val_acc did not improve from 0.68750

Epoch 13/130

300/300 [=====] - 140s 468ms/step - loss: 0.6744 - acc: 0.6633 - auc: 0.7197 -
val_loss: 0.6634 - val_acc: 0.6538 - val_auc: 0.6844

Epoch 00013: val_acc did not improve from 0.68750

Epoch 14/130

300/300 [=====] - 140s 468ms/step - loss: 0.6718 - acc: 0.6544 - auc: 0.7045 -

val_loss: 0.6577 - val_acc: 0.6787 - val_auc: 0.7026

Epoch 00014: val_acc did not improve from 0.68750

Epoch 15/130

300/300 [=====] - 142s 472ms/step - loss: 0.6691 - acc: 0.6627 - auc: 0.7093 -
val_loss: 0.6635 - val_acc: 0.6525 - val_auc: 0.7084

Epoch 00015: val_acc did not improve from 0.68750

Epoch 16/130

300/300 [=====] - 142s 473ms/step - loss: 0.6626 - acc: 0.6771 - auc: 0.7221 -
val_loss: 0.6535 - val_acc: 0.6800 - val_auc: 0.7254

Epoch 00016: val_acc did not improve from 0.68750

Epoch 17/130

300/300 [=====] - 142s 473ms/step - loss: 0.6596 - acc: 0.6869 - auc: 0.7290 -
val_loss: 0.6638 - val_acc: 0.6431 - val_auc: 0.6730

Epoch 00017: val_acc did not improve from 0.68750

Epoch 18/130

300/300 [=====] - 142s 472ms/step - loss: 0.6601 - acc: 0.6823 - auc: 0.7263 -
val_loss: 0.6744 - val_acc: 0.6106 - val_auc: 0.6707

Epoch 00018: val_acc did not improve from 0.68750

Epoch 19/130

300/300 [=====] - 141s 470ms/step - loss: 0.6552 - acc: 0.6873 - auc: 0.7282 -
val_loss: 0.6600 - val_acc: 0.6544 - val_auc: 0.6785

Epoch 00019: val_acc did not improve from 0.68750

Epoch 20/130

300/300 [=====] - 141s 470ms/step - loss: 0.6503 - acc: 0.6898 - auc: 0.7185 -
val_loss: 0.6676 - val_acc: 0.6481 - val_auc: 0.6751

Epoch 00020: val_acc did not improve from 0.68750

Epoch 21/130

300/300 [=====] - 140s 468ms/step - loss: 0.6525 - acc: 0.6927 - auc: 0.7267 -
val_loss: 0.6552 - val_acc: 0.6700 - val_auc: 0.7130

Epoch 00021: val_acc did not improve from 0.68750

Epoch 22/130

300/300 [=====] - 141s 468ms/step - loss: 0.6486 - acc: 0.6944 - auc: 0.7268 -
val_loss: 0.6538 - val_acc: 0.6694 - val_auc: 0.6868

Epoch 00022: val_acc did not improve from 0.68750

Epoch 23/130

300/300 [=====] - 141s 469ms/step - loss: 0.6467 - acc: 0.7008 - auc: 0.7310 -
val_loss: 0.6724 - val_acc: 0.6275 - val_auc: 0.6394

Epoch 00023: val_acc did not improve from 0.68750

Epoch 24/130

300/300 [=====] - 140s 468ms/step - loss: 0.6501 - acc: 0.6906 - auc: 0.7257 -
val_loss: 0.6504 - val_acc: 0.6719 - val_auc: 0.6993

Epoch 00024: val_acc did not improve from 0.68750

Epoch 25/130

300/300 [=====] - 141s 469ms/step - loss: 0.6472 - acc: 0.6992 - auc: 0.7279 -
val_loss: 0.6519 - val_acc: 0.6625 - val_auc: 0.6807

Epoch 00025: val_acc did not improve from 0.68750

Epoch 26/130

300/300 [=====] - 141s 469ms/step - loss: 0.6452 - acc: 0.6954 - auc: 0.7264 -
val_loss: 0.6636 - val_acc: 0.6338 - val_auc: 0.6507

Epoch 00026: val_acc did not improve from 0.68750

Epoch 27/130

300/300 [=====] - 141s 469ms/step - loss: 0.6423 - acc: 0.6973 - auc: 0.7324 -
val_loss: 0.6541 - val_acc: 0.6594 - val_auc: 0.6852

Epoch 00027: val_acc did not improve from 0.68750

Epoch 28/130

300/300 [=====] - 141s 469ms/step - loss: 0.6391 - acc: 0.7042 - auc: 0.7396 -
val_loss: 0.6793 - val_acc: 0.6519 - val_auc: 0.6957

Epoch 00028: val_acc did not improve from 0.68750

Epoch 29/130

300/300 [=====] - 140s 468ms/step - loss: 0.6374 - acc: 0.7125 - auc: 0.7433 -
val_loss: 0.6674 - val_acc: 0.6550 - val_auc: 0.6699

Epoch 00029: val_acc did not improve from 0.68750
Epoch 30/130
300/300 [=====] - 140s 468ms/step - loss: 0.6340 - acc: 0.7090 - auc: 0.7356 -
val_loss: 0.6648 - val_acc: 0.6538 - val_auc: 0.6917

Epoch 00030: val_acc did not improve from 0.68750
Epoch 31/130
300/300 [=====] - 140s 468ms/step - loss: 0.6315 - acc: 0.7150 - auc: 0.7396 -
val_loss: 0.6497 - val_acc: 0.6925 - val_auc: 0.7289

Epoch 00031: val_acc improved from 0.68750 to 0.69250, saving model to ./model3_1.h5
Epoch 32/130
300/300 [=====] - 140s 468ms/step - loss: 0.6299 - acc: 0.7188 - auc: 0.7429 -
val_loss: 0.6489 - val_acc: 0.6731 - val_auc: 0.7024

Epoch 00032: val_acc did not improve from 0.69250
Epoch 33/130
300/300 [=====] - 140s 468ms/step - loss: 0.6328 - acc: 0.7121 - auc: 0.7364 -
val_loss: 0.6576 - val_acc: 0.6562 - val_auc: 0.6737

Epoch 00033: val_acc did not improve from 0.69250
Epoch 34/130
300/300 [=====] - 140s 468ms/step - loss: 0.6303 - acc: 0.7198 - auc: 0.7461 -
val_loss: 0.6706 - val_acc: 0.6531 - val_auc: 0.6887

Epoch 00034: val_acc did not improve from 0.69250
Epoch 35/130
300/300 [=====] - 140s 468ms/step - loss: 0.6328 - acc: 0.7094 - auc: 0.7351 -
val_loss: 0.6602 - val_acc: 0.6456 - val_auc: 0.6690

Epoch 00035: val_acc did not improve from 0.69250
Epoch 36/130
300/300 [=====] - 140s 468ms/step - loss: 0.6208 - acc: 0.7262 - auc: 0.7446 -
val_loss: 0.6608 - val_acc: 0.6706 - val_auc: 0.7095

Epoch 00036: val_acc did not improve from 0.69250
Epoch 37/130
300/300 [=====] - 140s 468ms/step - loss: 0.6236 - acc: 0.7202 - auc: 0.7464 -
val_loss: 0.6351 - val_acc: 0.6937 - val_auc: 0.7191

Epoch 00037: val_acc improved from 0.69250 to 0.69375, saving model to ./model3_1.h5
Epoch 38/130
300/300 [=====] - 140s 467ms/step - loss: 0.6265 - acc: 0.7171 - auc: 0.7350 -
val_loss: 0.6488 - val_acc: 0.6587 - val_auc: 0.6861

Epoch 00038: val_acc did not improve from 0.69375
Epoch 39/130
300/300 [=====] - 141s 469ms/step - loss: 0.6228 - acc: 0.7160 - auc: 0.7367 -
val_loss: 0.6574 - val_acc: 0.6669 - val_auc: 0.7067

Epoch 00039: val_acc did not improve from 0.69375
Epoch 40/130
300/300 [=====] - 141s 468ms/step - loss: 0.6227 - acc: 0.7200 - auc: 0.7497 -
val_loss: 0.6380 - val_acc: 0.6887 - val_auc: 0.7133

Epoch 00040: val_acc did not improve from 0.69375
Epoch 41/130
300/300 [=====] - 140s 468ms/step - loss: 0.6197 - acc: 0.7246 - auc: 0.7476 -
val_loss: 0.6464 - val_acc: 0.6806 - val_auc: 0.7398

Epoch 00041: val_acc did not improve from 0.69375
Epoch 42/130
300/300 [=====] - 140s 467ms/step - loss: 0.6194 - acc: 0.7181 - auc: 0.7527 -
val_loss: 0.6380 - val_acc: 0.6894 - val_auc: 0.7257

Epoch 00042: val_acc did not improve from 0.69375
Epoch 43/130
300/300 [=====] - 140s 468ms/step - loss: 0.6172 - acc: 0.7294 - auc: 0.7488 -
val_loss: 0.6439 - val_acc: 0.6844 - val_auc: 0.7069

Epoch 00043: val_acc did not improve from 0.69375
Epoch 44/130
300/300 [=====] - 140s 468ms/step - loss: 0.6119 - acc: 0.7321 - auc: 0.7629 -
val_loss: 0.6438 - val_acc: 0.6619 - val_auc: 0.7079

Epoch 00044: val_acc did not improve from 0.69375
Epoch 45/130

300/300 [=====] - 140s 468ms/step - loss: 0.6142 - acc: 0.7346 - auc: 0.7579 - val_loss: 0.6334 - val_acc: 0.6994 - val_auc: 0.7334

Epoch 00045: val_acc improved from 0.69375 to 0.69937, saving model to ./model3_1.h5
Epoch 46/130
300/300 [=====] - 140s 467ms/step - loss: 0.6153 - acc: 0.7227 - auc: 0.7515 - val_loss: 0.6481 - val_acc: 0.6625 - val_auc: 0.6831

Epoch 00046: val_acc did not improve from 0.69937
Epoch 47/130
300/300 [=====] - 140s 466ms/step - loss: 0.6150 - acc: 0.7258 - auc: 0.7575 - val_loss: 0.6587 - val_acc: 0.6894 - val_auc: 0.7150

Epoch 00047: val_acc did not improve from 0.69937
Epoch 48/130
300/300 [=====] - 141s 469ms/step - loss: 0.6145 - acc: 0.7277 - auc: 0.7469 - val_loss: 0.6558 - val_acc: 0.6675 - val_auc: 0.6894

Epoch 00048: val_acc did not improve from 0.69937
Epoch 49/130
300/300 [=====] - 140s 467ms/step - loss: 0.6019 - acc: 0.7525 - auc: 0.7844 - val_loss: 0.6821 - val_acc: 0.6856 - val_auc: 0.7170

Epoch 00049: val_acc did not improve from 0.69937
Epoch 50/130
300/300 [=====] - 140s 467ms/step - loss: 0.6042 - acc: 0.7423 - auc: 0.7699 - val_loss: 0.6529 - val_acc: 0.6856 - val_auc: 0.7107

Epoch 00050: val_acc did not improve from 0.69937
Epoch 51/130
300/300 [=====] - 140s 467ms/step - loss: 0.6024 - acc: 0.7473 - auc: 0.7743 - val_loss: 0.6401 - val_acc: 0.6937 - val_auc: 0.7248

Epoch 00051: val_acc did not improve from 0.69937
Epoch 52/130
300/300 [=====] - 140s 467ms/step - loss: 0.5969 - acc: 0.7596 - auc: 0.7872 - val_loss: 0.6963 - val_acc: 0.6575 - val_auc: 0.7373

Epoch 00052: val_acc did not improve from 0.69937
Epoch 53/130
300/300 [=====] - 140s 467ms/step - loss: 0.6056 - acc: 0.7431 - auc: 0.7824 - val_loss: 0.6472 - val_acc: 0.6794 - val_auc: 0.7122

Epoch 00053: val_acc did not improve from 0.69937
Epoch 54/130
300/300 [=====] - 140s 467ms/step - loss: 0.5999 - acc: 0.7515 - auc: 0.7768 - val_loss: 0.6565 - val_acc: 0.6756 - val_auc: 0.7236

Epoch 00054: val_acc did not improve from 0.69937
Epoch 55/130
300/300 [=====] - 140s 467ms/step - loss: 0.5980 - acc: 0.7542 - auc: 0.7795 - val_loss: 0.6308 - val_acc: 0.7075 - val_auc: 0.7541

Epoch 00055: val_acc improved from 0.69937 to 0.70750, saving model to ./model3_1.h5
Epoch 56/130
300/300 [=====] - 140s 467ms/step - loss: 0.5980 - acc: 0.7483 - auc: 0.7865 - val_loss: 0.6472 - val_acc: 0.6987 - val_auc: 0.7624

Epoch 00056: val_acc did not improve from 0.70750
Epoch 57/130
300/300 [=====] - 140s 467ms/step - loss: 0.5913 - acc: 0.7579 - auc: 0.7965 - val_loss: 0.6467 - val_acc: 0.7013 - val_auc: 0.7632

Epoch 00057: val_acc did not improve from 0.70750
Epoch 58/130
300/300 [=====] - 140s 466ms/step - loss: 0.5863 - acc: 0.7667 - auc: 0.8131 - val_loss: 0.6311 - val_acc: 0.6919 - val_auc: 0.7485

Epoch 00058: val_acc did not improve from 0.70750
Epoch 59/130
300/300 [=====] - 140s 467ms/step - loss: 0.5868 - acc: 0.7648 - auc: 0.8025 - val_loss: 0.6150 - val_acc: 0.7188 - val_auc: 0.7634

Epoch 00059: val_acc improved from 0.70750 to 0.71875, saving model to ./model3_1.h5
Epoch 60/130
300/300 [=====] - 140s 467ms/step - loss: 0.5813 - acc: 0.7750 - auc: 0.8131 - val_loss: 0.6188 - val_acc: 0.7188 - val_auc: 0.7799

Epoch 00060: val_acc did not improve from 0.71875
Epoch 61/130
300/300 [=====] - 140s 466ms/step - loss: 0.5754 - acc: 0.7781 - auc: 0.8259 -
val_loss: 0.6152 - val_acc: 0.7344 - val_auc: 0.7845

Epoch 00061: val_acc improved from 0.71875 to 0.73438, saving model to ./model3_1.h5
Epoch 62/130
300/300 [=====] - 140s 467ms/step - loss: 0.5707 - acc: 0.7942 - auc: 0.8396 -
val_loss: 0.6223 - val_acc: 0.7063 - val_auc: 0.7625

Epoch 00062: val_acc did not improve from 0.73438
Epoch 63/130
300/300 [=====] - 140s 468ms/step - loss: 0.5734 - acc: 0.7837 - auc: 0.8316 -
val_loss: 0.6223 - val_acc: 0.7294 - val_auc: 0.7702

Epoch 00063: val_acc did not improve from 0.73438
Epoch 64/130
300/300 [=====] - 140s 467ms/step - loss: 0.5791 - acc: 0.7754 - auc: 0.8273 -
val_loss: 0.6181 - val_acc: 0.7425 - val_auc: 0.7912

Epoch 00064: val_acc improved from 0.73438 to 0.74250, saving model to ./model3_1.h5
Epoch 65/130
300/300 [=====] - 140s 467ms/step - loss: 0.5657 - acc: 0.7915 - auc: 0.8379 -
val_loss: 0.6176 - val_acc: 0.7119 - val_auc: 0.7698

Epoch 00065: val_acc did not improve from 0.74250
Epoch 66/130
300/300 [=====] - 140s 466ms/step - loss: 0.5625 - acc: 0.7960 - auc: 0.8374 -
val_loss: 0.6093 - val_acc: 0.7169 - val_auc: 0.7712

Epoch 00066: val_acc did not improve from 0.74250
Epoch 67/130
300/300 [=====] - 140s 467ms/step - loss: 0.5640 - acc: 0.7979 - auc: 0.8365 -
val_loss: 0.6033 - val_acc: 0.7412 - val_auc: 0.7873

Epoch 00067: val_acc did not improve from 0.74250
Epoch 68/130
300/300 [=====] - 140s 467ms/step - loss: 0.5581 - acc: 0.8013 - auc: 0.8487 -
val_loss: 0.6089 - val_acc: 0.7469 - val_auc: 0.7852

Epoch 00068: val_acc improved from 0.74250 to 0.74687, saving model to ./model3_1.h5
Epoch 69/130
300/300 [=====] - 140s 466ms/step - loss: 0.5592 - acc: 0.7973 - auc: 0.8441 -
val_loss: 0.6152 - val_acc: 0.7306 - val_auc: 0.7803

Epoch 00069: val_acc did not improve from 0.74687
Epoch 70/130
300/300 [=====] - 140s 467ms/step - loss: 0.5618 - acc: 0.7960 - auc: 0.8324 -
val_loss: 0.5992 - val_acc: 0.7438 - val_auc: 0.8031

Epoch 00070: val_acc did not improve from 0.74687
Epoch 71/130
300/300 [=====] - 140s 467ms/step - loss: 0.5497 - acc: 0.8094 - auc: 0.8616 -
val_loss: 0.6027 - val_acc: 0.7400 - val_auc: 0.7990

Epoch 00071: val_acc did not improve from 0.74687
Epoch 72/130
300/300 [=====] - 140s 467ms/step - loss: 0.5521 - acc: 0.8031 - auc: 0.8516 -
val_loss: 0.6141 - val_acc: 0.7344 - val_auc: 0.8013

Epoch 00072: val_acc did not improve from 0.74687
Epoch 73/130
300/300 [=====] - 140s 466ms/step - loss: 0.5509 - acc: 0.8083 - auc: 0.8499 -
val_loss: 0.5845 - val_acc: 0.7519 - val_auc: 0.8148

Epoch 00073: val_acc improved from 0.74687 to 0.75187, saving model to ./model3_1.h5
Epoch 74/130
300/300 [=====] - 140s 466ms/step - loss: 0.5414 - acc: 0.8173 - auc: 0.8619 -
val_loss: 0.6462 - val_acc: 0.7269 - val_auc: 0.7998

Epoch 00074: val_acc did not improve from 0.75187
Epoch 75/130
300/300 [=====] - 140s 466ms/step - loss: 0.5486 - acc: 0.8096 - auc: 0.8547 -
val_loss: 0.6123 - val_acc: 0.7406 - val_auc: 0.7936

Epoch 00075: val_acc did not improve from 0.75187

Epoch 76/130
300/300 [=====] - 140s 467ms/step - loss: 0.5421 - acc: 0.8140 - auc: 0.8611 - val_loss: 0.6126 - val_acc: 0.7306 - val_auc: 0.7937

Epoch 00076: val_acc did not improve from 0.75187

Epoch 77/130
300/300 [=====] - 140s 466ms/step - loss: 0.5475 - acc: 0.8083 - auc: 0.8576 - val_loss: 0.6061 - val_acc: 0.7400 - val_auc: 0.7874

Epoch 00077: val_acc did not improve from 0.75187

Epoch 78/130
300/300 [=====] - 140s 467ms/step - loss: 0.5420 - acc: 0.8121 - auc: 0.8590 - val_loss: 0.5836 - val_acc: 0.7562 - val_auc: 0.8131

Epoch 00078: val_acc improved from 0.75187 to 0.75625, saving model to ./model3_1.h5

Epoch 79/130
300/300 [=====] - 140s 466ms/step - loss: 0.5366 - acc: 0.8194 - auc: 0.8647 - val_loss: 0.5898 - val_acc: 0.7612 - val_auc: 0.8225

Epoch 00079: val_acc improved from 0.75625 to 0.76125, saving model to ./model3_1.h5

Epoch 80/130
300/300 [=====] - 140s 466ms/step - loss: 0.5319 - acc: 0.8269 - auc: 0.8682 - val_loss: 0.5941 - val_acc: 0.7450 - val_auc: 0.7975

Epoch 00080: val_acc did not improve from 0.76125

Epoch 81/130
300/300 [=====] - 140s 466ms/step - loss: 0.5321 - acc: 0.8217 - auc: 0.8749 - val_loss: 0.5881 - val_acc: 0.7531 - val_auc: 0.8130

Epoch 00081: val_acc did not improve from 0.76125

Epoch 82/130
300/300 [=====] - 140s 467ms/step - loss: 0.5282 - acc: 0.8269 - auc: 0.8712 - val_loss: 0.5990 - val_acc: 0.7512 - val_auc: 0.8063

Epoch 00082: val_acc did not improve from 0.76125

Epoch 83/130
300/300 [=====] - 140s 467ms/step - loss: 0.5302 - acc: 0.8196 - auc: 0.8666 - val_loss: 0.5860 - val_acc: 0.7612 - val_auc: 0.8231

Epoch 00083: val_acc did not improve from 0.76125

Epoch 84/130
300/300 [=====] - 140s 466ms/step - loss: 0.5262 - acc: 0.8279 - auc: 0.8765 - val_loss: 0.5902 - val_acc: 0.7462 - val_auc: 0.8103

Epoch 00084: val_acc did not improve from 0.76125

Epoch 85/130
300/300 [=====] - 140s 467ms/step - loss: 0.5237 - acc: 0.8294 - auc: 0.8767 - val_loss: 0.5860 - val_acc: 0.7800 - val_auc: 0.8431

Epoch 00085: val_acc improved from 0.76125 to 0.78000, saving model to ./model3_1.h5

Epoch 86/130
300/300 [=====] - 140s 466ms/step - loss: 0.5251 - acc: 0.8283 - auc: 0.8842 - val_loss: 0.6211 - val_acc: 0.7269 - val_auc: 0.7824

Epoch 00086: val_acc did not improve from 0.78000

Epoch 87/130
300/300 [=====] - 140s 466ms/step - loss: 0.5154 - acc: 0.8356 - auc: 0.8889 - val_loss: 0.5846 - val_acc: 0.7556 - val_auc: 0.8386

Epoch 00087: val_acc did not improve from 0.78000

Epoch 88/130
300/300 [=====] - 140s 466ms/step - loss: 0.5217 - acc: 0.8302 - auc: 0.8810 - val_loss: 0.5996 - val_acc: 0.7406 - val_auc: 0.8045

Epoch 00088: val_acc did not improve from 0.78000

Epoch 89/130
300/300 [=====] - 140s 467ms/step - loss: 0.5165 - acc: 0.8354 - auc: 0.8784 - val_loss: 0.6343 - val_acc: 0.7288 - val_auc: 0.8124

Epoch 00089: val_acc did not improve from 0.78000

Epoch 90/130
300/300 [=====] - 140s 466ms/step - loss: 0.5156 - acc: 0.8335 - auc: 0.8838 - val_loss: 0.6046 - val_acc: 0.7406 - val_auc: 0.8202

Epoch 00090: val_acc did not improve from 0.78000

Epoch 91/130
300/300 [=====] - 140s 467ms/step - loss: 0.5040 - acc: 0.8483 - auc: 0.8999 -

val_loss: 0.6223 - val_acc: 0.7356 - val_auc: 0.7944

Epoch 00091: val_acc did not improve from 0.78000

Epoch 92/130

300/300 [=====] - 140s 466ms/step - loss: 0.5182 - acc: 0.8308 - auc: 0.8694 -
val_loss: 0.5841 - val_acc: 0.7438 - val_auc: 0.8010

Epoch 00092: val_acc did not improve from 0.78000

Epoch 93/130

300/300 [=====] - 140s 467ms/step - loss: 0.5096 - acc: 0.8352 - auc: 0.8857 -
val_loss: 0.6084 - val_acc: 0.7306 - val_auc: 0.8089

Epoch 00093: val_acc did not improve from 0.78000

Epoch 94/130

300/300 [=====] - 140s 466ms/step - loss: 0.5044 - acc: 0.8465 - auc: 0.8961 -
val_loss: 0.6103 - val_acc: 0.7294 - val_auc: 0.8149

Epoch 00094: val_acc did not improve from 0.78000

Epoch 95/130

300/300 [=====] - 140s 467ms/step - loss: 0.5096 - acc: 0.8352 - auc: 0.8843 -
val_loss: 0.6157 - val_acc: 0.7425 - val_auc: 0.8166

Epoch 00095: val_acc did not improve from 0.78000

Epoch 96/130

300/300 [=====] - 140s 467ms/step - loss: 0.5004 - acc: 0.8510 - auc: 0.8940 -
val_loss: 0.5772 - val_acc: 0.7562 - val_auc: 0.8220

Epoch 00096: val_acc did not improve from 0.78000

Epoch 97/130

300/300 [=====] - 140s 466ms/step - loss: 0.5083 - acc: 0.8383 - auc: 0.8859 -
val_loss: 0.5922 - val_acc: 0.7519 - val_auc: 0.8105

Epoch 00097: val_acc did not improve from 0.78000

Epoch 98/130

300/300 [=====] - 140s 466ms/step - loss: 0.4968 - acc: 0.8500 - auc: 0.8942 -
val_loss: 0.5956 - val_acc: 0.7544 - val_auc: 0.8295

Epoch 00098: val_acc did not improve from 0.78000

Epoch 99/130

300/300 [=====] - 140s 466ms/step - loss: 0.4939 - acc: 0.8523 - auc: 0.8970 -
val_loss: 0.6376 - val_acc: 0.7275 - val_auc: 0.8004

Epoch 00099: val_acc did not improve from 0.78000

Epoch 100/130

300/300 [=====] - 140s 466ms/step - loss: 0.5041 - acc: 0.8392 - auc: 0.8831 -
val_loss: 0.6006 - val_acc: 0.7338 - val_auc: 0.8057

Epoch 00100: val_acc did not improve from 0.78000

Epoch 101/130

300/300 [=====] - 140s 467ms/step - loss: 0.4925 - acc: 0.8506 - auc: 0.8993 -
val_loss: 0.6389 - val_acc: 0.7325 - val_auc: 0.8216

Epoch 00101: val_acc did not improve from 0.78000

Epoch 102/130

300/300 [=====] - 141s 469ms/step - loss: 0.4912 - acc: 0.8496 - auc: 0.8996 -
val_loss: 0.6004 - val_acc: 0.7262 - val_auc: 0.7849

Epoch 00102: val_acc did not improve from 0.78000

Epoch 103/130

300/300 [=====] - 141s 470ms/step - loss: 0.4846 - acc: 0.8602 - auc: 0.9010 -
val_loss: 0.6214 - val_acc: 0.7188 - val_auc: 0.7960

Epoch 00103: val_acc did not improve from 0.78000

Epoch 104/130

300/300 [=====] - 141s 469ms/step - loss: 0.4800 - acc: 0.8604 - auc: 0.9058 -
val_loss: 0.5993 - val_acc: 0.7450 - val_auc: 0.8124

Epoch 00104: val_acc did not improve from 0.78000

Epoch 105/130

300/300 [=====] - 141s 469ms/step - loss: 0.4794 - acc: 0.8640 - auc: 0.9112 -
val_loss: 0.6297 - val_acc: 0.7538 - val_auc: 0.8209

Epoch 00105: val_acc did not improve from 0.78000

Epoch 00105: ReduceLROnPlateau reducing learning rate to 9.999999747378752e-07.

Epoch 106/130

300/300 [=====] - 141s 469ms/step - loss: 0.4788 - acc: 0.8587 - auc: 0.9056 -

val_loss: 0.5917 - val_acc: 0.7450 - val_auc: 0.8180

Epoch 00106: val_acc did not improve from 0.78000

Epoch 107/130

300/300 [=====] - 141s 469ms/step - loss: 0.4784 - acc: 0.8615 - auc: 0.9049 -
val_loss: 0.5864 - val_acc: 0.7638 - val_auc: 0.8362

Epoch 00107: val_acc did not improve from 0.78000

Epoch 108/130

300/300 [=====] - 142s 472ms/step - loss: 0.4733 - acc: 0.8631 - auc: 0.9116 -
val_loss: 0.5917 - val_acc: 0.7569 - val_auc: 0.8294

Epoch 00108: val_acc did not improve from 0.78000

Epoch 109/130

300/300 [=====] - 141s 471ms/step - loss: 0.4752 - acc: 0.8633 - auc: 0.9089 -
val_loss: 0.5833 - val_acc: 0.7762 - val_auc: 0.8398

Epoch 00109: val_acc did not improve from 0.78000

Epoch 110/130

300/300 [=====] - 141s 471ms/step - loss: 0.4691 - acc: 0.8708 - auc: 0.9135 -
val_loss: 0.5933 - val_acc: 0.7644 - val_auc: 0.8250

Epoch 00110: val_acc did not improve from 0.78000

Epoch 111/130

300/300 [=====] - 142s 473ms/step - loss: 0.4745 - acc: 0.8642 - auc: 0.9104 -
val_loss: 0.6011 - val_acc: 0.7494 - val_auc: 0.8130

Epoch 00111: val_acc did not improve from 0.78000

Epoch 112/130

300/300 [=====] - 141s 471ms/step - loss: 0.4696 - acc: 0.8721 - auc: 0.9123 -
val_loss: 0.6082 - val_acc: 0.7425 - val_auc: 0.8241

Epoch 00112: val_acc did not improve from 0.78000

Epoch 113/130

300/300 [=====] - 141s 471ms/step - loss: 0.4607 - acc: 0.8833 - auc: 0.9233 -
val_loss: 0.5930 - val_acc: 0.7512 - val_auc: 0.8175

Epoch 00113: val_acc did not improve from 0.78000

Epoch 114/130

300/300 [=====] - 141s 470ms/step - loss: 0.4688 - acc: 0.8723 - auc: 0.9183 -
val_loss: 0.6050 - val_acc: 0.7531 - val_auc: 0.8258

Epoch 00114: val_acc did not improve from 0.78000

Epoch 115/130

300/300 [=====] - 141s 471ms/step - loss: 0.4644 - acc: 0.8765 - auc: 0.9232 -
val_loss: 0.6032 - val_acc: 0.7519 - val_auc: 0.8220

Epoch 00115: val_acc did not improve from 0.78000

Epoch 116/130

300/300 [=====] - 141s 470ms/step - loss: 0.4618 - acc: 0.8740 - auc: 0.9251 -
val_loss: 0.6046 - val_acc: 0.7369 - val_auc: 0.8150

Epoch 00116: val_acc did not improve from 0.78000

Epoch 117/130

300/300 [=====] - 141s 471ms/step - loss: 0.4576 - acc: 0.8821 - auc: 0.9259 -
val_loss: 0.6077 - val_acc: 0.7381 - val_auc: 0.8252

Epoch 00117: val_acc did not improve from 0.78000

Epoch 118/130

300/300 [=====] - 141s 471ms/step - loss: 0.4632 - acc: 0.8808 - auc: 0.9201 -
val_loss: 0.6052 - val_acc: 0.7456 - val_auc: 0.8241

Epoch 00118: val_acc did not improve from 0.78000

Epoch 119/130

300/300 [=====] - 141s 471ms/step - loss: 0.4646 - acc: 0.8763 - auc: 0.9160 -
val_loss: 0.6101 - val_acc: 0.7444 - val_auc: 0.8238

Epoch 00119: val_acc did not improve from 0.78000

Epoch 120/130

300/300 [=====] - 141s 470ms/step - loss: 0.4618 - acc: 0.8781 - auc: 0.9190 -
val_loss: 0.5968 - val_acc: 0.7469 - val_auc: 0.8358

Epoch 00120: val_acc did not improve from 0.78000

Epoch 121/130

300/300 [=====] - 141s 470ms/step - loss: 0.4590 - acc: 0.8804 - auc: 0.9245 -
val_loss: 0.6051 - val_acc: 0.7462 - val_auc: 0.8228

```

Epoch 00121: val_acc did not improve from 0.78000
Epoch 122/130
300/300 [=====] - 141s 470ms/step - loss: 0.4512 - acc: 0.8858 - auc: 0.9302 -
val_loss: 0.6019 - val_acc: 0.7444 - val_auc: 0.8024

Epoch 00122: val_acc did not improve from 0.78000
Epoch 123/130
300/300 [=====] - 141s 471ms/step - loss: 0.4544 - acc: 0.8873 - auc: 0.9287 -
val_loss: 0.5858 - val_acc: 0.7600 - val_auc: 0.8431

Epoch 00123: val_acc did not improve from 0.78000
Epoch 124/130
300/300 [=====] - 141s 469ms/step - loss: 0.4547 - acc: 0.8869 - auc: 0.9258 -
val_loss: 0.6109 - val_acc: 0.7294 - val_auc: 0.8048

Epoch 00124: val_acc did not improve from 0.78000
Epoch 125/130
300/300 [=====] - 141s 469ms/step - loss: 0.4559 - acc: 0.8813 - auc: 0.9252 -
val_loss: 0.5906 - val_acc: 0.7375 - val_auc: 0.8211

Epoch 00125: val_acc did not improve from 0.78000

Epoch 00125: ReduceLROnPlateau reducing learning rate to 9.999999974752428e-08.
Epoch 126/130
300/300 [=====] - 141s 469ms/step - loss: 0.4601 - acc: 0.8777 - auc: 0.9191 -
val_loss: 0.5997 - val_acc: 0.7475 - val_auc: 0.8170

Epoch 00126: val_acc did not improve from 0.78000
Epoch 127/130
300/300 [=====] - 141s 469ms/step - loss: 0.4515 - acc: 0.8877 - auc: 0.9279 -
val_loss: 0.6114 - val_acc: 0.7488 - val_auc: 0.8147

Epoch 00127: val_acc did not improve from 0.78000
Epoch 128/130
300/300 [=====] - 141s 470ms/step - loss: 0.4549 - acc: 0.8865 - auc: 0.9286 -
val_loss: 0.6059 - val_acc: 0.7406 - val_auc: 0.8217

Epoch 00128: val_acc did not improve from 0.78000
Epoch 129/130
300/300 [=====] - 141s 469ms/step - loss: 0.4574 - acc: 0.8781 - auc: 0.9265 -
val_loss: 0.6136 - val_acc: 0.7381 - val_auc: 0.8189

Epoch 00129: val_acc did not improve from 0.78000
Epoch 130/130
300/300 [=====] - 141s 469ms/step - loss: 0.4560 - acc: 0.8823 - auc: 0.9265 -
val_loss: 0.5985 - val_acc: 0.7531 - val_auc: 0.8325

Epoch 00130: val_acc did not improve from 0.78000

```

Out[0]:

```
<keras.callbacks.History at 0x7facac017be0>
```

At epoch 85 we can see Train accuracy is 82 and validation accuracy is approx 78 so considering epoch 85 for trained model

Epoch 85/130 => acc: 0.8294 - val_acc: 0.7800 --> saving model to ./model3_1.h5

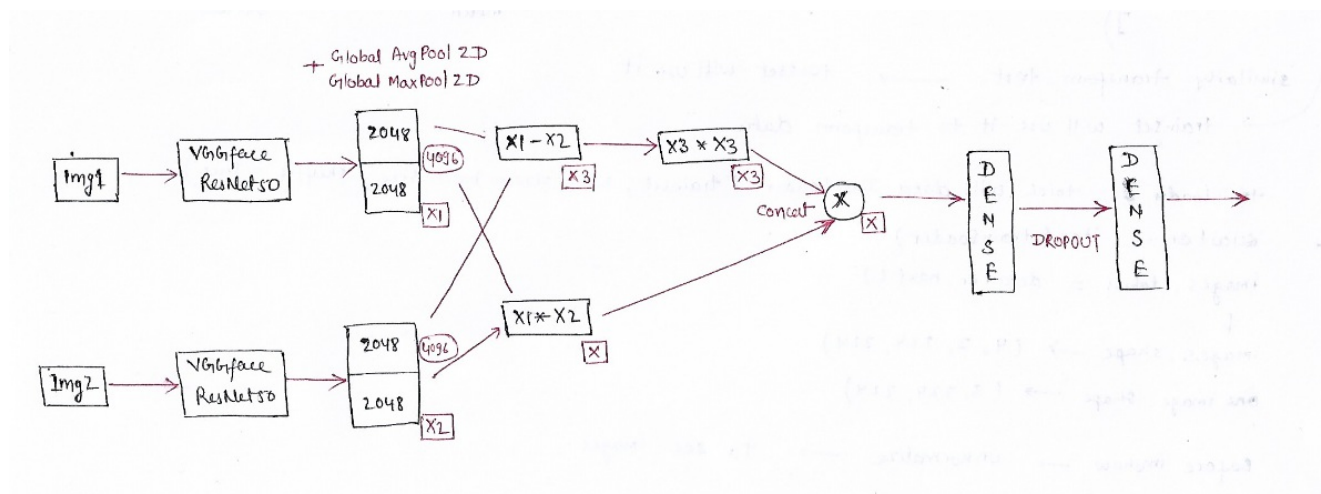
In [0]:

```
!mv model3_1.h5 ./drive/My\ Drive/COLAB_FILES/
```

In [0]:

In [0]:

MODEL 4



In [0]:

```
!ls
```

```
drive          setup.py  test.zip
image          test     tools
keras_vggface  test-faces.zip train
keras_vggface  test-private-faces.zip train-faces.zip
LICENSE.txt    test-private-labels.zip train_relationships.csv
MANIFEST.in    test-private-lists.zip train.zip
README.md      test-public-faces.zip visualize
sample_data    test-public-lists.zip
sample_submission.csv test.py
```

In [0]:

```
def read_img_cv(path):
    img = cv2.imread(path)
    img = np.array(img).astype(np.float)
    return preprocess_input(img, version=2)

def gen(list_tuples, person_to_images_map, batch_size=16):
    ppl = list(person_to_images_map.keys())
    while True:
        batch_tuples = sample(list_tuples, batch_size // 2)
        labels = [1] * len(batch_tuples)
        while len(batch_tuples) < batch_size:
            p1 = choice(ppl)
            p2 = choice(ppl)

            if p1 != p2 and (p1, p2) not in list_tuples and (p2, p1) not in list_tuples:
                batch_tuples.append((p1, p2))
                labels.append(0)

        for x in batch_tuples:
            if not len(person_to_images_map[x[0]]):
                print(x[0])

        X1 = [choice(person_to_images_map[x[0]]) for x in batch_tuples]
        X1 = np.array([read_img_cv(x) for x in X1])

        X2 = [choice(person_to_images_map[x[1]]) for x in batch_tuples]
        X2 = np.array([read_img_cv(x) for x in X2])

        yield [X1, X2], labels
```

In [0]:

In [0]:

```
def baseline_model4():
    input_1 = Input(shape=(224, 224, 3))
    input_2 = Input(shape=(224, 224, 3))

    base_model4 = VGGFace(model='resnet50', include_top=False)

    for x in base_model4.layers[:-3]:
        x.trainable = True

    x1 = base_model4(input_1)
    x2 = base_model4(input_2)

    x1 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x1), GlobalAvgPool2D() (x1)])
    x2 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x2), GlobalAvgPool2D() (x2)])

    x3 = Subtract() ([x1, x2])
    x3 = Multiply() ([x3, x3])

    x = Multiply() ([x1, x2])

    x = Concatenate(axis=-1) ([x, x3])
    x = BatchNormalization() (x)
    x = Dense(128, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)
    x = Activation('relu') (x)
    x = Dropout(0.4) (x)
    x = Dense(64, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)
    x = Activation('relu') (x)
    x = Dropout(0.4) (x)
    x = Dense(32, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)
    x = Activation('relu') (x)
    x = Dropout(0.3) (x)
    x = Dense(16, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)
    x = Activation('relu') (x)
    x = Dropout(0.3) (x)
    out = Dense(1, activation="sigmoid") (x)
    model4 = Model([input_1, input_2], out)
    model4.compile(loss="binary_crossentropy", metrics=['acc', auc], optimizer=Adam(0.00001))
    model4.summary()
    return model4
```

In [0]:

```
file_path4 = "./model4.h5"

checkpoint4 = ModelCheckpoint(file_path4, monitor='val_acc', verbose=1, save_best_only=True, mode='max'
)

reduce_on_plateau4 = ReduceLROnPlateau(monitor="val_acc", mode="max", factor=0.1, patience=20, verbose=
1)

callbacks_list4 = [checkpoint4, reduce_on_plateau4]

model4 = baseline_model4()
```

Model: "model_2"

Layer (type)	Output Shape	Param #	Connected to
input_4 (InputLayer)	(None, 224, 224, 3)	0	
input_5 (InputLayer)	(None, 224, 224, 3)	0	
vggface_resnet50 (Model)	multiple	23561152	input_4[0][0] input_5[0][0]
global_max_pooling2d_3 (GlobalM	(None, 2048)	0	vggface_resnet50[1][0]
global_average_pooling2d_3 (Glo	(None, 2048)	0	vggface_resnet50[1][0]
global_max_pooling2d_4 (GlobalM	(None, 2048)	0	vggface_resnet50[2][0]

global_average_pooling2d_4 (Global Average Pooling)	(None, 2048)	0	vggface_resnet50[2][0]
concatenate_4 (Concatenate)	(None, 4096)	0	global_max_pooling2d_3[0][0] global_average_pooling2d_3[0][0]
concatenate_5 (Concatenate)	(None, 4096)	0	global_max_pooling2d_4[0][0] global_average_pooling2d_4[0][0]
subtract_2 (Subtract)	(None, 4096)	0	concatenate_4[0][0] concatenate_5[0][0]
multiply_4 (Multiply)	(None, 4096)	0	concatenate_4[0][0] concatenate_5[0][0]
multiply_3 (Multiply)	(None, 4096)	0	subtract_2[0][0] subtract_2[0][0]
concatenate_6 (Concatenate)	(None, 8192)	0	multiply_4[0][0] multiply_3[0][0]
batch_normalization_2 (Batch Normalization)	(None, 8192)	32768	concatenate_6[0][0]
dense_6 (Dense)	(None, 128)	1048704	batch_normalization_2[0][0]
activation_103 (Activation)	(None, 128)	0	dense_6[0][0]
dropout_2 (Dropout)	(None, 128)	0	activation_103[0][0]
dense_7 (Dense)	(None, 64)	8256	dropout_2[0][0]
activation_104 (Activation)	(None, 64)	0	dense_7[0][0]
dropout_3 (Dropout)	(None, 64)	0	activation_104[0][0]
dense_8 (Dense)	(None, 32)	2080	dropout_3[0][0]
activation_105 (Activation)	(None, 32)	0	dense_8[0][0]
dropout_4 (Dropout)	(None, 32)	0	activation_105[0][0]
dense_9 (Dense)	(None, 16)	528	dropout_4[0][0]
activation_106 (Activation)	(None, 16)	0	dense_9[0][0]
dropout_5 (Dropout)	(None, 16)	0	activation_106[0][0]
dense_10 (Dense)	(None, 1)	17	dropout_5[0][0]

Total params: 24,653,505
 Trainable params: 24,584,001
 Non-trainable params: 69,504

In [0]:

```

model4.fit_generator(gen(train, train_person_to_images_map, batch_size=16), use_multiprocessing=True,
                    validation_data=gen(val, val_person_to_images_map, batch_size=16), epochs=100, verbose=1,
                    workers = 4, callbacks=callbacks_list4, steps_per_epoch=300, validation_steps=100)

```

Epoch 1/100
 300/300 [=====] - 116s 388ms/step - loss: 1.2418 - acc: 0.5033 - auc: 0.5185 - val_loss: 1.1100 - val_acc: 0.5181 - val_auc: 0.5350

Epoch 00001: val_acc improved from -inf to 0.51812, saving model to ./model4.h5

Epoch 2/100
 300/300 [=====] - 90s 301ms/step - loss: 1.2246 - acc: 0.4956 - auc: 0.5023 - val_loss: 1.1092 - val_acc: 0.5306 - val_auc: 0.5273

Epoch 00002: val_acc improved from 0.51812 to 0.53063, saving model to ./model4.h5

Epoch 3/100
 300/300 [=====] - 91s 302ms/step - loss: 1.2009 - acc: 0.5129 - auc: 0.5223 - val_loss: 1.0929 - val_acc: 0.5644 - val_auc: 0.5781

Epoch 00003: val_acc improved from 0.53063 to 0.56437, saving model to ./model4.h5

Epoch 4/100
300/300 [=====] - 90s 300ms/step - loss: 1.1889 - acc: 0.5252 - auc: 0.5292 -
val_loss: 1.0857 - val_acc: 0.5506 - val_auc: 0.5734

Epoch 00004: val_acc did not improve from 0.56437

Epoch 5/100
300/300 [=====] - 90s 299ms/step - loss: 1.1697 - acc: 0.5240 - auc: 0.5404 -
val_loss: 1.0731 - val_acc: 0.5763 - val_auc: 0.6123

Epoch 00005: val_acc improved from 0.56437 to 0.57625, saving model to ./model4.h5

Epoch 6/100
300/300 [=====] - 90s 300ms/step - loss: 1.1756 - acc: 0.5196 - auc: 0.5206 -
val_loss: 1.0631 - val_acc: 0.6156 - val_auc: 0.6533

Epoch 00006: val_acc improved from 0.57625 to 0.61562, saving model to ./model4.h5

Epoch 7/100
300/300 [=====] - 90s 301ms/step - loss: 1.1477 - acc: 0.5265 - auc: 0.5420 -
val_loss: 1.0733 - val_acc: 0.5950 - val_auc: 0.6373

Epoch 00007: val_acc did not improve from 0.61562

Epoch 8/100
300/300 [=====] - 90s 301ms/step - loss: 1.1426 - acc: 0.5442 - auc: 0.5535 -
val_loss: 1.0741 - val_acc: 0.5794 - val_auc: 0.6197

Epoch 00008: val_acc did not improve from 0.61562

Epoch 9/100
300/300 [=====] - 91s 303ms/step - loss: 1.1217 - acc: 0.5508 - auc: 0.5764 -
val_loss: 1.0535 - val_acc: 0.6025 - val_auc: 0.6577

Epoch 00009: val_acc did not improve from 0.61562

Epoch 10/100
300/300 [=====] - 90s 300ms/step - loss: 1.1234 - acc: 0.5435 - auc: 0.5748 -
val_loss: 1.0139 - val_acc: 0.6494 - val_auc: 0.7362

Epoch 00010: val_acc improved from 0.61562 to 0.64938, saving model to ./model4.h5

Epoch 11/100
300/300 [=====] - 90s 300ms/step - loss: 1.1116 - acc: 0.5533 - auc: 0.5861 -
val_loss: 1.0214 - val_acc: 0.6375 - val_auc: 0.7047

Epoch 00011: val_acc did not improve from 0.64938

Epoch 12/100
300/300 [=====] - 91s 302ms/step - loss: 1.1136 - acc: 0.5613 - auc: 0.5901 -
val_loss: 1.0275 - val_acc: 0.6481 - val_auc: 0.7267

Epoch 00012: val_acc did not improve from 0.64938

Epoch 13/100
300/300 [=====] - 90s 301ms/step - loss: 1.0968 - acc: 0.5723 - auc: 0.6023 -
val_loss: 1.0246 - val_acc: 0.6356 - val_auc: 0.6969

Epoch 00013: val_acc did not improve from 0.64938

Epoch 14/100
300/300 [=====] - 90s 301ms/step - loss: 1.0844 - acc: 0.5827 - auc: 0.6223 -
val_loss: 1.0113 - val_acc: 0.6606 - val_auc: 0.7381

Epoch 00014: val_acc improved from 0.64938 to 0.66063, saving model to ./model4.h5

Epoch 15/100
300/300 [=====] - 90s 300ms/step - loss: 1.0800 - acc: 0.5804 - auc: 0.6277 -
val_loss: 1.0078 - val_acc: 0.6488 - val_auc: 0.7209

Epoch 00015: val_acc did not improve from 0.66063

Epoch 16/100
300/300 [=====] - 91s 303ms/step - loss: 1.0814 - acc: 0.5831 - auc: 0.6310 -
val_loss: 1.0042 - val_acc: 0.6562 - val_auc: 0.7273

Epoch 00016: val_acc did not improve from 0.66063

Epoch 17/100
300/300 [=====] - 91s 303ms/step - loss: 1.0576 - acc: 0.5998 - auc: 0.6543 -
val_loss: 0.9937 - val_acc: 0.6787 - val_auc: 0.7517

Epoch 00017: val_acc improved from 0.66063 to 0.67875, saving model to ./model4.h5

Epoch 18/100
300/300 [=====] - 90s 301ms/step - loss: 1.0547 - acc: 0.6019 - auc: 0.6517 -
val_loss: 0.9830 - val_acc: 0.6913 - val_auc: 0.7625

Epoch 00018: val_acc improved from 0.67875 to 0.69125, saving model to ./model4.h5

Epoch 19/100
300/300 [=====] - 90s 302ms/step - loss: 1.0398 - acc: 0.6131 - auc: 0.6739 -

val_loss: 0.9673 - val_acc: 0.6800 - val_auc: 0.7795

Epoch 00019: val_acc did not improve from 0.69125

Epoch 20/100

300/300 [=====] - 90s 301ms/step - loss: 1.0377 - acc: 0.6252 - auc: 0.6814 -
val_loss: 0.9614 - val_acc: 0.6850 - val_auc: 0.7856

Epoch 00020: val_acc did not improve from 0.69125

Epoch 21/100

300/300 [=====] - 90s 300ms/step - loss: 1.0391 - acc: 0.6250 - auc: 0.6783 -
val_loss: 0.9561 - val_acc: 0.7194 - val_auc: 0.7927

Epoch 00021: val_acc improved from 0.69125 to 0.71937, saving model to ./model4.h5

Epoch 22/100

300/300 [=====] - 92s 305ms/step - loss: 1.0172 - acc: 0.6442 - auc: 0.7071 -
val_loss: 0.9618 - val_acc: 0.7013 - val_auc: 0.7825

Epoch 00022: val_acc did not improve from 0.71937

Epoch 23/100

300/300 [=====] - 91s 303ms/step - loss: 1.0091 - acc: 0.6371 - auc: 0.7102 -
val_loss: 0.9548 - val_acc: 0.7044 - val_auc: 0.7848

Epoch 00023: val_acc did not improve from 0.71937

Epoch 24/100

300/300 [=====] - 92s 305ms/step - loss: 1.0146 - acc: 0.6442 - auc: 0.7165 -
val_loss: 0.9530 - val_acc: 0.7063 - val_auc: 0.7934

Epoch 00024: val_acc did not improve from 0.71937

Epoch 25/100

300/300 [=====] - 91s 304ms/step - loss: 0.9949 - acc: 0.6642 - auc: 0.7284 -
val_loss: 0.9470 - val_acc: 0.7131 - val_auc: 0.7980

Epoch 00025: val_acc did not improve from 0.71937

Epoch 26/100

300/300 [=====] - 91s 304ms/step - loss: 1.0099 - acc: 0.6496 - auc: 0.7197 -
val_loss: 0.9553 - val_acc: 0.6850 - val_auc: 0.7816

Epoch 00026: val_acc did not improve from 0.71937

Epoch 27/100

300/300 [=====] - 91s 304ms/step - loss: 1.0050 - acc: 0.6579 - auc: 0.7211 -
val_loss: 0.9434 - val_acc: 0.6981 - val_auc: 0.8013

Epoch 00027: val_acc did not improve from 0.71937

Epoch 28/100

300/300 [=====] - 92s 307ms/step - loss: 0.9926 - acc: 0.6635 - auc: 0.7380 -
val_loss: 0.9512 - val_acc: 0.6963 - val_auc: 0.7939

Epoch 00028: val_acc did not improve from 0.71937

Epoch 29/100

300/300 [=====] - 91s 303ms/step - loss: 0.9960 - acc: 0.6627 - auc: 0.7327 -
val_loss: 0.9455 - val_acc: 0.6981 - val_auc: 0.7877

Epoch 00029: val_acc did not improve from 0.71937

Epoch 30/100

300/300 [=====] - 91s 304ms/step - loss: 0.9757 - acc: 0.6767 - auc: 0.7502 -
val_loss: 0.9444 - val_acc: 0.7000 - val_auc: 0.7936

Epoch 00030: val_acc did not improve from 0.71937

Epoch 31/100

300/300 [=====] - 91s 303ms/step - loss: 0.9703 - acc: 0.6794 - auc: 0.7554 -
val_loss: 0.9177 - val_acc: 0.7181 - val_auc: 0.8172

Epoch 00031: val_acc did not improve from 0.71937

Epoch 32/100

300/300 [=====] - 91s 302ms/step - loss: 0.9710 - acc: 0.6783 - auc: 0.7609 -
val_loss: 0.9304 - val_acc: 0.7137 - val_auc: 0.8105

Epoch 00032: val_acc did not improve from 0.71937

Epoch 33/100

300/300 [=====] - 91s 305ms/step - loss: 0.9657 - acc: 0.6898 - auc: 0.7627 -
val_loss: 0.9182 - val_acc: 0.7238 - val_auc: 0.8223

Epoch 00033: val_acc improved from 0.71937 to 0.72375, saving model to ./model4.h5

Epoch 34/100

300/300 [=====] - 91s 303ms/step - loss: 0.9627 - acc: 0.6840 - auc: 0.7627 -
val_loss: 0.9401 - val_acc: 0.7069 - val_auc: 0.7983

Epoch 00034: val_acc did not improve from 0.72375
Epoch 35/100
300/300 [=====] - 90s 302ms/step - loss: 0.9460 - acc: 0.6917 - auc: 0.7740 -
val_loss: 0.9454 - val_acc: 0.7231 - val_auc: 0.7978

Epoch 00035: val_acc did not improve from 0.72375
Epoch 36/100
300/300 [=====] - 91s 303ms/step - loss: 0.9443 - acc: 0.7038 - auc: 0.7854 -
val_loss: 0.9280 - val_acc: 0.7013 - val_auc: 0.8070

Epoch 00036: val_acc did not improve from 0.72375
Epoch 37/100
300/300 [=====] - 92s 307ms/step - loss: 0.9296 - acc: 0.7148 - auc: 0.7903 -
val_loss: 0.9134 - val_acc: 0.7106 - val_auc: 0.8208

Epoch 00037: val_acc did not improve from 0.72375
Epoch 38/100
300/300 [=====] - 91s 303ms/step - loss: 0.9286 - acc: 0.7081 - auc: 0.7877 -
val_loss: 0.8897 - val_acc: 0.7406 - val_auc: 0.8389

Epoch 00038: val_acc improved from 0.72375 to 0.74062, saving model to ./model4.h5
Epoch 39/100
300/300 [=====] - 91s 302ms/step - loss: 0.9296 - acc: 0.7117 - auc: 0.7892 -
val_loss: 0.9165 - val_acc: 0.7106 - val_auc: 0.8172

Epoch 00039: val_acc did not improve from 0.74062
Epoch 40/100
300/300 [=====] - 90s 301ms/step - loss: 0.9171 - acc: 0.7302 - auc: 0.8057 -
val_loss: 0.9357 - val_acc: 0.7137 - val_auc: 0.8114

Epoch 00040: val_acc did not improve from 0.74062
Epoch 41/100
300/300 [=====] - 90s 301ms/step - loss: 0.9019 - acc: 0.7235 - auc: 0.8105 -
val_loss: 0.9135 - val_acc: 0.7331 - val_auc: 0.8164

Epoch 00041: val_acc did not improve from 0.74062
Epoch 42/100
300/300 [=====] - 91s 302ms/step - loss: 0.9077 - acc: 0.7285 - auc: 0.8075 -
val_loss: 0.8947 - val_acc: 0.7269 - val_auc: 0.8364

Epoch 00042: val_acc did not improve from 0.74062
Epoch 43/100
300/300 [=====] - 91s 303ms/step - loss: 0.8980 - acc: 0.7379 - auc: 0.8129 -
val_loss: 0.9102 - val_acc: 0.7181 - val_auc: 0.8216

Epoch 00043: val_acc did not improve from 0.74062
Epoch 44/100
300/300 [=====] - 90s 300ms/step - loss: 0.8891 - acc: 0.7413 - auc: 0.8268 -
val_loss: 0.8860 - val_acc: 0.7350 - val_auc: 0.8411

Epoch 00044: val_acc did not improve from 0.74062
Epoch 45/100
300/300 [=====] - 90s 299ms/step - loss: 0.9032 - acc: 0.7321 - auc: 0.8145 -
val_loss: 0.9284 - val_acc: 0.7238 - val_auc: 0.8319

Epoch 00045: val_acc did not improve from 0.74062
Epoch 46/100
300/300 [=====] - 90s 299ms/step - loss: 0.8807 - acc: 0.7392 - auc: 0.8244 -
val_loss: 0.9059 - val_acc: 0.7281 - val_auc: 0.8264

Epoch 00046: val_acc did not improve from 0.74062
Epoch 47/100
300/300 [=====] - 90s 300ms/step - loss: 0.8784 - acc: 0.7452 - auc: 0.8286 -
val_loss: 0.9060 - val_acc: 0.7262 - val_auc: 0.8241

Epoch 00047: val_acc did not improve from 0.74062
Epoch 48/100
300/300 [=====] - 90s 300ms/step - loss: 0.8616 - acc: 0.7637 - auc: 0.8465 -
val_loss: 0.9210 - val_acc: 0.7037 - val_auc: 0.8217

Epoch 00048: val_acc did not improve from 0.74062
Epoch 49/100
300/300 [=====] - 90s 300ms/step - loss: 0.8693 - acc: 0.7625 - auc: 0.8441 -
val_loss: 0.9327 - val_acc: 0.7206 - val_auc: 0.8152

Epoch 00049: val_acc did not improve from 0.74062
Epoch 50/100

300/300 [=====] - 90s 301ms/step - loss: 0.8670 - acc: 0.7654 - auc: 0.8445 -
val_loss: 0.8857 - val_acc: 0.7425 - val_auc: 0.8295

Epoch 00050: val_acc improved from 0.74062 to 0.74250, saving model to ./model4.h5
Epoch 51/100
300/300 [=====] - 89s 297ms/step - loss: 0.8516 - acc: 0.7688 - auc: 0.8560 -
val_loss: 0.9308 - val_acc: 0.7212 - val_auc: 0.7941

Epoch 00051: val_acc did not improve from 0.74250
Epoch 52/100
300/300 [=====] - 90s 299ms/step - loss: 0.8444 - acc: 0.7717 - auc: 0.8507 -
val_loss: 0.9076 - val_acc: 0.7244 - val_auc: 0.8395

Epoch 00052: val_acc did not improve from 0.74250
Epoch 53/100
300/300 [=====] - 90s 300ms/step - loss: 0.8336 - acc: 0.7754 - auc: 0.8585 -
val_loss: 0.9352 - val_acc: 0.7056 - val_auc: 0.8263

Epoch 00053: val_acc did not improve from 0.74250
Epoch 54/100
300/300 [=====] - 90s 299ms/step - loss: 0.8125 - acc: 0.7881 - auc: 0.8712 -
val_loss: 0.9096 - val_acc: 0.7244 - val_auc: 0.8189

Epoch 00054: val_acc did not improve from 0.74250
Epoch 55/100
300/300 [=====] - 89s 298ms/step - loss: 0.8358 - acc: 0.7865 - auc: 0.8639 -
val_loss: 0.9338 - val_acc: 0.7156 - val_auc: 0.8070

Epoch 00055: val_acc did not improve from 0.74250
Epoch 56/100
300/300 [=====] - 90s 299ms/step - loss: 0.8110 - acc: 0.7921 - auc: 0.8760 -
val_loss: 0.9426 - val_acc: 0.7056 - val_auc: 0.8178

Epoch 00056: val_acc did not improve from 0.74250
Epoch 57/100
300/300 [=====] - 90s 299ms/step - loss: 0.8225 - acc: 0.7931 - auc: 0.8737 -
val_loss: 0.8981 - val_acc: 0.7431 - val_auc: 0.8530

Epoch 00057: val_acc improved from 0.74250 to 0.74313, saving model to ./model4.h5
Epoch 58/100
300/300 [=====] - 90s 298ms/step - loss: 0.8082 - acc: 0.7992 - auc: 0.8817 -
val_loss: 0.9506 - val_acc: 0.7144 - val_auc: 0.8080

Epoch 00058: val_acc did not improve from 0.74313
Epoch 59/100
300/300 [=====] - 90s 300ms/step - loss: 0.7974 - acc: 0.8077 - auc: 0.8855 -
val_loss: 0.9466 - val_acc: 0.7163 - val_auc: 0.8264

Epoch 00059: val_acc did not improve from 0.74313
Epoch 60/100
300/300 [=====] - 90s 300ms/step - loss: 0.7993 - acc: 0.8013 - auc: 0.8794 -
val_loss: 0.9198 - val_acc: 0.7331 - val_auc: 0.8295

Epoch 00060: val_acc did not improve from 0.74313
Epoch 61/100
300/300 [=====] - 90s 299ms/step - loss: 0.7895 - acc: 0.8073 - auc: 0.8848 -
val_loss: 0.9070 - val_acc: 0.7419 - val_auc: 0.8458

Epoch 00061: val_acc did not improve from 0.74313
Epoch 62/100
300/300 [=====] - 90s 300ms/step - loss: 0.7597 - acc: 0.8217 - auc: 0.8999 -
val_loss: 0.9757 - val_acc: 0.7094 - val_auc: 0.8206

Epoch 00062: val_acc did not improve from 0.74313
Epoch 63/100
300/300 [=====] - 92s 305ms/step - loss: 0.7785 - acc: 0.8169 - auc: 0.8909 -
val_loss: 0.9635 - val_acc: 0.7150 - val_auc: 0.8084

Epoch 00063: val_acc did not improve from 0.74313
Epoch 64/100
300/300 [=====] - 89s 298ms/step - loss: 0.7685 - acc: 0.8204 - auc: 0.8982 -
val_loss: 0.9358 - val_acc: 0.7250 - val_auc: 0.8302

Epoch 00064: val_acc did not improve from 0.74313
Epoch 65/100
300/300 [=====] - 89s 298ms/step - loss: 0.7638 - acc: 0.8206 - auc: 0.8963 -
val_loss: 0.9910 - val_acc: 0.7081 - val_auc: 0.8108

Epoch 00065: val_acc did not improve from 0.74313
Epoch 66/100
300/300 [=====] - 89s 296ms/step - loss: 0.7384 - acc: 0.8335 - auc: 0.9105 -
val_loss: 0.9782 - val_acc: 0.7131 - val_auc: 0.8192

Epoch 00066: val_acc did not improve from 0.74313
Epoch 67/100
300/300 [=====] - 89s 297ms/step - loss: 0.7505 - acc: 0.8308 - auc: 0.9070 -
val_loss: 0.9475 - val_acc: 0.7181 - val_auc: 0.8305

Epoch 00067: val_acc did not improve from 0.74313
Epoch 68/100
300/300 [=====] - 89s 297ms/step - loss: 0.7518 - acc: 0.8262 - auc: 0.9030 -
val_loss: 1.0039 - val_acc: 0.6994 - val_auc: 0.8148

Epoch 00068: val_acc did not improve from 0.74313
Epoch 69/100
300/300 [=====] - 89s 297ms/step - loss: 0.7443 - acc: 0.8375 - auc: 0.9099 -
val_loss: 0.9958 - val_acc: 0.7113 - val_auc: 0.8252

Epoch 00069: val_acc did not improve from 0.74313
Epoch 70/100
300/300 [=====] - 89s 297ms/step - loss: 0.7282 - acc: 0.8408 - auc: 0.9134 -
val_loss: 0.9926 - val_acc: 0.7044 - val_auc: 0.8183

Epoch 00070: val_acc did not improve from 0.74313
Epoch 71/100
300/300 [=====] - 88s 295ms/step - loss: 0.7271 - acc: 0.8512 - auc: 0.9149 -
val_loss: 1.0256 - val_acc: 0.7044 - val_auc: 0.8141

Epoch 00071: val_acc did not improve from 0.74313
Epoch 72/100
300/300 [=====] - 89s 295ms/step - loss: 0.7365 - acc: 0.8373 - auc: 0.9123 -
val_loss: 0.9672 - val_acc: 0.7219 - val_auc: 0.8358

Epoch 00072: val_acc did not improve from 0.74313
Epoch 73/100
300/300 [=====] - 88s 295ms/step - loss: 0.7297 - acc: 0.8394 - auc: 0.9170 -
val_loss: 0.9631 - val_acc: 0.7125 - val_auc: 0.8281

Epoch 00073: val_acc did not improve from 0.74313
Epoch 74/100
300/300 [=====] - 89s 296ms/step - loss: 0.7067 - acc: 0.8575 - auc: 0.9230 -
val_loss: 0.9780 - val_acc: 0.7231 - val_auc: 0.8403

Epoch 00074: val_acc did not improve from 0.74313
Epoch 75/100
300/300 [=====] - 89s 295ms/step - loss: 0.7100 - acc: 0.8558 - auc: 0.9232 -
val_loss: 0.9968 - val_acc: 0.7156 - val_auc: 0.8327

Epoch 00075: val_acc did not improve from 0.74313
Epoch 76/100
300/300 [=====] - 89s 297ms/step - loss: 0.6885 - acc: 0.8610 - auc: 0.9307 -
val_loss: 0.9604 - val_acc: 0.7225 - val_auc: 0.8466

Epoch 00076: val_acc did not improve from 0.74313
Epoch 77/100
300/300 [=====] - 89s 296ms/step - loss: 0.7028 - acc: 0.8627 - auc: 0.9258 -
val_loss: 1.0680 - val_acc: 0.7131 - val_auc: 0.8197

Epoch 00077: val_acc did not improve from 0.74313
Epoch 78/100
300/300 [=====] - 89s 297ms/step - loss: 0.6858 - acc: 0.8712 - auc: 0.9329 -
val_loss: 1.0294 - val_acc: 0.7137 - val_auc: 0.8411

Epoch 00078: val_acc did not improve from 0.74313
Epoch 79/100
300/300 [=====] - 89s 297ms/step - loss: 0.6834 - acc: 0.8621 - auc: 0.9311 -
val_loss: 1.0052 - val_acc: 0.7044 - val_auc: 0.8170

Epoch 00079: val_acc did not improve from 0.74313
Epoch 80/100
300/300 [=====] - 90s 298ms/step - loss: 0.6700 - acc: 0.8679 - auc: 0.9388 -
val_loss: 1.0162 - val_acc: 0.7119 - val_auc: 0.8234

Epoch 00080: val_acc did not improve from 0.74313
Epoch 81/100
300/300 [=====] - 89s 298ms/step - loss: 0.6709 - acc: 0.8710 - auc: 0.9371 -
val_loss: 1.0375 - val_acc: 0.7075 - val_auc: 0.8366

Epoch 00081: val_acc did not improve from 0.74313
Epoch 82/100
300/300 [=====] - 89s 297ms/step - loss: 0.6660 - acc: 0.8725 - auc: 0.9385 -
val_loss: 0.9707 - val_acc: 0.7181 - val_auc: 0.8444

Epoch 00082: val_acc did not improve from 0.74313
Epoch 83/100
300/300 [=====] - 89s 296ms/step - loss: 0.6655 - acc: 0.8727 - auc: 0.9406 -
val_loss: 1.0353 - val_acc: 0.7000 - val_auc: 0.8300

Epoch 00083: val_acc did not improve from 0.74313
Epoch 84/100
300/300 [=====] - 89s 297ms/step - loss: 0.6806 - acc: 0.8675 - auc: 0.9356 -
val_loss: 1.0364 - val_acc: 0.7144 - val_auc: 0.8494

Epoch 00084: val_acc did not improve from 0.74313
Epoch 85/100
300/300 [=====] - 89s 296ms/step - loss: 0.6654 - acc: 0.8777 - auc: 0.9412 -
val_loss: 1.0220 - val_acc: 0.7231 - val_auc: 0.8400

Epoch 00085: val_acc did not improve from 0.74313
Epoch 86/100
300/300 [=====] - 89s 296ms/step - loss: 0.6592 - acc: 0.8775 - auc: 0.9432 -
val_loss: 1.0002 - val_acc: 0.7200 - val_auc: 0.8403

Epoch 00086: val_acc did not improve from 0.74313
Epoch 87/100
300/300 [=====] - 89s 298ms/step - loss: 0.6619 - acc: 0.8700 - auc: 0.9392 -
val_loss: 1.0169 - val_acc: 0.7169 - val_auc: 0.8455

Epoch 00087: val_acc did not improve from 0.74313
Epoch 88/100
300/300 [=====] - 89s 297ms/step - loss: 0.6563 - acc: 0.8810 - auc: 0.9435 -
val_loss: 1.0063 - val_acc: 0.7019 - val_auc: 0.8305

Epoch 00088: val_acc did not improve from 0.74313
Epoch 89/100
300/300 [=====] - 90s 299ms/step - loss: 0.6622 - acc: 0.8800 - auc: 0.9424 -
val_loss: 1.0016 - val_acc: 0.7131 - val_auc: 0.8392

Epoch 00089: val_acc did not improve from 0.74313
Epoch 90/100
300/300 [=====] - 89s 297ms/step - loss: 0.6458 - acc: 0.8840 - auc: 0.9460 -
val_loss: 1.0323 - val_acc: 0.7013 - val_auc: 0.8311

Epoch 00090: val_acc did not improve from 0.74313
Epoch 91/100
300/300 [=====] - 89s 297ms/step - loss: 0.6412 - acc: 0.8904 - auc: 0.9483 -
val_loss: 1.0452 - val_acc: 0.7100 - val_auc: 0.8256

Epoch 00091: val_acc did not improve from 0.74313
Epoch 92/100
300/300 [=====] - 89s 296ms/step - loss: 0.6600 - acc: 0.8731 - auc: 0.9409 -
val_loss: 1.0238 - val_acc: 0.7031 - val_auc: 0.8372

Epoch 00092: val_acc did not improve from 0.74313
Epoch 93/100
300/300 [=====] - 89s 298ms/step - loss: 0.6558 - acc: 0.8846 - auc: 0.9461 -
val_loss: 1.0364 - val_acc: 0.7025 - val_auc: 0.8286

Epoch 00093: val_acc did not improve from 0.74313
Epoch 94/100
300/300 [=====] - 89s 297ms/step - loss: 0.6608 - acc: 0.8754 - auc: 0.9409 -
val_loss: 1.0276 - val_acc: 0.7069 - val_auc: 0.8314

Epoch 00094: val_acc did not improve from 0.74313
Epoch 95/100
300/300 [=====] - 89s 298ms/step - loss: 0.6418 - acc: 0.8915 - auc: 0.9511 -
val_loss: 1.0458 - val_acc: 0.7069 - val_auc: 0.8348

Epoch 00095: val acc did not improve from 0.74313

```

Epoch 96/100
300/300 [=====] - 89s 298ms/step - loss: 0.6368 - acc: 0.8842 - auc: 0.9489 -
val_loss: 1.0220 - val_acc: 0.7306 - val_auc: 0.8447

Epoch 00096: val_acc did not improve from 0.74313
Epoch 97/100
300/300 [=====] - 89s 297ms/step - loss: 0.6506 - acc: 0.8835 - auc: 0.9469 -
val_loss: 1.0212 - val_acc: 0.7338 - val_auc: 0.8538

Epoch 00097: val_acc did not improve from 0.74313

Epoch 00097: ReduceLROnPlateau reducing learning rate to 9.999999974752428e-08.
Epoch 98/100
300/300 [=====] - 89s 296ms/step - loss: 0.6315 - acc: 0.8946 - auc: 0.9521 -
val_loss: 1.0288 - val_acc: 0.7238 - val_auc: 0.8434

Epoch 00098: val_acc did not improve from 0.74313
Epoch 99/100
300/300 [=====] - 90s 299ms/step - loss: 0.6443 - acc: 0.8917 - auc: 0.9490 -
val_loss: 1.0658 - val_acc: 0.7087 - val_auc: 0.8228

Epoch 00099: val_acc did not improve from 0.74313
Epoch 100/100
300/300 [=====] - 89s 296ms/step - loss: 0.6381 - acc: 0.8944 - auc: 0.9523 -
val_loss: 1.0473 - val_acc: 0.7200 - val_auc: 0.8497

Epoch 00100: val_acc did not improve from 0.74313

```

Out[0]:

```
<keras.callbacks.History at 0x7f21e5ed86d8>
```

At epoch 57 we can see Train accuracy is 79 and validation accuracy is approx 74 so considering epoch 57 for trained model

Epoch 57/100 => acc: 0.7931 - val_acc: 0.7431 --> saving model to ./model4.h5

In [0]:

```
!mv model4.h5 ./drive/My\ Drive/COILAB/SIMILAR_FACES/
```

MODEL 5

In [36]:

```
from random import choice, sample
```

In [37]:

```

def read_img_cv(path):
    img = cv2.imread(path)
    img = np.array(img).astype(np.float)
    return preprocess_input(img, version=2)

def gen(list_tuples, person_to_images_map, batch_size=16):
    ppl = list(person_to_images_map.keys())
    while True:
        batch_tuples = sample(list_tuples, batch_size // 2)
        labels = [1] * len(batch_tuples)
        while len(batch_tuples) < batch_size:
            p1 = choice(ppl)
            p2 = choice(ppl)

            if p1 != p2 and (p1, p2) not in list_tuples and (p2, p1) not in list_tuples:
                batch_tuples.append((p1, p2))
                labels.append(0)

        for x in batch_tuples:
            if not len(person_to_images_map[x[0]]):

```

```

        print(x[0])

    X1 = [choice(person_to_images_map[x[0]]) for x in batch_tuples]
    X1 = np.array([read_img_cv(x) for x in X1])

    X2 = [choice(person_to_images_map[x[1]]) for x in batch_tuples]
    X2 = np.array([read_img_cv(x) for x in X2])

    yield [X1, X2], labels

```

In [38]:

```

import gc
import psutil
print("available RAM:", psutil.virtual_memory())
gc.collect()
print("available RAM:", psutil.virtual_memory())

```

```

available RAM: svmem(total=54881505280, available=53530177536, percent=2.5, used=788684800, free=533423
75936, active=781651968, inactive=473862144, buffers=72650752, cached=677793792, shared=12136448, slab=
82657280)
available RAM: svmem(total=54881505280, available=53530177536, percent=2.5, used=788676608, free=533423
75936, active=781651968, inactive=473862144, buffers=72658944, cached=677793792, shared=12136448, slab=
82657280)

```

In [39]:

```

from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau, EarlyStopping
from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, MaxPooling1D, GlobalAveragePooling1D, LSTM
from keras.layers import Concatenate, Multiply, Dropout, Subtract, Add, Conv2D, Conv1D
from keras.models import Model
from keras.layers import BatchNormalization, Reshape
from keras.preprocessing import image
from keras.optimizers import Adam, SGD, rmsprop
import h5py
from keras.regularizers import l2
from keras.losses import binary_crossentropy
import tensorflow as tf
from sklearn.metrics import roc_auc_score
from keras.layers import LeakyReLU
from keras import regularizers
from keras.initializers import RandomNormal

def auc(y_true, y_pred):
    return tf.py_function(roc_auc_score, (y_true, y_pred), tf.double)

```

In [40]:

```

def siamese_model():
    input_1 = Input(shape=(224, 224, 3))
    input_2 = Input(shape=(224, 224, 3))
    base_model = VGGFace(model='resnet50', include_top=False, input_shape=(224, 224, 3))

    for x in base_model.layers[:-3]:
        x.trainable = True

    x1 = base_model(input_1)
    x2 = base_model(input_2)

    x1 = Concatenate(axis=-1) ([GlobalAvgPool2D()(x1), GlobalAvgPool2D()(x1)])
    x2 = Concatenate(axis=-1) ([GlobalMaxPool2D()(x2), GlobalAvgPool2D()(x2)])
    x3 = Subtract() ([x1, x2])
    x3 = Multiply() ([x3, x3])
    x1_ = Multiply() ([x1, x1])
    x2_ = Multiply() ([x2, x2])
    x4 = Subtract() ([x1_, x2_])
    x = Concatenate(axis=-1) ([x4, x3])
    x = Reshape(input_shape=(8192,), target_shape=(8192, 1)) (x)
    x = Conv1D(filters=100,
               kernel_size=8,
               strides=8, input_shape=(8192, 1))

```



```

        strides=0, input_shape=( 64, 2, 1),
        activation='relu',
        padding='same')(x)
    x = Conv1D( 100, 8, activation= 'relu')(x)
    x = LSTM(128, kernel_initializer='glorot_normal', recurrent_dropout=0.5, kernel_regularizer=l2(0.001),
return_sequences=True)(x)
    #x = LSTM(128, return_sequences=True)(x)
    x = Dropout(0.2)(x)
    x = LSTM(128, kernel_initializer='glorot_normal', recurrent_dropout=0.5, kernel_regularizer=l2(0.001),
return_sequences=True)(x)
    #x = LSTM(128, return_sequences=True)(x)
    x = BatchNormalization ()(x)
    x = Flatten()(x)
    x = Dense(100, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001))(x)
    x = Activation('relu')(x)
    x = Dropout(0.2)(x)
    x = Dense(64, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001))(x)
    x = Activation('relu')(x)
    x = Dropout(0.1)(x)
    x = Dense(32, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001))(x)
    x = Activation('relu')(x)
    x = Dropout(0.1)(x)
    x = Dense(16, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001))(x)
    x = Activation('relu')(x)
    x = Dropout(0.01)(x)
    out = Dense(1, activation="sigmoid")(x)
    model = Model([input_1, input_2], out)
    model.compile(loss="binary_crossentropy", metrics=['acc', auc], optimizer=Adam(0.00001, decay=1e-6))
    model.summary()

    return model

```

In []:

In [41]:

```

file_path5 = "./model5.h5"

checkpoint5 = ModelCheckpoint(file_path5, monitor='val_acc', verbose=1, save_best_only=True, mode='max'
)
early_stopping5 = EarlyStopping(monitor='val_loss', min_delta=0.0001, patience=20, verbose=0, mode='au
to')
reduce_on_plateau5 = ReduceLROnPlateau(monitor="val_loss", mode="max", factor=0.2, patience=10, verbose
=1)

callbacks_list5 = [checkpoint5, reduce_on_plateau5, early_stopping5]

model5 = siamese_model()

```

WARNING:tensorflow:From /usr/local/lib/python3.5/dist-packages/tensorflow_core/python/ops/resource_variable_ops.py:1630: calling BaseResourceVariable.__init__ (from tensorflow.python.ops.resource_variable_o

ps) with constraint is deprecated and will be removed in a future version.

Instructions for updating:

If using Keras pass *_constraint arguments to layers.

WARNING:tensorflow:From /usr/local/lib/python3.5/dist-packages/keras/backend/tensorflow_backend.py:4070 : The name tf.nn.max_pool is deprecated. Please use tf.nn.max_pool2d instead.

WARNING:tensorflow:From /usr/local/lib/python3.5/dist-packages/keras/backend/tensorflow_backend.py:4074 : The name tf.nn.avg_pool is deprecated. Please use tf.nn.avg_pool2d instead.

WARNING:tensorflow:From /usr/local/lib/python3.5/dist-packages/tensorflow_core/python/ops/nn_impl.py:183: where (from tensorflow.python.ops.array_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use tf.where in 2.0, which has the same broadcast rule as np.where

Model: "model_1"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 224, 224, 3)	0	
input_2 (InputLayer)	(None, 224, 224, 3)	0	

vggface_resnet50 (Model)	(None, 1, 1, 2048)	23561152	input_1[0][0] input_2[0][0]
global_average_pooling2d_1 (GlobalAveragePooling2D)	(None, 2048)	0	vggface_resnet50[1][0]
global_average_pooling2d_2 (GlobalAveragePooling2D)	(None, 2048)	0	vggface_resnet50[1][0]
global_max_pooling2d_1 (GlobalMaxPooling2D)	(None, 2048)	0	vggface_resnet50[2][0]
global_average_pooling2d_3 (GlobalAveragePooling2D)	(None, 2048)	0	vggface_resnet50[2][0]
concatenate_1 (Concatenate)	(None, 4096)	0	global_average_pooling2d_1[0][0] global_average_pooling2d_2[0][0]
concatenate_2 (Concatenate)	(None, 4096)	0	global_max_pooling2d_1[0][0] global_average_pooling2d_3[0][0]
multiply_2 (Multiply)	(None, 4096)	0	concatenate_1[0][0] concatenate_1[0][0]
multiply_3 (Multiply)	(None, 4096)	0	concatenate_2[0][0] concatenate_2[0][0]
subtract_1 (Subtract)	(None, 4096)	0	concatenate_1[0][0] concatenate_2[0][0]
subtract_2 (Subtract)	(None, 4096)	0	multiply_2[0][0] multiply_3[0][0]
multiply_1 (Multiply)	(None, 4096)	0	subtract_1[0][0] subtract_1[0][0]
concatenate_3 (Concatenate)	(None, 8192)	0	subtract_2[0][0] multiply_1[0][0]
reshape_1 (Reshape)	(None, 8192, 1)	0	concatenate_3[0][0]
conv1d_1 (Conv1D)	(None, 1024, 100)	900	reshape_1[0][0]
conv1d_2 (Conv1D)	(None, 1017, 100)	80100	conv1d_1[0][0]
lstm_1 (LSTM)	(None, 1017, 128)	117248	conv1d_2[0][0]
dropout_1 (Dropout)	(None, 1017, 128)	0	lstm_1[0][0]
lstm_2 (LSTM)	(None, 1017, 128)	131584	dropout_1[0][0]
batch_normalization_1 (Batch Normalization)	(None, 1017, 128)	512	lstm_2[0][0]
flatten_1 (Flatten)	(None, 130176)	0	batch_normalization_1[0][0]
dense_1 (Dense)	(None, 100)	13017700	flatten_1[0][0]
activation_50 (Activation)	(None, 100)	0	dense_1[0][0]
dropout_2 (Dropout)	(None, 100)	0	activation_50[0][0]
dense_2 (Dense)	(None, 64)	6464	dropout_2[0][0]
activation_51 (Activation)	(None, 64)	0	dense_2[0][0]
dropout_3 (Dropout)	(None, 64)	0	activation_51[0][0]
dense_3 (Dense)	(None, 32)	2080	dropout_3[0][0]
activation_52 (Activation)	(None, 32)	0	dense_3[0][0]
dropout_4 (Dropout)	(None, 32)	0	activation_52[0][0]
dense_4 (Dense)	(None, 16)	528	dropout_4[0][0]
activation_53 (Activation)	(None, 16)	0	dense_4[0][0]
dropout_5 (Dropout)	(None, 16)	0	activation_53[0][0]
dense_5 (Dense)	(None, 1)	17	dropout_5[0][0]

Total params: 36,918,285
Trainable params: 36,864,909
Non-trainable params: 53,376

In [42]:

```
model5.fit_generator(gen(train, train_person_to_images_map, batch_size=8), use_multiprocessing=True,  
                    validation_data=gen(val, val_person_to_images_map, batch_size=8), epochs=30, verbo  
se=1,  
                    workers=16, callbacks=callbacks_list5, steps_per_epoch=300, validation_steps=100)
```

WARNING:tensorflow:From /usr/local/lib/python3.5/dist-packages/keras/backend/tensorflow_backend.py:422:
The name tf.global_variables is deprecated. Please use tf.compat.v1.global_variables instead.

Epoch 1/30

300/300 [=====] - 1400s 5s/step - loss: 1.4704 - acc: 0.4983 - auc: 0.4863 - v
al_loss: 1.4107 - val_acc: 0.5387 - val_auc: 0.5569

Epoch 00001: val_auc improved from -inf to 0.55687, saving model to ./model5.h5

Epoch 2/30

300/300 [=====] - 1376s 5s/step - loss: 1.4383 - acc: 0.5208 - auc: 0.5354 - v
al_loss: 1.3606 - val_acc: 0.5962 - val_auc: 0.6494

Epoch 00002: val_auc improved from 0.55687 to 0.64938, saving model to ./model5.h5

Epoch 3/30

300/300 [=====] - 1362s 5s/step - loss: 1.4213 - acc: 0.5412 - auc: 0.5569 - v
al_loss: 1.3343 - val_acc: 0.6275 - val_auc: 0.6687

Epoch 00003: val_auc improved from 0.64938 to 0.66875, saving model to ./model5.h5

Epoch 4/30

300/300 [=====] - 1356s 5s/step - loss: 1.4031 - acc: 0.5592 - auc: 0.6006 - v
al_loss: 1.3580 - val_acc: 0.7013 - val_auc: 0.7681

Epoch 00004: val_auc improved from 0.66875 to 0.76812, saving model to ./model5.h5

Epoch 5/30

300/300 [=====] - 1359s 5s/step - loss: 1.3732 - acc: 0.5942 - auc: 0.6317 - v
al_loss: 1.5030 - val_acc: 0.6725 - val_auc: 0.7256

Epoch 00005: val_auc did not improve from 0.76812

Epoch 6/30

300/300 [=====] - 1349s 4s/step - loss: 1.3886 - acc: 0.5967 - auc: 0.6317 - v
al_loss: 1.2673 - val_acc: 0.6787 - val_auc: 0.7594

Epoch 00006: val_auc did not improve from 0.76812

Epoch 7/30

300/300 [=====] - 1342s 4s/step - loss: 1.3453 - acc: 0.6229 - auc: 0.6702 - v
al_loss: 1.3428 - val_acc: 0.6550 - val_auc: 0.7406

Epoch 00007: val_auc did not improve from 0.76812

Epoch 8/30

300/300 [=====] - 1341s 4s/step - loss: 1.3376 - acc: 0.6371 - auc: 0.6900 - v
al_loss: 1.1294 - val_acc: 0.6438 - val_auc: 0.7688

Epoch 00008: val_auc improved from 0.76812 to 0.76875, saving model to ./model5.h5

Epoch 9/30

300/300 [=====] - 1382s 5s/step - loss: 1.3288 - acc: 0.6508 - auc: 0.6981 - v
al_loss: 1.2949 - val_acc: 0.6837 - val_auc: 0.8188

Epoch 00009: val_auc improved from 0.76875 to 0.81875, saving model to ./model5.h5

Epoch 10/30

300/300 [=====] - 1403s 5s/step - loss: 1.3290 - acc: 0.6263 - auc: 0.6835 - v
al_loss: 1.2635 - val_acc: 0.6137 - val_auc: 0.7788

Epoch 00010: val_auc did not improve from 0.81875

Epoch 11/30

300/300 [=====] - 1391s 5s/step - loss: 1.3033 - acc: 0.6571 - auc: 0.7233 - v
al_loss: 1.2668 - val_acc: 0.6125 - val_auc: 0.7925

Epoch 00011: val_auc did not improve from 0.81875

Epoch 12/30

300/300 [=====] - 1405s 5s/step - loss: 1.3092 - acc: 0.6654 - auc: 0.7217 - v
al_loss: 1.6192 - val_acc: 0.6187 - val_auc: 0.8031

Epoch 00012: val_auc did not improve from 0.81875

Epoch 12/30

```
Epoch 13/30
300/300 [=====] - 1392s 5s/step - loss: 1.2747 - acc: 0.6904 - auc: 0.7421 - v
al_loss: 1.3045 - val_acc: 0.6812 - val_auc: 0.8087

Epoch 00013: val_auc did not improve from 0.81875
Epoch 14/30
300/300 [=====] - 1407s 5s/step - loss: 1.2737 - acc: 0.6854 - auc: 0.7535 - v
al_loss: 1.2144 - val_acc: 0.6888 - val_auc: 0.8181

Epoch 00014: val_auc did not improve from 0.81875
Epoch 15/30
300/300 [=====] - 1402s 5s/step - loss: 1.2688 - acc: 0.6883 - auc: 0.7600 - v
al_loss: 1.4941 - val_acc: 0.6700 - val_auc: 0.8469

Epoch 00015: val_auc improved from 0.81875 to 0.84688, saving model to ./model5.h5
Epoch 16/30
300/300 [=====] - 1413s 5s/step - loss: 1.2532 - acc: 0.7042 - auc: 0.7650 - v
al_loss: 1.0344 - val_acc: 0.6637 - val_auc: 0.8594

Epoch 00016: val_auc improved from 0.84688 to 0.85938, saving model to ./model5.h5
Epoch 17/30
300/300 [=====] - 1401s 5s/step - loss: 1.2357 - acc: 0.7054 - auc: 0.7763 - v
al_loss: 1.5083 - val_acc: 0.6587 - val_auc: 0.8500

Epoch 00017: val_auc did not improve from 0.85938
Epoch 18/30
300/300 [=====] - 1393s 5s/step - loss: 1.2434 - acc: 0.7088 - auc: 0.7704 - v
al_loss: 1.3254 - val_acc: 0.6888 - val_auc: 0.8388

Epoch 00018: val_auc did not improve from 0.85938
Epoch 19/30
300/300 [=====] - 1403s 5s/step - loss: 1.2278 - acc: 0.6988 - auc: 0.7725 - v
al_loss: 0.9712 - val_acc: 0.6575 - val_auc: 0.8425

Epoch 00019: val_auc did not improve from 0.85938
Epoch 20/30
300/300 [=====] - 1403s 5s/step - loss: 1.2278 - acc: 0.7192 - auc: 0.7858 - v
al_loss: 1.2096 - val_acc: 0.6913 - val_auc: 0.8319

Epoch 00020: val_auc did not improve from 0.85938
Epoch 21/30
300/300 [=====] - 1387s 5s/step - loss: 1.2153 - acc: 0.7063 - auc: 0.7904 - v
al_loss: 1.5849 - val_acc: 0.7212 - val_auc: 0.8381

Epoch 00021: val_auc did not improve from 0.85938
Epoch 22/30
300/300 [=====] - 1384s 5s/step - loss: 1.1996 - acc: 0.7325 - auc: 0.8087 - v
al_loss: 1.6508 - val_acc: 0.6712 - val_auc: 0.8419

Epoch 00022: val_auc did not improve from 0.85938
Epoch 23/30
300/300 [=====] - 1401s 5s/step - loss: 1.1929 - acc: 0.7258 - auc: 0.8096 - v
al_loss: 1.0912 - val_acc: 0.6800 - val_auc: 0.8363

Epoch 00023: val_auc did not improve from 0.85938
Epoch 24/30
300/300 [=====] - 1403s 5s/step - loss: 1.1646 - acc: 0.7442 - auc: 0.8198 - v
al_loss: 1.0079 - val_acc: 0.6913 - val_auc: 0.8506

Epoch 00024: val_auc did not improve from 0.85938
Epoch 25/30
300/300 [=====] - 1402s 5s/step - loss: 1.1683 - acc: 0.7271 - auc: 0.8121 - v
al_loss: 0.8623 - val_acc: 0.7013 - val_auc: 0.8431

Epoch 00025: val_auc did not improve from 0.85938
Epoch 26/30
300/300 [=====] - 1401s 5s/step - loss: 1.1551 - acc: 0.7412 - auc: 0.8171 - v
al_loss: 0.9408 - val_acc: 0.6950 - val_auc: 0.8456

Epoch 00026: val_auc did not improve from 0.85938
Epoch 27/30
300/300 [=====] - 1413s 5s/step - loss: 1.1627 - acc: 0.7367 - auc: 0.8079 - v
al_loss: 1.2624 - val_acc: 0.6850 - val_auc: 0.8350

Epoch 00027: val_auc did not improve from 0.85938
Epoch 28/30
300/300 [=====] - 1408s 5s/step - loss: 1.1403 - acc: 0.7417 - auc: 0.8290 - v
al_loss: 1.0550 - val_acc: 0.6825 - val_auc: 0.8270
```

```
al_loss: 1.0559 - val_acc: 0.6925 - val_auc: 0.8378
```

```
Epoch 00028: val_auc did not improve from 0.85938
```

```
Epoch 29/30
```

```
300/300 [=====] - 1414s 5s/step - loss: 1.1394 - acc: 0.7387 - auc: 0.8129 - v  
al_loss: 1.3482 - val_acc: 0.6913 - val_auc: 0.8406
```

```
Epoch 00029: val_auc did not improve from 0.85938
```

```
Epoch 30/30
```

```
300/300 [=====] - 1401s 5s/step - loss: 1.1142 - acc: 0.7433 - auc: 0.8256 - v  
al_loss: 1.1110 - val_acc: 0.7038 - val_auc: 0.8300
```

```
Epoch 00030: val_auc did not improve from 0.85938
```

```
Out[42]:
```

```
<keras.callbacks.callbacks.History at 0x7f5a27bfc080>
```

```
In [44]:
```

```
def siamese_model_1():  
    input_1 = Input(shape=(224, 224, 3))  
    input_2 = Input(shape=(224, 224, 3))  
    base_model = VGGFace(model='resnet50', include_top=False, input_shape=(224, 224, 3))  
  
    for x in base_model.layers[:-3]:  
        x.trainable = True  
  
    x1 = base_model(input_1)  
    x2 = base_model(input_2)  
  
    x1 = Concatenate(axis=-1) ([GlobalAvgPool2D() (x1), GlobalAvgPool2D() (x1)])  
    x2 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x2), GlobalAvgPool2D() (x2)])  
    x3 = Subtract() ([x1, x2])  
    x3 = Multiply() ([x3, x3])  
    x1_ = Multiply() ([x1, x1])  
    x2_ = Multiply() ([x2, x2])  
    x4 = Subtract() ([x1_, x2_])  
    x = Concatenate(axis=-1) ([x4, x3])  
    x = Reshape(input_shape=(8192,), target_shape=(8192, 1)) (x)  
    x = Conv1D(filters=100,  
               kernel_size=8,  
               strides=8, input_shape=( 8192,1),  
               activation='relu',  
               padding='same') (x)  
    x = Conv1D( 100, activation='relu') (x)  
    x = LSTM(128, kernel_initializer='glorot_normal', recurrent_dropout=0.5, kernel_regularizer=l2(0.001),  
return_sequences=True) (x)  
    #x = LSTM(128, return_sequences=True) (x)  
    x = Dropout(0.4) (x)  
    x = LSTM(128, kernel_initializer='glorot_normal', recurrent_dropout=0.5, kernel_regularizer=l2(0.001),  
return_sequences=True) (x)  
    #x = LSTM(128, return_sequences=True) (x)  
    x = BatchNormalization () (x)  
    x = Flatten() (x)  
    x = Dense(100, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)  
    x = Activation('relu') (x)  
    x = Dropout(0.3) (x)  
    x = Dense(64, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)  
    x = Activation('relu') (x)  
    x = Dropout(0.2) (x)  
    x = Dense(32, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)  
    x = Activation('relu') (x)  
    x = Dropout(0.2) (x)  
    x = Dense(16, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)  
    x = Activation('relu') (x)  
    x = Dropout(0.1) (x)  
    out = Dense(1, activation="sigmoid") (x)  
    model = Model([input_1, input_2], out)  
    model.compile(loss="binary_crossentropy", metrics=['acc', auc], optimizer=Adam(0.00001, decay=1e-6))  
    model.summary()  
    return model
```

```
In [ ]:
```

In [47]:

```
file_path5 = "./model5_1.h5"

checkpoint5_1 = ModelCheckpoint(file_path5, monitor='val_acc', verbose=1, save_best_only=True, mode='max')
early_stopping5_1 = EarlyStopping(monitor='val_loss', min_delta=0.0001, patience=50, verbose=0, mode='auto')
reduce_on_plateau5_1 = ReduceLROnPlateau(monitor="val_loss", mode="max", factor=0.2, patience=10, verbose=1)

callbacks_list5_1 = [checkpoint5_1, reduce_on_plateau5_1, early_stopping5_1]

model5_1 = siamese_model_1()
```

Model: "model_3"

Layer (type)	Output Shape	Param #	Connected to
input_7 (InputLayer)	(None, 224, 224, 3)	0	
input_8 (InputLayer)	(None, 224, 224, 3)	0	
vggface_resnet50 (Model)	(None, 1, 1, 2048)	23561152	input_7[0][0] input_8[0][0]
global_average_pooling2d_7 (GlobalAveragePooling2D)	(None, 2048)	0	vggface_resnet50[1][0]
global_average_pooling2d_8 (GlobalAveragePooling2D)	(None, 2048)	0	vggface_resnet50[1][0]
global_max_pooling2d_3 (GlobalMaxPooling2D)	(None, 2048)	0	vggface_resnet50[2][0]
global_average_pooling2d_9 (GlobalAveragePooling2D)	(None, 2048)	0	vggface_resnet50[2][0]
concatenate_7 (Concatenate)	(None, 4096)	0	global_average_pooling2d_7[0][0] global_average_pooling2d_8[0][0]
concatenate_8 (Concatenate)	(None, 4096)	0	global_max_pooling2d_3[0][0] global_average_pooling2d_9[0][0]
multiply_8 (Multiply)	(None, 4096)	0	concatenate_7[0][0] concatenate_7[0][0]
multiply_9 (Multiply)	(None, 4096)	0	concatenate_8[0][0] concatenate_8[0][0]
subtract_5 (Subtract)	(None, 4096)	0	concatenate_7[0][0] concatenate_8[0][0]
subtract_6 (Subtract)	(None, 4096)	0	multiply_8[0][0] multiply_9[0][0]
multiply_7 (Multiply)	(None, 4096)	0	subtract_5[0][0] subtract_5[0][0]
concatenate_9 (Concatenate)	(None, 8192)	0	subtract_6[0][0] multiply_7[0][0]
reshape_3 (Reshape)	(None, 8192, 1)	0	concatenate_9[0][0]
conv1d_5 (Conv1D)	(None, 1024, 100)	900	reshape_3[0][0]
conv1d_6 (Conv1D)	(None, 1017, 100)	80100	conv1d_5[0][0]
lstm_5 (LSTM)	(None, 1017, 128)	117248	conv1d_6[0][0]
dropout_11 (Dropout)	(None, 1017, 128)	0	lstm_5[0][0]
lstm_6 (LSTM)	(None, 1017, 128)	131584	dropout_11[0][0]
batch_normalization_3 (Batch Normalization)	(None, 1017, 128)	512	lstm_6[0][0]
flatten_3 (Flatten)	(None, 130176)	0	batch_normalization_3[0][0]

flatten_3 (Flatten)	(None, 1301770)	0	batch_normalization_3[0][0]
dense_11 (Dense)	(None, 100)	13017700	flatten_3[0][0]
activation_156 (Activation)	(None, 100)	0	dense_11[0][0]
dropout_12 (Dropout)	(None, 100)	0	activation_156[0][0]
dense_12 (Dense)	(None, 64)	6464	dropout_12[0][0]
activation_157 (Activation)	(None, 64)	0	dense_12[0][0]
dropout_13 (Dropout)	(None, 64)	0	activation_157[0][0]
dense_13 (Dense)	(None, 32)	2080	dropout_13[0][0]
activation_158 (Activation)	(None, 32)	0	dense_13[0][0]
dropout_14 (Dropout)	(None, 32)	0	activation_158[0][0]
dense_14 (Dense)	(None, 16)	528	dropout_14[0][0]
activation_159 (Activation)	(None, 16)	0	dense_14[0][0]
dropout_15 (Dropout)	(None, 16)	0	activation_159[0][0]
dense_15 (Dense)	(None, 1)	17	dropout_15[0][0]
=====			
Total params: 36,918,285			
Trainable params: 36,864,909			
Non-trainable params: 53,376			
=====			

In [49]:

```
model5_1.fit_generator(gen(train, train_person_to_images_map, batch_size=16), use_multiprocessing=True,
                        validation_data=gen(val, val_person_to_images_map, batch_size=16), epochs=50, verbose=1,
                        workers=16, callbacks=callbacks_list5_1, steps_per_epoch=300, validation_steps=100
)
```

Epoch 1/50

300/300 [=====] - 2412s 8s/step - loss: 1.4332 - acc: 0.5325 - auc: 0.5481 - val_loss: 1.3648 - val_acc: 0.6044 - val_auc: 0.6998

Epoch 00001: val_acc improved from -inf to 0.60438, saving model to ./model5_1.h5

Epoch 2/50

300/300 [=====] - 2408s 8s/step - loss: 1.3899 - acc: 0.5738 - auc: 0.6127 - val_loss: 1.4123 - val_acc: 0.6244 - val_auc: 0.7108

Epoch 00002: val_acc improved from 0.60438 to 0.62437, saving model to ./model5_1.h5

Epoch 3/50

300/300 [=====] - 2409s 8s/step - loss: 1.3729 - acc: 0.5960 - auc: 0.6380 - val_loss: 1.5730 - val_acc: 0.6150 - val_auc: 0.7177

Epoch 00003: val_acc did not improve from 0.62437

Epoch 4/50

300/300 [=====] - 2412s 8s/step - loss: 1.3514 - acc: 0.6183 - auc: 0.6661 - val_loss: 1.4229 - val_acc: 0.6344 - val_auc: 0.7167

Epoch 00004: val_acc improved from 0.62437 to 0.63437, saving model to ./model5_1.h5

Epoch 5/50

300/300 [=====] - 2411s 8s/step - loss: 1.3341 - acc: 0.6354 - auc: 0.6838 - val_loss: 1.3731 - val_acc: 0.6612 - val_auc: 0.7200

Epoch 00005: val_acc improved from 0.63437 to 0.66125, saving model to ./model5_1.h5

Epoch 6/50

300/300 [=====] - 2428s 8s/step - loss: 1.3323 - acc: 0.6319 - auc: 0.6779 - val_loss: 1.2908 - val_acc: 0.6463 - val_auc: 0.7120

Epoch 00006: val_acc did not improve from 0.66125

Epoch 7/50

300/300 [=====] - 2421s 8s/step - loss: 1.3067 - acc: 0.6488 - auc: 0.7086 - val_loss: 1.2434 - val_acc: 0.6650 - val_auc: 0.7425

Epoch 00007: val_acc improved from 0.66125 to 0.66500, saving model to ./model5_1.h5

Epoch 00007: val_acc improved from 0.66500 to 0.69563, saving model to ./model5_1.h5
Epoch 8/50
300/300 [=====] - 2434s 8s/step - loss: 1.3046 - acc: 0.6554 - auc: 0.7046 - val_loss: 1.3182 - val_acc: 0.6956 - val_auc: 0.7788

Epoch 00008: val_acc improved from 0.66500 to 0.69563, saving model to ./model5_1.h5
Epoch 9/50
300/300 [=====] - 2449s 8s/step - loss: 1.2894 - acc: 0.6660 - auc: 0.7302 - val_loss: 1.2035 - val_acc: 0.6956 - val_auc: 0.7677

Epoch 00009: val_acc did not improve from 0.69563
Epoch 10/50
300/300 [=====] - 2457s 8s/step - loss: 1.2601 - acc: 0.6888 - auc: 0.7574 - val_loss: 1.2242 - val_acc: 0.7088 - val_auc: 0.7806

Epoch 00010: val_acc improved from 0.69563 to 0.70875, saving model to ./model5_1.h5
Epoch 11/50
300/300 [=====] - 2452s 8s/step - loss: 1.2319 - acc: 0.7171 - auc: 0.7835 - val_loss: 1.1536 - val_acc: 0.7669 - val_auc: 0.8366

Epoch 00011: val_acc improved from 0.70875 to 0.76688, saving model to ./model5_1.h5
Epoch 12/50
300/300 [=====] - 2452s 8s/step - loss: 1.2137 - acc: 0.7360 - auc: 0.8021 - val_loss: 1.1006 - val_acc: 0.7563 - val_auc: 0.8238

Epoch 00012: val_acc did not improve from 0.76688
Epoch 13/50
300/300 [=====] - 2417s 8s/step - loss: 1.2017 - acc: 0.7333 - auc: 0.8048 - val_loss: 1.0603 - val_acc: 0.7825 - val_auc: 0.8727

Epoch 00013: val_acc improved from 0.76688 to 0.78250, saving model to ./model5_1.h5

Epoch 00013: ReduceLROnPlateau reducing learning rate to 1.9999999494757505e-06.
Epoch 14/50
300/300 [=====] - 2379s 8s/step - loss: 1.1694 - acc: 0.7519 - auc: 0.8269 - val_loss: 1.1775 - val_acc: 0.7950 - val_auc: 0.8672

Epoch 00014: val_acc improved from 0.78250 to 0.79500, saving model to ./model5_1.h5
Epoch 15/50
300/300 [=====] - 2408s 8s/step - loss: 1.1418 - acc: 0.7631 - auc: 0.8474 - val_loss: 1.1670 - val_acc: 0.7744 - val_auc: 0.8558

Epoch 00015: val_acc did not improve from 0.79500
Epoch 16/50
300/300 [=====] - 2392s 8s/step - loss: 1.1494 - acc: 0.7646 - auc: 0.8412 - val_loss: 1.3725 - val_acc: 0.7769 - val_auc: 0.8577

Epoch 00016: val_acc did not improve from 0.79500
Epoch 17/50
300/300 [=====] - 2377s 8s/step - loss: 1.1512 - acc: 0.7635 - auc: 0.8360 - val_loss: 1.1046 - val_acc: 0.7806 - val_auc: 0.8589

Epoch 00017: val_acc did not improve from 0.79500
Epoch 18/50
300/300 [=====] - 2372s 8s/step - loss: 1.1449 - acc: 0.7696 - auc: 0.8458 - val_loss: 1.1398 - val_acc: 0.7950 - val_auc: 0.8673

Epoch 00018: val_acc did not improve from 0.79500
Epoch 19/50
300/300 [=====] - 2379s 8s/step - loss: 1.1292 - acc: 0.7704 - auc: 0.8536 - val_loss: 1.1963 - val_acc: 0.7700 - val_auc: 0.8561

Epoch 00019: val_acc did not improve from 0.79500
Epoch 20/50
300/300 [=====] - 2412s 8s/step - loss: 1.1284 - acc: 0.7823 - auc: 0.8575 - val_loss: 1.0475 - val_acc: 0.7756 - val_auc: 0.8678

Epoch 00020: val_acc did not improve from 0.79500
Epoch 21/50
300/300 [=====] - 2402s 8s/step - loss: 1.0996 - acc: 0.7973 - auc: 0.8761 - val_loss: 1.2326 - val_acc: 0.7887 - val_auc: 0.8745

Epoch 00021: val_acc did not improve from 0.79500
Epoch 22/50
300/300 [=====] - 2454s 8s/step - loss: 1.0984 - acc: 0.7946 - auc: 0.8736 - val_loss: 1.1242 - val_acc: 0.7775 - val_auc: 0.8689

Epoch 00022: val_acc did not improve from 0.79500

Epoch 00022: val_acc did not improve from 0.79937
Epoch 23/50
300/300 [=====] - 2452s 8s/step - loss: 1.0939 - acc: 0.7954 - auc: 0.8720 - val_loss: 1.2493 - val_acc: 0.7969 - val_auc: 0.8823

Epoch 00023: val_acc improved from 0.79500 to 0.79688, saving model to ./model5_1.h5

Epoch 00023: ReduceLROnPlateau reducing learning rate to 3.99999989900971e-07.
Epoch 24/50
300/300 [=====] - 2446s 8s/step - loss: 1.0879 - acc: 0.8031 - auc: 0.8758 - val_loss: 1.2046 - val_acc: 0.7994 - val_auc: 0.8900

Epoch 00024: val_acc improved from 0.79688 to 0.79937, saving model to ./model5_1.h5
Epoch 25/50
300/300 [=====] - 2452s 8s/step - loss: 1.1012 - acc: 0.7912 - auc: 0.8664 - val_loss: 1.1487 - val_acc: 0.7763 - val_auc: 0.8687

Epoch 00025: val_acc did not improve from 0.79937
Epoch 26/50
300/300 [=====] - 2451s 8s/step - loss: 1.0909 - acc: 0.8008 - auc: 0.8720 - val_loss: 1.1911 - val_acc: 0.7719 - val_auc: 0.8656

Epoch 00026: val_acc did not improve from 0.79937
Epoch 27/50
300/300 [=====] - 2473s 8s/step - loss: 1.0735 - acc: 0.8067 - auc: 0.8867 - val_loss: 1.0655 - val_acc: 0.7887 - val_auc: 0.8712

Epoch 00027: val_acc did not improve from 0.79937
Epoch 28/50
300/300 [=====] - 2479s 8s/step - loss: 1.0813 - acc: 0.8006 - auc: 0.8783 - val_loss: 0.9317 - val_acc: 0.7919 - val_auc: 0.8722

Epoch 00028: val_acc did not improve from 0.79937
Epoch 29/50
300/300 [=====] - 2465s 8s/step - loss: 1.0790 - acc: 0.8073 - auc: 0.8802 - val_loss: 1.0364 - val_acc: 0.7738 - val_auc: 0.8553

Epoch 00029: val_acc did not improve from 0.79937
Epoch 30/50
300/300 [=====] - 2468s 8s/step - loss: 1.0745 - acc: 0.8117 - auc: 0.8859 - val_loss: 1.0505 - val_acc: 0.8012 - val_auc: 0.8731

Epoch 00030: val_acc improved from 0.79937 to 0.80125, saving model to ./model5_1.h5
Epoch 31/50
300/300 [=====] - 2478s 8s/step - loss: 1.0897 - acc: 0.7946 - auc: 0.8733 - val_loss: 0.8719 - val_acc: 0.8175 - val_auc: 0.8997

Epoch 00031: val_acc improved from 0.80125 to 0.81750, saving model to ./model5_1.h5
Epoch 32/50
300/300 [=====] - 2474s 8s/step - loss: 1.0733 - acc: 0.8133 - auc: 0.8852 - val_loss: 1.3500 - val_acc: 0.7806 - val_auc: 0.8703

Epoch 00032: val_acc did not improve from 0.81750
Epoch 33/50
300/300 [=====] - 2472s 8s/step - loss: 1.0834 - acc: 0.8135 - auc: 0.8829 - val_loss: 1.1542 - val_acc: 0.7625 - val_auc: 0.8645

Epoch 00033: val_acc did not improve from 0.81750

Epoch 00033: ReduceLROnPlateau reducing learning rate to 8.00000009348878e-08.
Epoch 34/50
300/300 [=====] - 2480s 8s/step - loss: 1.0882 - acc: 0.8025 - auc: 0.8780 - val_loss: 1.0896 - val_acc: 0.7850 - val_auc: 0.8763

Epoch 00034: val_acc did not improve from 0.81750
Epoch 35/50
300/300 [=====] - 2484s 8s/step - loss: 1.0825 - acc: 0.8031 - auc: 0.8778 - val_loss: 1.3321 - val_acc: 0.7850 - val_auc: 0.8791

Epoch 00035: val_acc did not improve from 0.81750
Epoch 36/50
300/300 [=====] - 2493s 8s/step - loss: 1.0744 - acc: 0.8090 - auc: 0.8836 - val_loss: 1.5253 - val_acc: 0.7931 - val_auc: 0.8813

Epoch 00036: val_acc did not improve from 0.81750
Epoch 37/50
300/300 [=====] - 2450s 8s/step - loss: 1.0814 - acc: 0.8035 - auc: 0.8772 - val_loss: 0.9447 - val_acc: 0.7931 - val_auc: 0.8822

```
al_loss: 0.9447 - val_acc: 0.7991 - val_auc: 0.8822
```

Epoch 00037: val_acc did not improve from 0.81750

Epoch 38/50

```
300/300 [=====] - 2472s 8s/step - loss: 1.0754 - acc: 0.8044 - auc: 0.8795 - v  
al_loss: 1.3675 - val_acc: 0.7919 - val_auc: 0.8808
```

Epoch 00038: val_acc did not improve from 0.81750

Epoch 39/50

```
300/300 [=====] - 2458s 8s/step - loss: 1.0764 - acc: 0.8104 - auc: 0.8824 - v  
al_loss: 1.6520 - val_acc: 0.7788 - val_auc: 0.8748
```

Epoch 00039: val_acc did not improve from 0.81750

Epoch 40/50

```
300/300 [=====] - 2454s 8s/step - loss: 1.0646 - acc: 0.8106 - auc: 0.8910 - v  
al_loss: 1.0561 - val_acc: 0.7919 - val_auc: 0.8830
```

Epoch 00040: val_acc did not improve from 0.81750

Epoch 41/50

```
300/300 [=====] - 2452s 8s/step - loss: 1.0790 - acc: 0.8158 - auc: 0.8863 - v  
al_loss: 1.0373 - val_acc: 0.7925 - val_auc: 0.8734
```

Epoch 00041: val_acc did not improve from 0.81750

Epoch 42/50

```
300/300 [=====] - 2477s 8s/step - loss: 1.0599 - acc: 0.8179 - auc: 0.8925 - v  
al_loss: 1.1467 - val_acc: 0.7869 - val_auc: 0.8922
```

Epoch 00042: val_acc did not improve from 0.81750

Epoch 43/50

```
300/300 [=====] - 2468s 8s/step - loss: 1.0724 - acc: 0.8092 - auc: 0.8825 - v  
al_loss: 0.9809 - val_acc: 0.7894 - val_auc: 0.8811
```

Epoch 00043: val_acc did not improve from 0.81750

Epoch 44/50

```
300/300 [=====] - 2443s 8s/step - loss: 1.0700 - acc: 0.8092 - auc: 0.8867 - v  
al_loss: 1.1400 - val_acc: 0.8019 - val_auc: 0.8883
```

Epoch 00044: val_acc did not improve from 0.81750

Epoch 45/50

```
300/300 [=====] - 2421s 8s/step - loss: 1.0757 - acc: 0.8081 - auc: 0.8815 - v  
al_loss: 1.8584 - val_acc: 0.7912 - val_auc: 0.8827
```

Epoch 00045: val_acc did not improve from 0.81750

Epoch 46/50

```
300/300 [=====] - 2453s 8s/step - loss: 1.0729 - acc: 0.8046 - auc: 0.8837 - v  
al_loss: 1.3261 - val_acc: 0.7919 - val_auc: 0.8823
```

Epoch 00046: val_acc did not improve from 0.81750

Epoch 47/50

```
300/300 [=====] - 2411s 8s/step - loss: 1.0605 - acc: 0.8202 - auc: 0.8899 - v  
al_loss: 0.9758 - val_acc: 0.7663 - val_auc: 0.8692
```

Epoch 00047: val_acc did not improve from 0.81750

Epoch 48/50

```
300/300 [=====] - 2354s 8s/step - loss: 1.0503 - acc: 0.8221 - auc: 0.8948 - v  
al_loss: 1.0674 - val_acc: 0.7875 - val_auc: 0.8794
```

Epoch 00048: val_acc did not improve from 0.81750

Epoch 49/50

```
300/300 [=====] - 2396s 8s/step - loss: 1.0632 - acc: 0.8179 - auc: 0.8892 - v  
al_loss: 1.1514 - val_acc: 0.7763 - val_auc: 0.8628
```

Epoch 00049: val_acc did not improve from 0.81750

Epoch 50/50

```
300/300 [=====] - 2438s 8s/step - loss: 1.0701 - acc: 0.8129 - auc: 0.8831 - v  
al_loss: 1.1050 - val_acc: 0.7750 - val_auc: 0.8680
```

Epoch 00050: val_acc did not improve from 0.81750

Out[49]:

```
<keras.callbacks.callbacks.History at 0x7f5a27beca20>
```

In [51]:

```
!mv ./model5_1.h5 ./drive/My\ Drive/COLAB/SIMILAR_FACES/
```

In [53]:

```
!ls

driver_installer.run      sample_submission.csv

model5_1.h5              SAVE_MODEL

model5.h5                TENSORBOARD

Northeastern_GCP_submit2_1_final_1.ipynb  test

Northeastern_GCP_submit2_1_test.ipynb     train

recognizing-faces-in-the-wild  train_relationships.csv

recognizing-faces-in-the-wild.zip
```

In [0]:

In [0]:

In [0]:

In [0]:

MODEL 6

In [0]:

```
#https://www.kaggle.com/janpreets/just-another-feature-extractor-0-824-lb
```

In [34]:

```
def initialize_bias(shape, name=None):
    """
    The paper, http://www.cs.utoronto.ca/~gkoch/files/msc-thesis.pdf
    suggests to initialize CNN layer bias with mean as 0.5 and standard deviation of 0.01
    """
    return np.random.normal(loc = 0.5, scale = 1e-2, size = shape)

def initialize_weights(shape, name=None):
    """
    The paper, http://www.cs.utoronto.ca/~gkoch/files/msc-thesis.pdf
    suggests to initialize CNN layer weights with mean as 0.0 and standard deviation of 0.01
    """
    return np.random.normal(loc = 0.0, scale = 1e-2, size = shape)
```

In [0]:

```
from keras.layers import Conv2D, ZeroPadding2D, Activation, Input, concatenate
from keras.layers.core import Lambda, Flatten, Dense
from keras.layers.normalization import BatchNormalization
from keras.layers.pooling import MaxPooling2D, AveragePooling2D
from keras.models import Model
```

```

from keras.models import Model
from keras import backend as K
import pandas as pd
import numpy as np
import keras
from keras.models import *
from keras.layers import *
from keras.preprocessing import *
from keras.callbacks import *
from keras.optimizers import *
from tqdm import tqdm
import glob
from keras.applications.resnet50 import preprocess_input, decode_predictions

# from keras.applications.nasnet import preprocess_input, decode_predictions
# from keras.applications.densenet import preprocess_input, decode_predictions
from keras.preprocessing import image
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt
import sys

import cv2
from random import choice, sample
import datetime
from sklearn.metrics import roc_auc_score
from keras.callbacks import ReduceLROnPlateau, ModelCheckpoint, EarlyStopping
import threading

from keras.initializers import glorot_normal

import tensorflow as tf
import numpy as np
import os

from numpy import genfromtxt
from keras.layers import Conv2D, ZeroPadding2D, Activation
from keras.layers.normalization import BatchNormalization

```

In []:

```

import tensorflow as tf
from tensorflow.python.keras import backend as K
sess = K.get_session()
#array = sess.run(your_tensor)

```

In []:

```

# https://github.com/rcmalli/keras-vggface/blob/master/keras_vggface/utils.py
def preprocess_input(x, data_format=None, version=1):
    """
    preprocess_input will take the numpy array of image and remove the unnecessary values
    The out ut numpy only contains thermal image of face which will have the umporatant chracters
    Here we subtract the values from image numpy array after subtraction redundant pixels are removed
    """
    x_temp = np.copy(x)
    if data_format is None:
        data_format = K.image_data_format()
    assert data_format in {'channels_last', 'channels_first'}

    if version == 1:
        if data_format == 'channels_first':
            x_temp = x_temp[:, ::-1, ...]
            x_temp[:, 0, :, :] -= 93.5940
            x_temp[:, 1, :, :] -= 104.7624
            x_temp[:, 2, :, :] -= 129.1863
        else:
            x_temp = x_temp[..., ::-1]
            x_temp[..., 0] -= 93.5940
            x_temp[..., 1] -= 104.7624
            x_temp[..., 2] -= 129.1863

    elif version == 2:

```

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    if data_format == 'channels_first':
        x_temp = x_temp[:, ::-1, ...]
        x_temp[:, 0, :, :] -= 91.4953
        x_temp[:, 1, :, :] -= 103.8827
        x_temp[:, 2, :, :] -= 131.0912
    else:
        x_temp = x_temp[..., ::-1]
        x_temp[..., 0] -= 91.4953
        x_temp[..., 1] -= 103.8827
        x_temp[..., 2] -= 131.0912
else:
    raise NotImplementedError
return x_temp

```

In []:

```

# Function for reading images
def read_img(path, IMG_SIZE):
    img = image.load_img(path, target_size=(IMG_SIZE, IMG_SIZE))
    img = np.array(img).astype(np.float)
    return preprocess_input(img, version=2)

def gen(list_tuples, person_to_images_map, batch_size=16):
    ppl = list(person_to_images_map.keys())
    while True:
        batch_tuples = sample(list_tuples, batch_size // 2)
        labels = [1] * len(batch_tuples)
        while len(batch_tuples) < batch_size:
            p1 = choice(ppl)
            p2 = choice(ppl)
            if p1 != p2 and (p1, p2) not in list_tuples and (p2, p1) not in list_tuples:
                batch_tuples.append((p1, p2))
                labels.append(0)
        for x in batch_tuples:
            if not len(person_to_images_map[x[0]]):
                print(x[0])
            X1 = [choice(person_to_images_map[x[0]]) for x in batch_tuples]
            X1 = np.array([read_img(x, 96) for x in X1])
            X2 = [choice(person_to_images_map[x[1]]) for x in batch_tuples]
            X2 = np.array([read_img(x, 96) for x in X2])
            yield [X1, X2], labels

```

In [0]:

```

def inception_based_model():
    myInput = Input(shape=(96, 96, 3))

    x = ZeroPadding2D(padding=(3, 3), input_shape=(96, 96, 3))(myInput)
    x = Conv2D(64, (7, 7), strides=(2, 2), name='conv1')(x)
    x = BatchNormalization(axis=3, epsilon=0.00001, name='bn1')(x)
    x = Activation('relu')(x)
    x = ZeroPadding2D(padding=(1, 1))(x)
    x = MaxPooling2D(pool_size=3, strides=2)(x)
    x = Lambda(LRN2D, name='lrn_1')(x)
    x = Conv2D(64, (1, 1), name='conv2')(x)
    x = BatchNormalization(axis=3, epsilon=0.00001, name='bn2')(x)
    x = Activation('relu')(x)
    x = ZeroPadding2D(padding=(1, 1))(x)
    x = Conv2D(192, (3, 3), name='conv3')(x)
    x = BatchNormalization(axis=3, epsilon=0.00001, name='bn3')(x)
    x = Activation('relu')(x)
    x = Lambda(LRN2D, name='lrn_2')(x)
    x = ZeroPadding2D(padding=(1, 1))(x)
    x = MaxPooling2D(pool_size=3, strides=2)(x)

    # Inception3a
    inception_3a_3x3 = Conv2D(96, (1, 1), name='inception_3a_3x3_conv1')(x)
    inception_3a_3x3 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_3x3_bn1')(inception_3a_3x3)
    inception_3a_3x3 = Activation('relu')(inception_3a_3x3)
    inception_3a_3x3 = ZeroPadding2D(padding=(1, 1))(inception_3a_3x3)
    inception_3a_3x3 = Conv2D(128, (3, 3), name='inception_3a_3x3_conv2')(inception_3a_3x3)
    inception_3a_3x3 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_3x3_bn2')(inception_3a_3x3)

```

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inception_3a_3x3 = Activation('relu')(inception_3a_3x3)

inception_3a_5x5 = Conv2D(16, (1, 1), name='inception_3a_5x5_conv1')(x)
inception_3a_5x5 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_5x5_bn1')(inception_3a_5x5)
inception_3a_5x5 = Activation('relu')(inception_3a_5x5)
inception_3a_5x5 = ZeroPadding2D(padding=(2, 2))(inception_3a_5x5)
inception_3a_5x5 = Conv2D(32, (5, 5), name='inception_3a_5x5_conv2')(inception_3a_5x5)
inception_3a_5x5 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_5x5_bn2')(inception_3a_5x5)
inception_3a_5x5 = Activation('relu')(inception_3a_5x5)

inception_3a_pool = MaxPooling2D(pool_size=3, strides=2)(x)
inception_3a_pool = Conv2D(32, (1, 1), name='inception_3a_pool_conv')(inception_3a_pool)
inception_3a_pool = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_pool_bn')(inception_3a_pool)
inception_3a_pool = Activation('relu')(inception_3a_pool)
inception_3a_pool = ZeroPadding2D(padding=((3, 4), (3, 4)))(inception_3a_pool)

inception_3a_1x1 = Conv2D(64, (1, 1), name='inception_3a_1x1_conv')(x)
inception_3a_1x1 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_1x1_bn')(inception_3a_1x1)
inception_3a_1x1 = Activation('relu')(inception_3a_1x1)

inception_3a = concatenate([inception_3a_3x3, inception_3a_5x5, inception_3a_pool, inception_3a_1x1], axis=3)

# Inception3b
inception_3b_3x3 = Conv2D(96, (1, 1), name='inception_3b_3x3_conv1')(inception_3a)
inception_3b_3x3 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_3x3_bn1')(inception_3b_3x3)
inception_3b_3x3 = Activation('relu')(inception_3b_3x3)
inception_3b_3x3 = ZeroPadding2D(padding=(1, 1))(inception_3b_3x3)
inception_3b_3x3 = Conv2D(128, (3, 3), name='inception_3b_3x3_conv2')(inception_3b_3x3)
inception_3b_3x3 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_3x3_bn2')(inception_3b_3x3)
inception_3b_3x3 = Activation('relu')(inception_3b_3x3)

inception_3b_5x5 = Conv2D(32, (1, 1), name='inception_3b_5x5_conv1')(inception_3a)
inception_3b_5x5 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_5x5_bn1')(inception_3b_5x5)
inception_3b_5x5 = Activation('relu')(inception_3b_5x5)
inception_3b_5x5 = ZeroPadding2D(padding=(2, 2))(inception_3b_5x5)
inception_3b_5x5 = Conv2D(64, (5, 5), name='inception_3b_5x5_conv2')(inception_3b_5x5)
inception_3b_5x5 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_5x5_bn2')(inception_3b_5x5)
inception_3b_5x5 = Activation('relu')(inception_3b_5x5)

inception_3b_pool = AveragePooling2D(pool_size=(3, 3), strides=(3, 3))(inception_3a)
inception_3b_pool = Conv2D(64, (1, 1), name='inception_3b_pool_conv')(inception_3b_pool)
inception_3b_pool = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_pool_bn')(inception_3b_pool)
inception_3b_pool = Activation('relu')(inception_3b_pool)
inception_3b_pool = ZeroPadding2D(padding=(4, 4))(inception_3b_pool)

inception_3b_1x1 = Conv2D(64, (1, 1), name='inception_3b_1x1_conv')(inception_3a)
inception_3b_1x1 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_1x1_bn')(inception_3b_1x1)
inception_3b_1x1 = Activation('relu')(inception_3b_1x1)

inception_3b = concatenate([inception_3b_3x3, inception_3b_5x5, inception_3b_pool, inception_3b_1x1], axis=3)

# Inception3c
inception_3c_3x3 = conv2d_bn(inception_3b,
                             layer='inception_3c_3x3',
                             cv1_out=128,
                             cv1_filter=(1, 1),
                             cv2_out=256,
                             cv2_filter=(3, 3),
                             cv2_strides=(2, 2),
                             padding=(1, 1))

inception_3c_5x5 = conv2d_bn(inception_3b,
                             layer='inception_3c_5x5',
                             cv1_out=32,
                             cv1_filter=(1, 1),

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        cv2_out=64,
        cv2_filter=(5, 5),
        cv2_strides=(2, 2),
        padding=(2, 2))

inception_3c_pool = MaxPooling2D(pool_size=3, strides=2)(inception_3b)
inception_3c_pool = ZeroPadding2D(padding=((0, 1), (0, 1)))(inception_3c_pool)

inception_3c = concatenate([inception_3c_3x3, inception_3c_5x5, inception_3c_pool], axis=3)

#inception 4a
inception_4a_3x3 = conv2d_bn(inception_3c,
                             layer='inception_4a_3x3',
                             cv1_out=96,
                             cv1_filter=(1, 1),
                             cv2_out=192,
                             cv2_filter=(3, 3),
                             cv2_strides=(1, 1),
                             padding=(1, 1))
inception_4a_5x5 = conv2d_bn(inception_3c,
                             layer='inception_4a_5x5',
                             cv1_out=32,
                             cv1_filter=(1, 1),
                             cv2_out=64,
                             cv2_filter=(5, 5),
                             cv2_strides=(1, 1),
                             padding=(2, 2))

inception_4a_pool = AveragePooling2D(pool_size=(3, 3), strides=(3, 3))(inception_3c)
inception_4a_pool = conv2d_bn(inception_4a_pool,
                              layer='inception_4a_pool',
                              cv1_out=128,
                              cv1_filter=(1, 1),
                              padding=(2, 2))
inception_4a_1x1 = conv2d_bn(inception_3c,
                              layer='inception_4a_1x1',
                              cv1_out=256,
                              cv1_filter=(1, 1))

inception_4a = concatenate([inception_4a_3x3, inception_4a_5x5, inception_4a_pool, inception_4a_1x1
], axis=3)

#inception4e
inception_4e_3x3 = conv2d_bn(inception_4a,
                             layer='inception_4e_3x3',
                             cv1_out=160,
                             cv1_filter=(1, 1),
                             cv2_out=256,
                             cv2_filter=(3, 3),
                             cv2_strides=(2, 2),
                             padding=(1, 1))
inception_4e_5x5 = conv2d_bn(inception_4a,
                             layer='inception_4e_5x5',
                             cv1_out=64,
                             cv1_filter=(1, 1),
                             cv2_out=128,
                             cv2_filter=(5, 5),
                             cv2_strides=(2, 2),
                             padding=(2, 2))
inception_4e_pool = MaxPooling2D(pool_size=3, strides=2)(inception_4a)
inception_4e_pool = ZeroPadding2D(padding=((0, 1), (0, 1)))(inception_4e_pool)

inception_4e = concatenate([inception_4e_3x3, inception_4e_5x5, inception_4e_pool], axis=3)

#inception5a
inception_5a_3x3 = conv2d_bn(inception_4e,
                             layer='inception_5a_3x3',
                             cv1_out=96,
                             cv1_filter=(1, 1),
                             cv2_out=384,
                             cv2_filter=(3, 3),
                             cv2_strides=(1, 1),
                             padding=(1, 1))

inception_5a_pool = AveragePooling2D(pool_size=(3, 3), strides=(3, 3))(inception_4e)
inception_5a_pool = conv2d_bn(inception_5a_pool,
                              layer='inception_5a_pool',
                              cv1_out=96,

```

```

cv1_out=96,
cv1_filter=(1, 1),
padding=(1, 1))
inception_5a_1x1 = conv2d_bn(inception_4e,
                             layer='inception_5a_1x1',
                             cv1_out=256,
                             cv1_filter=(1, 1))

inception_5a = concatenate([inception_5a_3x3, inception_5a_pool, inception_5a_1x1], axis=3)

#inception 5b
inception_5b_3x3 = conv2d_bn(inception_5a,
                              layer='inception_5b_3x3',
                              cv1_out=96,
                              cv1_filter=(1, 1),
                              cv2_out=384,
                              cv2_filter=(3, 3),
                              cv2_strides=(1, 1),
                              padding=(1, 1))
inception_5b_pool = MaxPooling2D(pool_size=3, strides=2)(inception_5a)
inception_5b_pool = conv2d_bn(inception_5b_pool,
                              layer='inception_5b_pool',
                              cv1_out=96,
                              cv1_filter=(1, 1))
inception_5b_pool = ZeroPadding2D(padding=(1, 1))(inception_5b_pool)

inception_5b_1x1 = conv2d_bn(inception_5a,
                              layer='inception_5b_1x1',
                              cv1_out=256,
                              cv1_filter=(1, 1))
inception_5b = concatenate([inception_5b_3x3, inception_5b_pool, inception_5b_1x1], axis=3)

av_pool = AveragePooling2D(pool_size=(3, 3), strides=(1, 1))(inception_5b)
reshape_layer = Flatten()(av_pool)
dense_layer = Dense(128, name='dense_layer')(reshape_layer)
norm_layer = Lambda(lambda x: K.l2_normalize(x, axis=1), name='norm_layer')(dense_layer)

return Model(inputs=[myInput], outputs=norm_layer)

```

In [0]:

```

def LRN2D(x):
    return tf.nn.lrn(x, alpha=1e-4, beta=0.75)

def conv2d_bn(
    x,
    layer=None,
    cv1_out=None,
    cv1_filter=(1, 1),
    cv1_strides=(1, 1),
    cv2_out=None,
    cv2_filter=(3, 3),
    cv2_strides=(1, 1),
    padding=None,
):
    num = '1' if cv2_out == None else '2'
    tensor = Conv2D(cv1_out, cv1_filter, strides=cv1_strides, name=layer+'conv'+num)(x)
    tensor = BatchNormalization(axis=3, epsilon=0.00001, name=layer+'bn'+num)(tensor)
    tensor = Activation('relu')(tensor)
    if padding == None:
        return tensor
    tensor = ZeroPadding2D(padding=padding)(tensor)
    if cv2_out == None:
        return tensor
    tensor = Conv2D(cv2_out, cv2_filter, strides=cv2_strides, name=layer+'conv'+num)(tensor)
    tensor = BatchNormalization(axis=3, epsilon=0.00001, name=layer+'bn'+num)(tensor)
    tensor = Activation('relu')(tensor)
    return tensor

def create_base_network(input_shape):
    mod = inception_base_model()
    new_mod = Model(mod.input, mod.layers[-5].output)
    return new_mod

def get_model6():
    input_shape = (96, 96, 3)

```



```

input_shape = (50, 50, 3)
input_a = Input(shape=input_shape)
input_b = Input(shape=input_shape)
base_network = create_base_network(input_shape)
# because we re-use the same instance 'base_network',
# the weights of the network
# will be shared across the two branches
processed_a = base_network(input_a)
processed_b = base_network(input_b)
# extract features from detector
x_detector = processed_a
shape_detector = processed_a.shape
# extract features from extractor , same with detector for symmetry DxD model
shape_extractor = processed_b.shape
x_extractor = processed_b
# reshape to (minibatch_size, total_pixels, filter_size)
x_detector = keras.layers.Reshape([shape_detector[1] * shape_detector[2] , shape_detector[-1]])(x_detector)
x_extractor = keras.layers.Reshape([shape_extractor[1] * shape_extractor[2] , shape_extractor[-1]])(x_extractor)
# outer products of features, output shape=(minibatch_size, filter_size_detector*filter_size_extractor)
x = keras.layers.Lambda(outer_product)([x_detector, x_extractor])
# reshape to (minibatch_size, filter_size_detector*filter_size_extractor)
x = keras.layers.Reshape([shape_detector[-1]*shape_extractor[-1]])(x)
# signed square-root
x = keras.layers.Lambda(signed_sqrt)(x)
# L2 normalization
x = keras.layers.Lambda(L2_norm)(x)
###
### attach FC-Layer
###
x = Dense(100, activation="relu")(x)
x = Dropout(0.01)(x)
#out = Dense(units=1, kernel_regularizer=keras.regularizers.l2(1e-8), kernel_initializer='glorot_normal', activation="sigmoid")(x)
out = Dense(units=1, kernel_initializer='glorot_normal', activation="sigmoid")(x)
model = Model([input_a, input_b], out)
model.compile(loss="binary_crossentropy", metrics=['acc', auc], optimizer=Adam(0.001))
print(model.summary())
return model

```

In [0]:

In [0]:

```

file_path6 = "./model6.h5"

callbacks_list6 = [keras.callbacks.EarlyStopping(monitor='val_loss', patience=50, verbose=1, min_delta=1e-4),
                  keras.callbacks.ReduceLROnPlateau(monitor='val_loss', factor=0.1, patience=20, verbose=1, epsilon=1e-4),
                  keras.callbacks.ModelCheckpoint(file_path6, save_best_only=True, verbose=1, save_weights_only=True, monitor='val_auc', mode='max'),
                  ]

#callbacks_list6 = [checkpoint6, reduce_on_plateau6, early_stopping6]
#callbacks_list6 = [checkpoint6]

model6 = get_model6()

```

In [0]:

```

model6_hist = model6.fit_generator(gen(train, train_person_to_images_map, batch_size=16), use_multiprocessing=True,
                                  validation_data=gen(val, val_person_to_images_map, batch_size=16), epochs=150, verbose=1,
                                  workers = 16, callbacks=callbacks_list6, steps_per_epoch=300, validation_steps=200)

```

```

Epoch 134/150
300/300 [=====] - 55s 182ms/step - loss: 0.4515 - acc: 0.7856 - auc: 0.8625 - val_loss: 0.5869 - val_acc: 0.7006 - val_auc: 0.7702
300/300 [=====] - 55s 182ms/step - loss: 0.4515 - acc: 0.7856 - auc: 0.8625 - val_loss: 0.5869 - val_acc: 0.7006 - val_auc: 0.7702

Epoch 00134: val_auc improved from 0.76961 to 0.77016, saving model to ./model6.h5

Epoch 00134: val_auc improved from 0.76961 to 0.77016, saving model to ./model6.h5
Epoch 135/150
 1/300 [.....] - ETA: 47s - loss: 0.5908 - acc: 0.7500 - auc: 0.8281Epoch 135/150
300/300 [=====] - 51s 172ms/step - loss: 0.4592 - acc: 0.7800 - auc: 0.8529 - val_loss: 0.5900 - val_acc: 0.6947 - val_auc: 0.7600
300/300 [=====] - 51s 172ms/step - loss: 0.4592 - acc: 0.7800 - auc: 0.8529 - val_loss: 0.5900 - val_acc: 0.6947 - val_auc: 0.7600

Epoch 00135: val_auc did not improve from 0.77016
Epoch 136/150

Epoch 00135: val_auc did not improve from 0.77016
Epoch 136/150
300/300 [=====] - 52s 172ms/step - loss: 0.4572 - acc: 0.7779 - auc: 0.8541 - val_loss: 0.5977 - val_acc: 0.6934 - val_auc: 0.7534
300/300 [=====] - 52s 172ms/step - loss: 0.4572 - acc: 0.7779 - auc: 0.8541 - val_loss: 0.5977 - val_acc: 0.6934 - val_auc: 0.7534

Epoch 00136: val_auc did not improve from 0.77016
Epoch 137/150
 1/300 [.....] - ETA: 56s - loss: 0.5061 - acc: 0.7500 - auc: 0.8594
Epoch 00136: val_auc did not improve from 0.77016
Epoch 137/150
300/300 [=====] - 52s 174ms/step - loss: 0.4610 - acc: 0.7765 - auc: 0.8529 - val_loss: 0.5922 - val_acc: 0.6913 - val_auc: 0.7585
300/300 [=====] - 52s 174ms/step - loss: 0.4610 - acc: 0.7765 - auc: 0.8529 - val_loss: 0.5922 - val_acc: 0.6913 - val_auc: 0.7585

Epoch 00137: val_auc did not improve from 0.77016
Epoch 138/150

Epoch 00137: val_auc did not improve from 0.77016
Epoch 138/150
300/300 [=====] - 52s 175ms/step - loss: 0.4494 - acc: 0.7790 - auc: 0.8625 - val_loss: 0.6292 - val_acc: 0.6684 - val_auc: 0.7419
300/300 [=====] - 52s 175ms/step - loss: 0.4494 - acc: 0.7790 - auc: 0.8625 - val_loss: 0.6292 - val_acc: 0.6684 - val_auc: 0.7419

Epoch 00138: val_auc did not improve from 0.77016
Epoch 00138: early stopping

Epoch 00138: val_auc did not improve from 0.77016
Epoch 00138: early stopping

```

In [0]:

Ensembling with best models

In [7]:

```

from collections import defaultdict
from glob import glob
from random import choice, sample

#import cv2
import numpy as np
import pandas as pd
from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau
from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Concatenate, Multiply
, Dropout, Subtract, Add, Conv2D
from keras.models import Model
from keras.preprocessing import image
from keras.optimizers import Adam
from keras_vggface.utils import preprocess_input
from keras_vggface.vggface import VGGFace

# import regularizer
from keras.regularizers import l1
from keras.regularizers import l2

import h5py
import itertools

```

In [8]:

```

from collections import defaultdict
from glob import glob
from random import choice, sample
import matplotlib.pyplot as plt
from tqdm import tqdm
import numpy as np
import pandas as pd
import pickle

```

```

import gc, psutil, random, base64, h5py, os,shutil
#import cv2
pd.options.mode.chained_assignment = None
pd.options.display.max_columns = 9999
pd.options.display.float_format = '{:20, .2f}'.format
from keras import backend as K

from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Concatenate, Multiply
, Dropout, Subtract, Add
from keras_vggface.vggface import VGGFace
from pathlib import Path

from PIL import Image
from io import BytesIO
from IPython.display import HTML
import plotly.offline as py
py.init_notebook_mode(connected=True)
import plotly.graph_objs as go
import plotly.tools as tls

import warnings
warnings.filterwarnings('ignore')
from keras.layers import Conv2D, ZeroPadding2D, Activation, concatenate
from keras.layers.pooling import MaxPooling2D, AveragePooling2D
from keras.layers.core import Lambda
import keras
from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau, EarlyStopping
from keras.layers import Conv1D, BatchNormalization, Reshape, LeakyReLU
from keras.preprocessing import image
from keras.optimizers import Adam, SGD, rmsprop
from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Conv2D, MaxPooling1D, GlobalAveragePooling1D, LSTM
from keras.models import Model
from keras.losses import binary_crossentropy
import tensorflow as tf
from sklearn.metrics import roc_auc_score
from keras import regularizers
%pylab inline
import matplotlib.image as mpimg

```

Populating the interactive namespace from numpy and matplotlib

In [9]:

```

def auc(y_true, y_pred):
    return tf.py_function(roc_auc_score, (y_true, y_pred), tf.double)

```

In [10]:

```

def gen(list_tuples, person_to_images_map, batch_size=16):
    ppl = list(person_to_images_map.keys())
    while True:
        batch_tuples = sample(list_tuples, batch_size // 2)
        labels = [1] * len(batch_tuples)
        while len(batch_tuples) < batch_size:
            p1 = choice(ppl)
            p2 = choice(ppl)

            if p1 != p2 and (p1, p2) not in list_tuples and (p2, p1) not in list_tuples:
                batch_tuples.append((p1, p2))
                labels.append(0)

        for x in batch_tuples:
            if not len(person_to_images_map[x[0]]):
                print(x[0])

        X1 = [choice(person_to_images_map[x[0]]) for x in batch_tuples]
        X1 = np.array([read_img(x, 197) for x in X1])

        X2 = [choice(person_to_images_map[x[1]]) for x in batch_tuples]
        X2 = np.array([read_img(x, 197) for x in X2])

```

```
yield [X1, X2], labels
```

MODEL 1

In [11]:

```
# Preparing baseline_model1
def baseline_model1():
    input_1 = Input(shape=(197, 197, 3))
    input_2 = Input(shape=(197, 197, 3))

    base_model1 = VGGFace(model='resnet50', include_top=False)

    for x in base_model1.layers[:-3]:
        x.trainable = True
    # for x in base_model1.layers[-3:]:
    #     x.trainable=False

    x1 = base_model1(input_1)
    x2 = base_model1(input_2)

    x1 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x1), GlobalAvgPool2D() (x1)])
    x2 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x2), GlobalAvgPool2D() (x2)])

    x3 = Subtract() ([x1, x2])
    x3 = Multiply() ([x3, x3])

    x1_ = Multiply() ([x1, x1])
    x2_ = Multiply() ([x2, x2])
    x4 = Subtract() ([x1_, x2_])
    x = Concatenate(axis=-1) ([x4, x3])

    x = Dense(100, activation="relu", activity_regularizer=l1(0.001)) (x)
    x = Dropout(0.3) (x)
    x = Dense(25, activation="relu", activity_regularizer=l1(0.001)) (x)
    x = Dropout(0.01) (x)
    out = Dense(1, activation="sigmoid") (x)

    model1 = Model([input_1, input_2], out)

    model1.compile(loss="binary_crossentropy", metrics=['acc'], optimizer=Adam(0.00001))

    model1.summary()

    return model1
```

In [12]:

```
model1 = baseline_model1()
```

Model: "model_1"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 197, 197, 3)	0	
input_2 (InputLayer)	(None, 197, 197, 3)	0	
vggface_resnet50 (Model)	multiple	23561152	input_1[0][0] input_2[0][0]
global_max_pooling2d_1 (GlobalM (None, 2048)		0	vggface_resnet50[1][0]
global_average_pooling2d_1 (Glo (None, 2048)		0	vggface_resnet50[1][0]
global_max_pooling2d_2 (GlobalM (None, 2048)		0	vggface_resnet50[2][0]
global_average_pooling2d_2 (Glo (None, 2048)		0	vggface_resnet50[2][0]
concatenate_1 (Concatenate)	(None, 4096)	0	global_max_pooling2d_1[0][0] global_average_pooling2d_1[0][0]

concatenate_2 (Concatenate)	(None, 4096)	0	global_max_pooling2d_2[0][0] global_average_pooling2d_2[0][0]
multiply_2 (Multiply)	(None, 4096)	0	concatenate_1[0][0] concatenate_1[0][0]
multiply_3 (Multiply)	(None, 4096)	0	concatenate_2[0][0] concatenate_2[0][0]
subtract_1 (Subtract)	(None, 4096)	0	concatenate_1[0][0] concatenate_2[0][0]
subtract_2 (Subtract)	(None, 4096)	0	multiply_2[0][0] multiply_3[0][0]
multiply_1 (Multiply)	(None, 4096)	0	subtract_1[0][0] subtract_1[0][0]
concatenate_3 (Concatenate)	(None, 8192)	0	subtract_2[0][0] multiply_1[0][0]
dense_1 (Dense)	(None, 100)	819300	concatenate_3[0][0]
dropout_1 (Dropout)	(None, 100)	0	dense_1[0][0]
dense_2 (Dense)	(None, 25)	2525	dropout_1[0][0]
dropout_2 (Dropout)	(None, 25)	0	dense_2[0][0]
dense_3 (Dense)	(None, 1)	26	dropout_2[0][0]
=====			
Total params: 24,383,003			
Trainable params: 24,329,883			
Non-trainable params: 53,120			

In []:

In []:

MODEL 2

In []:

In [13]:

```
def baseline_model2():
    input_1 = Input(shape=(197, 197, 3))
    input_2 = Input(shape=(197, 197, 3))

    base_model2 = VGGFace(model='resnet50', include_top=False)

    for x in base_model2.layers[:-3]:
        x.trainable = True

    x1 = base_model2(input_1)
    x2 = base_model2(input_2)

    x1 = Concatenate(axis=-1)([GlobalMaxPool2D()(x1), GlobalAvgPool2D()(x1)])
    x2 = Concatenate(axis=-1)([GlobalMaxPool2D()(x2), GlobalAvgPool2D()(x2)])

    x3 = Subtract()([x1, x2])
    x3 = Multiply()([x3, x3])

    x1 = Multiply()([x1, x1])
```

```

x2_ = Multiply()([x2, x2])
x4 = Subtract()([x1_, x2_])
x = Concatenate(axis=-1)([x4, x3])
x = Dropout(0.3)(x)
x = Dense(100, activation="relu", activity_regularizer=l1(0.001))(x)
x = Dropout(0.1)(x)
out = Dense(1, activation="sigmoid")(x)

model2 = Model([input_1, input_2], out)

model2.compile(loss="binary_crossentropy", metrics=['acc', auc], optimizer=Adam(0.00001))

model2.summary()

return model2

```

In [14]:

```

model2 = baseline_model2()
#model.load_weights(file_path)

```

Model: "model_2"

Layer (type)	Output Shape	Param #	Connected to
input_4 (InputLayer)	(None, 197, 197, 3)	0	
input_5 (InputLayer)	(None, 197, 197, 3)	0	
vggface_resnet50 (Model)	multiple	23561152	input_4[0][0] input_5[0][0]
global_max_pooling2d_3 (GlobalM	(None, 2048)	0	vggface_resnet50[1][0]
global_average_pooling2d_3 (Glo	(None, 2048)	0	vggface_resnet50[1][0]
global_max_pooling2d_4 (GlobalM	(None, 2048)	0	vggface_resnet50[2][0]
global_average_pooling2d_4 (Glo	(None, 2048)	0	vggface_resnet50[2][0]
concatenate_4 (Concatenate)	(None, 4096)	0	global_max_pooling2d_3[0][0] global_average_pooling2d_3[0][0]
concatenate_5 (Concatenate)	(None, 4096)	0	global_max_pooling2d_4[0][0] global_average_pooling2d_4[0][0]
multiply_5 (Multiply)	(None, 4096)	0	concatenate_4[0][0] concatenate_4[0][0]
multiply_6 (Multiply)	(None, 4096)	0	concatenate_5[0][0] concatenate_5[0][0]
subtract_3 (Subtract)	(None, 4096)	0	concatenate_4[0][0] concatenate_5[0][0]
subtract_4 (Subtract)	(None, 4096)	0	multiply_5[0][0] multiply_6[0][0]
multiply_4 (Multiply)	(None, 4096)	0	subtract_3[0][0] subtract_3[0][0]
concatenate_6 (Concatenate)	(None, 8192)	0	subtract_4[0][0] multiply_4[0][0]
dropout_3 (Dropout)	(None, 8192)	0	concatenate_6[0][0]
dense_4 (Dense)	(None, 100)	819300	dropout_3[0][0]
dropout_4 (Dropout)	(None, 100)	0	dense_4[0][0]
dense_5 (Dense)	(None, 1)	101	dropout_4[0][0]

```

Total params: 24,380,553
Trainable params: 24,327,433
Non-trainable params: 53,120

```

```
In [ ]:
```

MODEL 3

```
In [ ]:
```

```
In [15]:
```

```
def baseline_model3_1():
    input_1 = Input(shape=(224, 224, 3))
    input_2 = Input(shape=(224, 224, 3))

    base_model3 = VGGFace(model='resnet50', include_top=False)

    for layer in base_model3.layers[:-3]:
        layer.trainable = True

    x1 = base_model3(input_1)
    x2 = base_model3(input_2)

    merged_add = Add()([x1, x2])
    merged_sub = Subtract()([x1, x2])

    merged_add = Conv2D(100, [1,1])(merged_add)
    merged_sub = Conv2D(100, [1,1])(merged_sub)

    merged = Concatenate(axis=-1)([merged_add, merged_sub])

    merged = Flatten()(merged)
    merged = Dropout(0.2)(merged)
    merged = Dense(100, activation="relu", activity_regularizer=l1(0.001))(merged)
    merged = Dropout(0.2)(merged)
    merged = Dense(25, activation="relu", activity_regularizer=l1(0.001))(merged)
    merged = Dropout(0.2)(merged)
    out = Dense(1, activation="sigmoid")(merged)
    model3 = Model([input_1, input_2], out)
    model3.compile(loss="binary_crossentropy", metrics=['acc', 'auc'], optimizer=Adam(0.00001))
    model3.summary()
    return model3
```

```
In [16]:
```

```
model3 = baseline_model3_1()
```

Model: "model_3"

Layer (type)	Output Shape	Param #	Connected to
input_7 (InputLayer)	(None, 224, 224, 3)	0	
input_8 (InputLayer)	(None, 224, 224, 3)	0	
vggface_resnet50 (Model)	multiple	23561152	input_7[0][0] input_8[0][0]
add_49 (Add)	(None, 1, 1, 2048)	0	vggface_resnet50[1][0] vggface_resnet50[2][0]
subtract_5 (Subtract)	(None, 1, 1, 2048)	0	vggface_resnet50[1][0] vggface_resnet50[2][0]
conv2d_1 (Conv2D)	(None, 1, 1, 100)	204900	add_49[0][0]
conv2d_2 (Conv2D)	(None, 1, 1, 100)	204900	subtract_5[0][0]

concatenate_7 (Concatenate)	(None, 1, 1, 200)	0	conv2d_1[0][0] conv2d_2[0][0]
flatten_1 (Flatten)	(None, 200)	0	concatenate_7[0][0]
dropout_5 (Dropout)	(None, 200)	0	flatten_1[0][0]
dense_6 (Dense)	(None, 100)	20100	dropout_5[0][0]
dropout_6 (Dropout)	(None, 100)	0	dense_6[0][0]
dense_7 (Dense)	(None, 25)	2525	dropout_6[0][0]
dropout_7 (Dropout)	(None, 25)	0	dense_7[0][0]
dense_8 (Dense)	(None, 1)	26	dropout_7[0][0]
=====			
Total params: 23,993,603			
Trainable params: 23,940,483			
Non-trainable params: 53,120			

In []:

In []:

MODEL 4

In []:

In [17]:

```
def baseline_model4():
    input_1 = Input(shape=(224, 224, 3))
    input_2 = Input(shape=(224, 224, 3))

    base_model4 = VGGFace(model='resnet50', include_top=False)

    for x in base_model4.layers[:-3]:
        x.trainable = True

    x1 = base_model4(input_1)
    x2 = base_model4(input_2)

    x1 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x1), GlobalAvgPool2D() (x1)])
    x2 = Concatenate(axis=-1) ([GlobalMaxPool2D() (x2), GlobalAvgPool2D() (x2)])

    x3 = Subtract() ([x1, x2])
    x3 = Multiply() ([x3, x3])

    x = Multiply() ([x1, x2])

    x = Concatenate(axis=-1) ([x, x3])
    x = BatchNormalization() (x)
    x = Dense(128, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)
    x = Activation('relu') (x)
    x = Dropout(0.4) (x)
    x = Dense(64, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)
    x = Activation('relu') (x)
    x = Dropout(0.4) (x)
    x = Dense(32, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)
    x = Activation('relu') (x)
    x = Dropout(0.3) (x)
    x = Dense(16, kernel_initializer='glorot_normal', kernel_regularizer=l2(0.001)) (x)
```



```

x = Activation('relu')(x)
x = Dropout(0.3)(x)
out = Dense(1, activation="sigmoid")(x)
model4 = Model([input_1, input_2], out)
model4.compile(loss="binary_crossentropy", metrics=['acc', auc], optimizer=Adam(0.00001))
model4.summary()
return model4

```

In [18]:

```
model4 = baseline_model4()
```

Model: "model_4"

Layer (type)	Output Shape	Param #	Connected to
input_10 (InputLayer)	(None, 224, 224, 3)	0	
input_11 (InputLayer)	(None, 224, 224, 3)	0	
vggface_resnet50 (Model)	multiple	23561152	input_10[0][0] input_11[0][0]
global_max_pooling2d_5 (GlobalM	(None, 2048)	0	vggface_resnet50[1][0]
global_average_pooling2d_5 (Glo	(None, 2048)	0	vggface_resnet50[1][0]
global_max_pooling2d_6 (GlobalM	(None, 2048)	0	vggface_resnet50[2][0]
global_average_pooling2d_6 (Glo	(None, 2048)	0	vggface_resnet50[2][0]
concatenate_8 (Concatenate)	(None, 4096)	0	global_max_pooling2d_5[0][0] global_average_pooling2d_5[0][0]
concatenate_9 (Concatenate)	(None, 4096)	0	global_max_pooling2d_6[0][0] global_average_pooling2d_6[0][0]
subtract_6 (Subtract)	(None, 4096)	0	concatenate_8[0][0] concatenate_9[0][0]
multiply_8 (Multiply)	(None, 4096)	0	concatenate_8[0][0] concatenate_9[0][0]
multiply_7 (Multiply)	(None, 4096)	0	subtract_6[0][0] subtract_6[0][0]
concatenate_10 (Concatenate)	(None, 8192)	0	multiply_8[0][0] multiply_7[0][0]
batch_normalization_1 (BatchNor	(None, 8192)	32768	concatenate_10[0][0]
dense_9 (Dense)	(None, 128)	1048704	batch_normalization_1[0][0]
activation_197 (Activation)	(None, 128)	0	dense_9[0][0]
dropout_8 (Dropout)	(None, 128)	0	activation_197[0][0]
dense_10 (Dense)	(None, 64)	8256	dropout_8[0][0]
activation_198 (Activation)	(None, 64)	0	dense_10[0][0]
dropout_9 (Dropout)	(None, 64)	0	activation_198[0][0]
dense_11 (Dense)	(None, 32)	2080	dropout_9[0][0]
activation_199 (Activation)	(None, 32)	0	dense_11[0][0]
dropout_10 (Dropout)	(None, 32)	0	activation_199[0][0]
dense_12 (Dense)	(None, 16)	528	dropout_10[0][0]
activation_200 (Activation)	(None, 16)	0	dense_12[0][0]
dropout_11 (Dropout)	(None, 16)	0	activation_200[0][0]
dense_13 (Dense)	(None, 1)	17	dropout_11[0][0]

```
dense_13 (Dense) (None, 1) 1 / dropout_11[0][0]
```

```
Total params: 24,653,505  
Trainable params: 24,584,001  
Non-trainable params: 69,504
```

```
In [ ]:
```

MODEL 5

```
In [ ]:
```

```
In [19]:
```

```
from random import choice, sample
```

```
In [20]:
```

```
from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau, EarlyStopping  
from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, MaxPooling1D, GlobalAveragePooling1D, LSTM  
from keras.layers import Concatenate, Multiply, Dropout, Subtract, Add, Conv2D, Conv1D  
from keras.models import Model  
from keras.layers import BatchNormalization, Reshape  
from keras.preprocessing import image  
from keras.optimizers import Adam, SGD, rmsprop  
import h5py  
from keras.regularizers import l2  
from keras.losses import binary_crossentropy  
import tensorflow as tf  
from sklearn.metrics import roc_auc_score  
from keras.layers import LeakyReLU  
from keras import regularizers  
from keras.initializers import RandomNormal  
  
def auc(y_true, y_pred):  
    return tf.py_function(roc_auc_score, (y_true, y_pred), tf.double)
```

```
In [21]:
```

```
def siamese_model_1():  
    input_1 = Input(shape=(224, 224, 3))  
    input_2 = Input(shape=(224, 224, 3))  
    base_model = VGGFace(model='resnet50', include_top=False, input_shape=(224, 224, 3))  
  
    for x in base_model.layers[:-3]:  
        x.trainable = True  
  
    x1 = base_model(input_1)  
    x2 = base_model(input_2)  
  
    x1 = Concatenate(axis=-1) ([GlobalAvgPool2D()(x1), GlobalAvgPool2D()(x1)])  
    x2 = Concatenate(axis=-1) ([GlobalMaxPool2D()(x2), GlobalAvgPool2D()(x2)])  
    x3 = Subtract() ([x1, x2])  
    x3 = Multiply() ([x3, x3])  
    x1_ = Multiply() ([x1, x1])  
    x2_ = Multiply() ([x2, x2])  
    x4 = Subtract() ([x1_, x2_])  
    x = Concatenate(axis=-1) ([x4, x3])  
    x = Reshape(input_shape=(8192,), target_shape=(8192, 1))(x)  
    x = Conv1D(filters=100,  
               kernel_size=8,  
               strides=8, input_shape=(8192, 1),  
               activation='relu',  
               padding='same')(x)
```

```

padding='same')(x)
x = Conv1D(100,8,activation='relu')(x)
x = LSTM(128,kernel_initializer='glorot_normal',recurrent_dropout=0.5,kernel_regularizer=l2(0.001),
return_sequences=True)(x)
#x = LSTM(128, return_sequences=True)(x)
x = Dropout(0.4)(x)
x = LSTM(128,kernel_initializer='glorot_normal',recurrent_dropout=0.5,kernel_regularizer=l2(0.001),
return_sequences=True)(x)
#x = LSTM(128, return_sequences=True)(x)
x = BatchNormalization()(x)
x = Flatten()(x)
x = Dense(100,kernel_initializer='glorot_normal',kernel_regularizer=l2(0.001))(x)
x = Activation('relu')(x)
x = Dropout(0.3)(x)
x = Dense(64,kernel_initializer='glorot_normal',kernel_regularizer=l2(0.001))(x)
x = Activation('relu')(x)
x = Dropout(0.2)(x)
x = Dense(32,kernel_initializer='glorot_normal',kernel_regularizer=l2(0.001))(x)
x = Activation('relu')(x)
x = Dropout(0.2)(x)
x = Dense(16,kernel_initializer='glorot_normal',kernel_regularizer=l2(0.001))(x)
x = Activation('relu')(x)
x = Dropout(0.1)(x)
out = Dense(1, activation="sigmoid")(x)
model = Model([input_1, input_2], out)
model.compile(loss="binary_crossentropy", metrics=['acc',auc], optimizer=Adam(0.00001,decay=1e-6))
model.summary()
return model

```

In []:

In [22]:

```
model5 = siamese_model_1()
```

Model: "model_5"

Layer (type)	Output Shape	Param #	Connected to
input_13 (InputLayer)	(None, 224, 224, 3)	0	
input_14 (InputLayer)	(None, 224, 224, 3)	0	
vggface_resnet50 (Model)	(None, 1, 1, 2048)	23561152	input_13[0][0] input_14[0][0]
global_average_pooling2d_7 (Glo	(None, 2048)	0	vggface_resnet50[1][0]
global_average_pooling2d_8 (Glo	(None, 2048)	0	vggface_resnet50[1][0]
global_max_pooling2d_7 (GlobalM	(None, 2048)	0	vggface_resnet50[2][0]
global_average_pooling2d_9 (Glo	(None, 2048)	0	vggface_resnet50[2][0]
concatenate_11 (Concatenate)	(None, 4096)	0	global_average_pooling2d_7[0][0] global_average_pooling2d_8[0][0]
concatenate_12 (Concatenate)	(None, 4096)	0	global_max_pooling2d_7[0][0] global_average_pooling2d_9[0][0]
multiply_10 (Multiply)	(None, 4096)	0	concatenate_11[0][0] concatenate_11[0][0]
multiply_11 (Multiply)	(None, 4096)	0	concatenate_12[0][0] concatenate_12[0][0]
subtract_7 (Subtract)	(None, 4096)	0	concatenate_11[0][0] concatenate_12[0][0]
subtract_8 (Subtract)	(None, 4096)	0	multiply_10[0][0] multiply_11[0][0]
multiply_9 (Multiply)	(None, 4096)	0	subtract_7[0][0]

multiply_9 (Multiply)	(None, 4096)	0	subtract_7[0][0] subtract_7[0][0]
concatenate_13 (Concatenate)	(None, 8192)	0	subtract_8[0][0] multiply_9[0][0]
reshape_1 (Reshape)	(None, 8192, 1)	0	concatenate_13[0][0]
conv1d_1 (Conv1D)	(None, 1024, 100)	900	reshape_1[0][0]
conv1d_2 (Conv1D)	(None, 1017, 100)	80100	conv1d_1[0][0]
lstm_1 (LSTM)	(None, 1017, 128)	117248	conv1d_2[0][0]
dropout_12 (Dropout)	(None, 1017, 128)	0	lstm_1[0][0]
lstm_2 (LSTM)	(None, 1017, 128)	131584	dropout_12[0][0]
batch_normalization_2 (BatchNormalizatio	(None, 1017, 128)	512	lstm_2[0][0]
flatten_2 (Flatten)	(None, 130176)	0	batch_normalization_2[0][0]
dense_14 (Dense)	(None, 100)	13017700	flatten_2[0][0]
activation_250 (Activation)	(None, 100)	0	dense_14[0][0]
dropout_13 (Dropout)	(None, 100)	0	activation_250[0][0]
dense_15 (Dense)	(None, 64)	6464	dropout_13[0][0]
activation_251 (Activation)	(None, 64)	0	dense_15[0][0]
dropout_14 (Dropout)	(None, 64)	0	activation_251[0][0]
dense_16 (Dense)	(None, 32)	2080	dropout_14[0][0]
activation_252 (Activation)	(None, 32)	0	dense_16[0][0]
dropout_15 (Dropout)	(None, 32)	0	activation_252[0][0]
dense_17 (Dense)	(None, 16)	528	dropout_15[0][0]
activation_253 (Activation)	(None, 16)	0	dense_17[0][0]
dropout_16 (Dropout)	(None, 16)	0	activation_253[0][0]
dense_18 (Dense)	(None, 1)	17	dropout_16[0][0]
=====			
Total params: 36,918,285			
Trainable params: 36,864,909			
Non-trainable params: 53,376			

In []:

MODEL 6

In [23]:

```
from collections import defaultdict
from glob import glob
from random import choice, sample
import matplotlib.pyplot as plt
from tqdm import tqdm
import numpy as np
import pandas as pd
import pickle
import gc, psutil, random, base64, h5py, os,shutil
#import cv2
pd.options.mode.chained_assignment = None
pd.options.display.max_columns = 9999
```

```

pd.options.display.float_format = '{:20, .2f}'.format
from keras import backend as K

from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Concatenate, Multiply
, Dropout, Subtract, Add
from keras_vggface.vggface import VGGFace
from pathlib import Path

from PIL import Image
from io import BytesIO
from IPython.display import HTML
import plotly.offline as py
py.init_notebook_mode(connected=True)
import plotly.graph_objs as go
import plotly.tools as tls

import warnings
warnings.filterwarnings('ignore')
from keras.layers import Conv2D, ZeroPadding2D, Activation, concatenate
from keras.layers.pooling import MaxPooling2D, AveragePooling2D
from keras.layers.core import Lambda
import keras
from keras.callbacks import ModelCheckpoint, ReduceLROnPlateau, EarlyStopping
from keras.layers import Conv1D, BatchNormalization, Reshape, LeakyReLU
from keras.preprocessing import image
from keras.optimizers import Adam, SGD, rmsprop
from keras.layers import Input, Dense, Flatten, GlobalMaxPool2D, GlobalAvgPool2D, Conv2D, MaxPooling1D, GlobalAveragePooling1D, LSTM
from keras.models import Model
from keras.losses import binary_crossentropy
import tensorflow as tf
from sklearn.metrics import roc_auc_score
from keras import regularizers
%pylab inline
import matplotlib.image as mpimg

```

Populating the interactive namespace from numpy and matplotlib

In [24]:

```

from keras.layers import Conv2D, ZeroPadding2D, Activation, Input, concatenate
from keras.layers.core import Lambda, Flatten, Dense
from keras.layers.normalization import BatchNormalization
from keras.layers.pooling import MaxPooling2D, AveragePooling2D
from keras.models import Model
from keras import backend as K
import pandas as pd
import numpy as np
import keras
from keras.models import *
from keras.layers import *
from keras.preprocessing import *
from keras.callbacks import *
from keras.optimizers import *
from tqdm import tqdm
import glob
from keras.applications.resnet50 import preprocess_input, decode_predictions

# from keras.applications.nasnet import preprocess_input, decode_predictions
# from keras.applications.densenet import preprocess_input, decode_predictions
from keras.preprocessing import image
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt
import sys

#import cv2
from random import choice, sample
import datetime
from sklearn.metrics import roc_auc_score
from keras.callbacks import ReduceLROnPlateau, ModelCheckpoint, EarlyStopping
import threading

```

```

from keras.initializers import glorot_normal

import tensorflow as tf
import numpy as np
import os

from numpy import genfromtxt
from keras.layers import Conv2D, ZeroPadding2D, Activation
from keras.layers.normalization import BatchNormalization

```

In [25]:

```

import tensorflow as tf
from tensorflow.python.keras import backend as K
sess = K.get_session()
#array = sess.run(your_tensor)

```

In [26]:

```

# https://github.com/rcmalli/keras-vggface/blob/master/keras_vggface/utils.py
def preprocess_input(x, data_format=None, version=1):
    """
    preprocess_input will take the numpy array of image and remove the unnecessary values
    The out ut numpy only contains thermal image of face which will have the umporatant chracters
    Here we subtract the values from image numpy array after subtraction redundant pixels are removed

    """
    x_temp = np.copy(x)
    if data_format is None:
        data_format = K.image_data_format()
    assert data_format in {'channels_last', 'channels_first'}

    if version == 1:
        if data_format == 'channels_first':
            x_temp = x_temp[:, ::-1, ...]
            x_temp[:, 0, :, :] -= 93.5940
            x_temp[:, 1, :, :] -= 104.7624
            x_temp[:, 2, :, :] -= 129.1863
        else:
            x_temp = x_temp[..., ::-1]
            x_temp[..., 0] -= 93.5940
            x_temp[..., 1] -= 104.7624
            x_temp[..., 2] -= 129.1863

    elif version == 2:
        if data_format == 'channels_first':
            x_temp = x_temp[:, ::-1, ...]
            x_temp[:, 0, :, :] -= 91.4953
            x_temp[:, 1, :, :] -= 103.8827
            x_temp[:, 2, :, :] -= 131.0912
        else:
            x_temp = x_temp[..., ::-1]
            x_temp[..., 0] -= 91.4953
            x_temp[..., 1] -= 103.8827
            x_temp[..., 2] -= 131.0912
    else:
        raise NotImplementedError
    return x_temp

```

In [27]:

```

# Function for reading images
def read_img(path, IMG_SIZE):
    img = image.load_img(path, target_size=(IMG_SIZE, IMG_SIZE))
    img = np.array(img).astype(np.float)
    return preprocess_input(img, version=2)

```

In [28]:

```

def auc(y_true, y_pred):
    return tf.py_function(roc_auc_score, (y_true, y_pred), tf.double)

```

In [29]:

```
def get_thumbnail(path):
    if path and os.path.exists(path):
        i = Image.open(path)
        i.thumbnail((150, 150), Image.LANCZOS)
        return i

def image_base64(im):
    if isinstance(im, str):
        im = get_thumbnail(im)
    with BytesIO() as buffer:
        im.save(buffer, 'jpeg')
        return base64.b64encode(buffer.getvalue()).decode()

def add_image_path(x):
    image_path = 'train/' + x
    if os.path.exists(image_path):
        path = os.path.join(image_path, os.listdir(image_path)[0])
        return path
```

In [30]:

```
# https://www.kaggle.com/janpreets/just-another-feature-extractor-0-824-1b
def outer_product(x):
    """
    calculate outer-products of 2 tensors

    args
        x
            list of 2 tensors
            , assuming each of which has shape = (size_minibatch, total_pixels, size_filter)
    """
    return keras.backend.batch_dot(
        x[0]
        , x[1]
        , axes=[1,1]
    ) / x[0].get_shape().as_list()[1]

def signed_sqrt(x):
    """
    calculate element-wise signed square root

    args
        x
            a tensor
    """
    return keras.backend.sign(x) * keras.backend.sqrt(keras.backend.abs(x) + 1e-9)

def L2_norm(x, axis=-1):
    """
    calculate L2-norm

    args
        x
            a tensor
    """
    return keras.backend.l2_normalize(x, axis=axis)
```

In []:

In [32]:

```
def inception_based_model():
    myInput = Input(shape=(96, 96, 3))

    x = ZeroPadding2D(padding=(3, 3), input_shape=(96, 96, 3))(myInput)
    x = Conv2D(64, (7, 7), strides=(2, 2), name='conv1')(x)
    x = BatchNormalization(axis=3, epsilon=0.00001, name='bn1')(x)
```

```

x = Activation('relu')(x)
x = ZeroPadding2D(padding=(1, 1))(x)
x = MaxPooling2D(pool_size=3, strides=2)(x)
x = Lambda(LRN2D, name='lrn_1')(x)
x = Conv2D(64, (1, 1), name='conv2')(x)
x = BatchNormalization(axis=3, epsilon=0.00001, name='bn2')(x)
x = Activation('relu')(x)
x = ZeroPadding2D(padding=(1, 1))(x)
x = Conv2D(192, (3, 3), name='conv3')(x)
x = BatchNormalization(axis=3, epsilon=0.00001, name='bn3')(x)
x = Activation('relu')(x)
x = Lambda(LRN2D, name='lrn_2')(x)
x = ZeroPadding2D(padding=(1, 1))(x)
x = MaxPooling2D(pool_size=3, strides=2)(x)

# Inception3a
inception_3a_3x3 = Conv2D(96, (1, 1), name='inception_3a_3x3_conv1')(x)
inception_3a_3x3 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_3x3_bn1')(inception_3a_3x3)
inception_3a_3x3 = Activation('relu')(inception_3a_3x3)
inception_3a_3x3 = ZeroPadding2D(padding=(1, 1))(inception_3a_3x3)
inception_3a_3x3 = Conv2D(128, (3, 3), name='inception_3a_3x3_conv2')(inception_3a_3x3)
inception_3a_3x3 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_3x3_bn2')(inception_3a_3x3)
inception_3a_3x3 = Activation('relu')(inception_3a_3x3)

inception_3a_5x5 = Conv2D(16, (1, 1), name='inception_3a_5x5_conv1')(x)
inception_3a_5x5 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_5x5_bn1')(inception_3a_5x5)
inception_3a_5x5 = Activation('relu')(inception_3a_5x5)
inception_3a_5x5 = ZeroPadding2D(padding=(2, 2))(inception_3a_5x5)
inception_3a_5x5 = Conv2D(32, (5, 5), name='inception_3a_5x5_conv2')(inception_3a_5x5)
inception_3a_5x5 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_5x5_bn2')(inception_3a_5x5)
inception_3a_5x5 = Activation('relu')(inception_3a_5x5)

inception_3a_pool = MaxPooling2D(pool_size=3, strides=2)(x)
inception_3a_pool = Conv2D(32, (1, 1), name='inception_3a_pool_conv')(inception_3a_pool)
inception_3a_pool = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_pool_bn')(inception_3a_pool)
inception_3a_pool = Activation('relu')(inception_3a_pool)
inception_3a_pool = ZeroPadding2D(padding=((3, 4), (3, 4)))(inception_3a_pool)

inception_3a_1x1 = Conv2D(64, (1, 1), name='inception_3a_1x1_conv')(x)
inception_3a_1x1 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3a_1x1_bn')(inception_3a_1x1)
inception_3a_1x1 = Activation('relu')(inception_3a_1x1)

inception_3a = concatenate([inception_3a_3x3, inception_3a_5x5, inception_3a_pool, inception_3a_1x1], axis=3)

# Inception3b
inception_3b_3x3 = Conv2D(96, (1, 1), name='inception_3b_3x3_conv1')(inception_3a)
inception_3b_3x3 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_3x3_bn1')(inception_3b_3x3)
inception_3b_3x3 = Activation('relu')(inception_3b_3x3)
inception_3b_3x3 = ZeroPadding2D(padding=(1, 1))(inception_3b_3x3)
inception_3b_3x3 = Conv2D(128, (3, 3), name='inception_3b_3x3_conv2')(inception_3b_3x3)
inception_3b_3x3 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_3x3_bn2')(inception_3b_3x3)
inception_3b_3x3 = Activation('relu')(inception_3b_3x3)

inception_3b_5x5 = Conv2D(32, (1, 1), name='inception_3b_5x5_conv1')(inception_3a)
inception_3b_5x5 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_5x5_bn1')(inception_3b_5x5)
inception_3b_5x5 = Activation('relu')(inception_3b_5x5)
inception_3b_5x5 = ZeroPadding2D(padding=(2, 2))(inception_3b_5x5)
inception_3b_5x5 = Conv2D(64, (5, 5), name='inception_3b_5x5_conv2')(inception_3b_5x5)
inception_3b_5x5 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_5x5_bn2')(inception_3b_5x5)
inception_3b_5x5 = Activation('relu')(inception_3b_5x5)

inception_3b_pool = AveragePooling2D(pool_size=(3, 3), strides=(3, 3))(inception_3a)
inception_3b_pool = Conv2D(64, (1, 1), name='inception_3b_pool_conv')(inception_3b_pool)
inception_3b_pool = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_pool_bn')(inception_3b_pool)
inception_3b_pool = Activation('relu')(inception_3b_pool)

```



```

inception_3b_pool = Activation('relu', (inception_3b_pool,
inception_3b_pool = ZeroPadding2D(padding=(4, 4))(inception_3b_pool)

inception_3b_1x1 = Conv2D(64, (1, 1), name='inception_3b_1x1_conv')(inception_3a)
inception_3b_1x1 = BatchNormalization(axis=3, epsilon=0.00001, name='inception_3b_1x1_bn')(inception_3b_1x1)
inception_3b_1x1 = Activation('relu')(inception_3b_1x1)

inception_3b = concatenate([inception_3b_3x3, inception_3b_5x5, inception_3b_pool, inception_3b_1x1
], axis=3)

# Inception3c
inception_3c_3x3 = conv2d_bn(inception_3b,
                             layer='inception_3c_3x3',
                             cv1_out=128,
                             cv1_filter=(1, 1),
                             cv2_out=256,
                             cv2_filter=(3, 3),
                             cv2_strides=(2, 2),
                             padding=(1, 1))

inception_3c_5x5 = conv2d_bn(inception_3b,
                             layer='inception_3c_5x5',
                             cv1_out=32,
                             cv1_filter=(1, 1),
                             cv2_out=64,
                             cv2_filter=(5, 5),
                             cv2_strides=(2, 2),
                             padding=(2, 2))

inception_3c_pool = MaxPooling2D(pool_size=3, strides=2)(inception_3b)
inception_3c_pool = ZeroPadding2D(padding=((0, 1), (0, 1)))(inception_3c_pool)

inception_3c = concatenate([inception_3c_3x3, inception_3c_5x5, inception_3c_pool], axis=3)

#inception 4a
inception_4a_3x3 = conv2d_bn(inception_3c,
                             layer='inception_4a_3x3',
                             cv1_out=96,
                             cv1_filter=(1, 1),
                             cv2_out=192,
                             cv2_filter=(3, 3),
                             cv2_strides=(1, 1),
                             padding=(1, 1))

inception_4a_5x5 = conv2d_bn(inception_3c,
                             layer='inception_4a_5x5',
                             cv1_out=32,
                             cv1_filter=(1, 1),
                             cv2_out=64,
                             cv2_filter=(5, 5),
                             cv2_strides=(1, 1),
                             padding=(2, 2))

inception_4a_pool = AveragePooling2D(pool_size=(3, 3), strides=(3, 3))(inception_3c)
inception_4a_pool = conv2d_bn(inception_4a_pool,
                             layer='inception_4a_pool',
                             cv1_out=128,
                             cv1_filter=(1, 1),
                             padding=(2, 2))

inception_4a_1x1 = conv2d_bn(inception_3c,
                             layer='inception_4a_1x1',
                             cv1_out=256,
                             cv1_filter=(1, 1))

inception_4a = concatenate([inception_4a_3x3, inception_4a_5x5, inception_4a_pool, inception_4a_1x1
], axis=3)

#inception4e
inception_4e_3x3 = conv2d_bn(inception_4a,
                             layer='inception_4e_3x3',
                             cv1_out=160,
                             cv1_filter=(1, 1),
                             cv2_out=256,
                             cv2_filter=(3, 3),
                             cv2_strides=(2, 2),
                             padding=(1, 1))

inception_4e_5x5 = conv2d_bn(inception_4a,
                             layer='inception_4e_5x5',
                             cv1_out=64

```

```

        cv1_out=cv1,
        cv1_filter=(1, 1),
        cv2_out=128,
        cv2_filter=(5, 5),
        cv2_strides=(2, 2),
        padding=(2, 2))
inception_4e_pool = MaxPooling2D(pool_size=3, strides=2)(inception_4a)
inception_4e_pool = ZeroPadding2D(padding=((0, 1), (0, 1)))(inception_4e_pool)

inception_4e = concatenate([inception_4e_3x3, inception_4e_5x5, inception_4e_pool], axis=3)

#inception5a
inception_5a_3x3 = conv2d_bn(inception_4e,
                             layer='inception_5a_3x3',
                             cv1_out=96,
                             cv1_filter=(1, 1),
                             cv2_out=384,
                             cv2_filter=(3, 3),
                             cv2_strides=(1, 1),
                             padding=(1, 1))

inception_5a_pool = AveragePooling2D(pool_size=(3, 3), strides=(3, 3))(inception_4e)
inception_5a_pool = conv2d_bn(inception_5a_pool,
                              layer='inception_5a_pool',
                              cv1_out=96,
                              cv1_filter=(1, 1),
                              padding=(1, 1))

inception_5a_1x1 = conv2d_bn(inception_4e,
                              layer='inception_5a_1x1',
                              cv1_out=256,
                              cv1_filter=(1, 1))

inception_5a = concatenate([inception_5a_3x3, inception_5a_pool, inception_5a_1x1], axis=3)

#inception 5b
inception_5b_3x3 = conv2d_bn(inception_5a,
                              layer='inception_5b_3x3',
                              cv1_out=96,
                              cv1_filter=(1, 1),
                              cv2_out=384,
                              cv2_filter=(3, 3),
                              cv2_strides=(1, 1),
                              padding=(1, 1))

inception_5b_pool = MaxPooling2D(pool_size=3, strides=2)(inception_5a)
inception_5b_pool = conv2d_bn(inception_5b_pool,
                              layer='inception_5b_pool',
                              cv1_out=96,
                              cv1_filter=(1, 1))

inception_5b_pool = ZeroPadding2D(padding=(1, 1))(inception_5b_pool)

inception_5b_1x1 = conv2d_bn(inception_5a,
                              layer='inception_5b_1x1',
                              cv1_out=256,
                              cv1_filter=(1, 1))

inception_5b = concatenate([inception_5b_3x3, inception_5b_pool, inception_5b_1x1], axis=3)

av_pool = AveragePooling2D(pool_size=(3, 3), strides=(1, 1))(inception_5b)
reshape_layer = Flatten()(av_pool)
dense_layer = Dense(128, name='dense_layer')(reshape_layer)
norm_layer = Lambda(lambda x: K.l2_normalize(x, axis=1), name='norm_layer')(dense_layer)

return Model(inputs=[myInput], outputs=norm_layer)

```

In [33]:

```

def LRN2D(x):
    return tf.nn.lrn(x, alpha=1e-4, beta=0.75)

def conv2d_bn(
    x,
    layer=None,
    cv1_out=None,
    cv1_filter=(1, 1),
    cv1_strides=(1, 1),
    cv2_out=None,
    cv2_filter=(3, 3),

```

```

cv2_strides=(1, 1),
padding=None,
):
    num = '' if cv2_out == None else '1'
    tensor = Conv2D(cv1_out, cv1_filter, strides=cv1_strides, name=layer+'__conv'+num)(x)
    tensor = BatchNormalization(axis=3, epsilon=0.00001, name=layer+'__bn'+num)(tensor)
    tensor = Activation('relu')(tensor)
    if padding == None:
        return tensor
    tensor = ZeroPadding2D(padding=padding)(tensor)
    if cv2_out == None:
        return tensor
    tensor = Conv2D(cv2_out, cv2_filter, strides=cv2_strides, name=layer+'__conv'+num+2)(tensor)
    tensor = BatchNormalization(axis=3, epsilon=0.00001, name=layer+'__bn'+num+2)(tensor)
    tensor = Activation('relu')(tensor)
    return tensor

def create_base_network(input_shape):
    mod = inception_base_model()
    new_mod = Model(mod.input, mod.layers[-5].output)
    return new_mod

def get_model6():
    input_shape = (96, 96, 3)
    input_a = Input(shape=input_shape)
    input_b = Input(shape=input_shape)
    base_network = create_base_network(input_shape)

    # because we re-use the same instance `base_network`,
    # the weights of the network
    # will be shared across the two branches
    processed_a = base_network(input_a)
    processed_b = base_network(input_b)

    # extract features from detector
    x_detector = processed_a
    shape_detector = processed_a.shape

    # extract features from extractor, same with detector for symmetry DxD model
    shape_extractor = processed_b.shape
    x_extractor = processed_b

    # reshape to (minibatch_size, total_pixels, filter_size)
    x_detector = keras.layers.Reshape([shape_detector[1] * shape_detector[2], shape_detector[-1]])(x_detector)
    x_extractor = keras.layers.Reshape([shape_extractor[1] * shape_extractor[2], shape_extractor[-1]])(x_extractor)

    # outer products of features, output shape=(minibatch_size, filter_size_detector*filter_size_extractor)
    x = keras.layers.Lambda(outer_product)([x_detector, x_extractor])

    # reshape to (minibatch_size, filter_size_detector*filter_size_extractor)
    x = keras.layers.Reshape([shape_detector[-1]*shape_extractor[-1]])(x)
    # signed square-root
    x = keras.layers.Lambda(signed_sqrt)(x)
    # L2 normalization
    x = keras.layers.Lambda(L2_norm)(x)

    ###
    ### attach FC-Layer
    ###

    x = Dense(100, activation="relu")(x)
    x = Dropout(0.01)(x)

    out = Dense(units=1, kernel_regularizer=keras.regularizers.l2(1e-8), kernel_initializer='glorot_normal', activation="sigmoid")(x)

    model = Model([input_a, input_b], out)

    model.compile(loss="binary_crossentropy", metrics=['acc', 'auc'], optimizer=Adam(0.001))

```

```
print(model.summary())

return model
```

In []:

In [34]:

```
model6 = get_model6()
```

Model: "model_8"

Layer (type)	Output Shape	Param #	Connected to
input_16 (InputLayer)	(None, 96, 96, 3)	0	
input_17 (InputLayer)	(None, 96, 96, 3)	0	
model_7 (Model)	(None, 3, 3, 736)	3648944	input_16[0][0] input_17[0][0]
reshape_2 (Reshape)	(None, 9, 736)	0	model_7[1][0]
reshape_3 (Reshape)	(None, 9, 736)	0	model_7[2][0]
lambda_1 (Lambda)	(None, 736, 736)	0	reshape_2[0][0] reshape_3[0][0]
reshape_4 (Reshape)	(None, 541696)	0	lambda_1[0][0]
lambda_2 (Lambda)	(None, 541696)	0	reshape_4[0][0]
lambda_3 (Lambda)	(None, 541696)	0	lambda_2[0][0]
dense_19 (Dense)	(None, 100)	54169700	lambda_3[0][0]
dropout_17 (Dropout)	(None, 100)	0	dense_19[0][0]
dense_20 (Dense)	(None, 1)	101	dropout_17[0][0]
Total params: 57,818,745			
Trainable params: 57,809,433			
Non-trainable params: 9,312			
None			

In []:

FINAL MODEL ENSEMBLING

In [35]:

```
import numpy as np
import pandas as pd
```

In [36]:

```
# Function for reading images
def read_img(path, IMG_SIZE):
    img = image.load_img(path, target_size=(IMG_SIZE, IMG_SIZE))
    img = np.array(img).astype(np.float)
    return preprocess_input(img, version=2)
```

In []:

In [37]:

```
from tqdm import tqdm
test_path = "../test/"

def final_model(test_path):
    def chunker(seq, size=32):
        return (seq[pos:pos + size] for pos in range(0, len(seq), size))

    submission1 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission2 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission3 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission4 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission5 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission6 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')

    predictions1 = []
    predictions2 = []
    predictions3 = []
    predictions4 = []
    predictions5 = []
    predictions6 = []

    print('Model1 predictions ....')
    IMG_SIZE = 197
    for batch in tqdm(chunker(submission1.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        model1.load_weights('../SAVE_MODEL1/model1/model1.h5')
        pred1 = model1.predict([X1, X2]).ravel().tolist()
        predictions1 += pred1

    print('Model2 predictions ....')
    IMG_SIZE = 197
    for batch in tqdm(chunker(submission2.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        pred2 = model2.predict([X1, X2]).ravel().tolist()
        model2.load_weights('../SAVE_MODEL1/model2/model2.h5')
        predictions2 += pred2

    print('Model3 predictions ....')
    IMG_SIZE = 224
    for batch in tqdm(chunker(submission3.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        model3.load_weights('../SAVE_MODEL1/model3/model3.h5')
        pred3 = model3.predict([X1, X2]).ravel().tolist()
        predictions3 += pred3

    print('Model4 predictions ....')
    IMG_SIZE = 224
    for batch in tqdm(chunker(submission4.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        model4.load_weights('../SAVE_MODEL1/model4/model4.h5')
        pred4 = model4.predict([X1, X2]).ravel().tolist()
        predictions4 += pred4

    print('Model5 predictions ....')
    IMG_SIZE = 224
    for batch in tqdm(chunker(submission5.img_pair.values)):
```

```

X1 = [x.split("-")[0] for x in batch]
X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
X2 = [x.split("-")[1] for x in batch]
X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
model5.load_weights('../SAVE_MODEL1/model5/model5.h5')
pred5 = model5.predict([X1, X2]).ravel().tolist()
predictions5 += pred5

print('Model6 predictions ....')
IMG_SIZE = 96
for batch in tqdm(chunker(submission6.img_pair.values)):
    X1 = [x.split("-")[0] for x in batch]
    X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
    X2 = [x.split("-")[1] for x in batch]
    X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
    model6.load_weights('../SAVE_MODEL1/model6/model6.h5')
    pred6 = model6.predict([X1, X2]).ravel().tolist()
    predictions6 += pred6

submission1['is_related'] = predictions1
submission2['is_related'] = predictions2
submission3['is_related'] = predictions3
submission4['is_related'] = predictions4
submission5['is_related'] = predictions5
submission6['is_related'] = predictions6

submission1.to_csv("model1.csv", index=False)
submission2.to_csv("model2.csv", index=False)
submission3.to_csv("model3.csv", index=False)
submission4.to_csv("model4.csv", index=False)
submission5.to_csv("model5.csv", index=False)
submission6.to_csv("model6.csv", index=False)

sub1 = pd.read_csv('./model1.csv')
sub2 = pd.read_csv('./model2.csv')
sub3 = pd.read_csv('./model3.csv')
sub4 = pd.read_csv('./model4.csv')
sub5 = pd.read_csv('./model5.csv')
sub6 = pd.read_csv('./model6.csv')

#final=pd.read_csv('./model2.csv')

#print('Combining Models predictions ....')
#final['is_related'] = 0.18*sub2['is_related'] + 0.16*sub3['is_related'] + 0.22*sub4['is_related']
+ 0.22*sub5['is_related'] + 0.22*sub6['is_related']
#final.to_csv('submission_final1.csv', index=False )

return sub1, sub2, sub3, sub4, sub5, sub6

```

In [38]:

```
sub_1, sub_2, sub_3, sub_4, sub_5, sub_6 = final_model(test_path)
```

0it [00:00, ?it/s]

Model1 predictions

166it [24:52, 8.99s/it]

0it [00:00, ?it/s]

Model2 predictions

166it [10:46, 3.89s/it]

0it [00:00, ?it/s]

Model3 predictions

166it [11:56, 4.31s/it]

0it [00:00, ?it/s]

Model4 predictions

```
166it [11:54, 4.31s/it]
0it [00:00, ?it/s]
```

Model5 predictions

```
166it [13:41, 4.95s/it]
0it [00:00, ?it/s]
```

Model6 predictions

```
166it [03:18, 1.19s/it]
```

In []:

```
final=pd.read_csv('./model2.csv')
print('Combining Models predictions ....')
final['is_related'] = 0.18*sub_2['is_related'] + 0.16*sub_3['is_related'] + 0.22*sub_4['is_related'] +
0.22*sub_5['is_related'] + 0.22*sub_6['is_related']
final.to_csv('submission_final1.csv', index=False )
```

In [39]:

```
final=pd.read_csv('./model2.csv')
print('Combining Models predictions ....')
final['is_related'] = 0.17*sub_1['is_related'] + 0.17*sub_2['is_related'] + 0.16*sub_3['is_related'] +
0.16*sub_4['is_related'] + 0.17*sub_5['is_related'] + 0.17*sub_6['is_related']
final.to_csv('submission_final2.csv', index=False )
```

Combining Models predictions

In [40]:

```
from tqdm import tqdm
test_path = "../test/"

def final_model1(test_path):
    def chunker(seq, size=32):
        return (seq[pos:pos + size] for pos in range(0, len(seq), size))

    submission1 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission2 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission3 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission4 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission5 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission6 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')

    predictions1 = []
    predictions2 = []
    predictions3 = []
    predictions4 = []
    predictions5 = []
    predictions6 = []

    print('Model1 predictions ....')
    IMG_SIZE = 197
    for batch in tqdm(chunker(submission1.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        model1.load_weights('../SAVE_MODEL1/model1/model1_0.h5')
        pred1 = model1.predict([X1, X2]).ravel().tolist()
        predictions1 += pred1

    print('Model2 predictions ....')
    IMG_SIZE = 197
    for batch in tqdm(chunker(submission2.img_pair.values)):
```

```

for batch in tqdm(chunker(submission1.img_pair.values)):
    X1 = [x.split("-")[0] for x in batch]
    X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
    X2 = [x.split("-")[1] for x in batch]
    X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
    pred2 = model2.predict([X1, X2]).ravel().tolist()
    model2.load_weights('../SAVE_MODEL1/model2/model2_0.h5')
    predictions2 += pred2

print('Model3 predictions ....')
IMG_SIZE = 224
for batch in tqdm(chunker(submission3.img_pair.values)):
    X1 = [x.split("-")[0] for x in batch]
    X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
    X2 = [x.split("-")[1] for x in batch]
    X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
    model3.load_weights('../SAVE_MODEL1/model3/model3.h5')
    pred3 = model3.predict([X1, X2]).ravel().tolist()
    predictions3 += pred3

print('Model4 predictions ....')
IMG_SIZE = 224
for batch in tqdm(chunker(submission4.img_pair.values)):
    X1 = [x.split("-")[0] for x in batch]
    X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
    X2 = [x.split("-")[1] for x in batch]
    X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
    model4.load_weights('../SAVE_MODEL1/model4/model4.h5')
    pred4 = model4.predict([X1, X2]).ravel().tolist()
    predictions4 += pred4

print('Model5 predictions ....')
IMG_SIZE = 224
for batch in tqdm(chunker(submission5.img_pair.values)):
    X1 = [x.split("-")[0] for x in batch]
    X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
    X2 = [x.split("-")[1] for x in batch]
    X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
    model5.load_weights('../SAVE_MODEL1/model5/model5.h5')
    pred5 = model5.predict([X1, X2]).ravel().tolist()
    predictions5 += pred5

print('Model6 predictions ....')
IMG_SIZE = 96
for batch in tqdm(chunker(submission6.img_pair.values)):
    X1 = [x.split("-")[0] for x in batch]
    X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
    X2 = [x.split("-")[1] for x in batch]
    X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
    model6.load_weights('../SAVE_MODEL1/model6/model6.h5')
    pred6 = model6.predict([X1, X2]).ravel().tolist()
    predictions6 += pred6

submission1['is_related'] = predictions1
submission2['is_related'] = predictions2
submission3['is_related'] = predictions3
submission4['is_related'] = predictions4
submission5['is_related'] = predictions5
submission6['is_related'] = predictions6

submission1.to_csv("model1.csv", index=False)
submission2.to_csv("model2.csv", index=False)
submission3.to_csv("model3.csv", index=False)
submission4.to_csv("model4.csv", index=False)
submission5.to_csv("model5.csv", index=False)
submission6.to_csv("model6.csv", index=False)

sub1 = pd.read_csv('./model1.csv')
sub2 = pd.read_csv('./model2.csv')
sub3 = pd.read_csv('./model3.csv')
sub4 = pd.read_csv('./model4.csv')
sub5 = pd.read_csv('./model5.csv')
sub6 = pd.read_csv('./model6.csv')

#final=pd.read_csv('./model2.csv')

```



```

    #print('Combining Models predictions ....')
    #final['is_related'] = 0.18*sub2['is_related'] + 0.16*sub3['is_related'] + 0.22*sub4['is_related']
    + 0.22*sub5['is_related'] + 0.22*sub6['is_related']
    #final.to_csv('submission_final1.csv', index=False )

    return sub1, sub2, sub3, sub4, sub5, sub6

```

In []:

In [41]:

```
sub1_1, sub1_2, sub1_3, sub1_4, sub1_5, sub1_6 = final_model1(test_path)
```

0it [00:00, ?it/s]

Model1 predictions

166it [10:06, 3.66s/it]
0it [00:00, ?it/s]

Model2 predictions

166it [10:14, 3.70s/it]
0it [00:00, ?it/s]

Model3 predictions

166it [11:59, 4.33s/it]
0it [00:00, ?it/s]

Model4 predictions

166it [11:52, 4.29s/it]
0it [00:00, ?it/s]

Model5 predictions

166it [13:50, 5.00s/it]
0it [00:00, ?it/s]

Model6 predictions

166it [03:20, 1.21s/it]

In [42]:

```

final1=pd.read_csv('./model2.csv')
print('Combining Models predictions ....')
final1['is_related'] = 0.18*sub1_2['is_related'] + 0.16*sub1_3['is_related'] + 0.22*sub1_4['is_related']
+ 0.22*sub1_5['is_related'] + 0.22*sub1_6['is_related']
final1.to_csv('submission_final3.csv', index=False )

```

Combining Models predictions

In [43]:

```

final1=pd.read_csv('./model2.csv')
print('Combining Models predictions ....')
final1['is_related'] = 0.17*sub1_1['is_related'] + 0.17*sub1_2['is_related'] + 0.16*sub1_3['is_related']
+ 0.16*sub1_4['is_related'] + 0.17*sub1_5['is_related'] + 0.17*sub1_6['is_related']
final1.to_csv('submission_final4.csv', index=False )

```

```
final1.to_csv('submission_final4.csv', index=False)
```

Combining Models predictions

In []:

In [44]:

```
from tqdm import tqdm
test_path = "../test/"

def final_model2(test_path):
    def chunker(seq, size=32):
        return (seq[pos:pos + size] for pos in range(0, len(seq), size))

    submission1 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission2 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission3 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission4 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission5 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')
    submission6 = pd.read_csv('../recognizing-faces-in-the-wild/sample_submission.csv')

    predictions1 = []
    predictions2 = []
    predictions3 = []
    predictions4 = []
    predictions5 = []
    predictions6 = []

    print('Model1 predictions ....')
    IMG_SIZE = 197
    for batch in tqdm(chunker(submission1.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        model1.load_weights('../SAVE_MODEL1/model1/model1_0.h5')
        pred1 = model1.predict([X1, X2]).ravel().tolist()
        predictions1 += pred1

    print('Model2 predictions ....')
    IMG_SIZE = 197
    for batch in tqdm(chunker(submission2.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        pred2 = model2.predict([X1, X2]).ravel().tolist()
        model2.load_weights('../SAVE_MODEL1/model2/model2_0.h5')
        predictions2 += pred2

    print('Model3 predictions ....')
    IMG_SIZE = 224
    for batch in tqdm(chunker(submission3.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        model3.load_weights('../SAVE_MODEL1/model3/model3.h5')
        pred3 = model3.predict([X1, X2]).ravel().tolist()
        predictions3 += pred3

    print('Model4 predictions ....')
    IMG_SIZE = 224
    for batch in tqdm(chunker(submission4.img_pair.values)):
        X1 = [x.split("-")[0] for x in batch]
        X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
        X2 = [x.split("-")[1] for x in batch]
        X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
        model4.load_weights('../SAVE_MODEL1/model4/model4.h5')
        pred4 = model4.predict([X1, X2]).ravel().tolist()
```

```

        predictions4 += pred4

print('Model5 predictions ....')
IMG_SIZE = 224
for batch in tqdm(chunker(submission5.img_pair.values)):
    X1 = [x.split("-")[0] for x in batch]
    X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
    X2 = [x.split("-")[1] for x in batch]
    X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
    model5.load_weights('../SAVE_MODEL1/model5/model5.h5')
    pred5 = model5.predict([X1, X2]).ravel().tolist()
    predictions5 += pred5

print('Model6 predictions ....')
IMG_SIZE = 96
for batch in tqdm(chunker(submission6.img_pair.values)):
    X1 = [x.split("-")[0] for x in batch]
    X1 = np.array([read_img(test_path + x, IMG_SIZE) for x in X1])
    X2 = [x.split("-")[1] for x in batch]
    X2 = np.array([read_img(test_path + x, IMG_SIZE) for x in X2])
    model6.load_weights('../SAVE_MODEL1/model6/model6_0.h5')
    pred6 = model6.predict([X1, X2]).ravel().tolist()
    predictions6 += pred6

submission1['is_related'] = predictions1
submission2['is_related'] = predictions2
submission3['is_related'] = predictions3
submission4['is_related'] = predictions4
submission5['is_related'] = predictions5
submission6['is_related'] = predictions6

submission1.to_csv("model1.csv", index=False)
submission2.to_csv("model2.csv", index=False)
submission3.to_csv("model3.csv", index=False)
submission4.to_csv("model4.csv", index=False)
submission5.to_csv("model5.csv", index=False)
submission6.to_csv("model6.csv", index=False)

sub1 = pd.read_csv('./model1.csv')
sub2 = pd.read_csv('./model2.csv')
sub3 = pd.read_csv('./model3.csv')
sub4 = pd.read_csv('./model4.csv')
sub5 = pd.read_csv('./model5.csv')
sub6 = pd.read_csv('./model6.csv')

#final=pd.read_csv('./model2.csv')

#print('Combining Models predictions ....')
#final['is_related'] = 0.18*sub2['is_related'] + 0.16*sub3['is_related'] + 0.22*sub4['is_related']
+ 0.22*sub5['is_related'] + 0.22*sub6['is_related']
#final.to_csv('submission_final1.csv', index=False )

return sub1, sub2, sub3, sub4, sub5, sub6

```

In []:

In [45]:

```
sub2_1, sub2_2, sub2_3, sub2_4, sub2_5, sub2_6 = final_model2(test_path)
```

0it [00:00, ?it/s]

Model1 predictions

166it [10:15, 3.71s/it]

0it [00:00, ?it/s]

Model2 predictions

```
166it [10:10, 3.68s/it]
0it [00:00, ?it/s]
```

Model3 predictions

```
166it [11:46, 4.25s/it]
0it [00:00, ?it/s]
```

Model4 predictions

```
166it [11:50, 4.28s/it]
0it [00:00, ?it/s]
```

Model5 predictions

```
166it [13:44, 4.97s/it]
0it [00:00, ?it/s]
```

Model6 predictions

```
166it [03:18, 1.20s/it]
```

In [46]:

```
final2=pd.read_csv('./model2.csv')
print('Combining Models predictions ....')
final2['is_related'] = 0.18*sub2_2['is_related'] + 0.16*sub2_3['is_related'] + 0.22*sub2_4['is_related']
+ 0.22*sub2_5['is_related'] + 0.22*sub2_6['is_related']
final2.to_csv('submission_final5.csv', index=False )
```

Combining Models predictions

In [50]:

```
final2=pd.read_csv('./model2.csv')
print('Combining Models predictions ....')
final2['is_related'] = 0.17*sub2_1['is_related'] + 0.17*sub2_2['is_related'] + 0.16*sub2_3['is_related']
+ 0.16*sub2_4['is_related'] + 0.17*sub2_5['is_related'] + 0.17*sub2_6['is_related']
final2.to_csv('submission_final6.csv', index=False )
```

Combining Models predictions

In [49]:

```
final2=pd.read_csv('./model2.csv')
print('Combining Models predictions ....')
final2['is_related'] = 0.15*sub2_1['is_related'] + 0.15*sub2_2['is_related'] + 0.15*sub2_3['is_related']
+ 0.15*sub2_4['is_related'] + 0.15*sub2_5['is_related'] + 0.25*sub2_6['is_related']
final2.to_csv('submission_final7.csv', index=False )
```

Combining Models predictions

References: -->

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<https://www.kaggle.com/gowrishankarin/eda-with-plotly-smart-cute-and-pretty-people>

In []:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]:

In [0]: