

Face Recognition

June 16, 2017

Overview of Intelligent face recognition system using deep neural networks

Faces are detected irrespective of expressions and emotions. Detected faces are then fed into neural network to create dataset for clustering. After clustering every cluster has a unique label, by which face recognition(validation) is performed.

Dependencies to be installed

- 1) Linux operating system (Ubuntu 16.04 LTS)
- 2) Python 2.7 (codes are tested with 2.7.13 comes by default with ubuntu)
- 3) Lua (recent stable version) and dependencies installation
 1. `curl -R -O http://www.lua.org/ftp/lua-5.3.4.tar.gz`
 2. `tar xzf lua-5.3.4.tar.gz`
 3. `cd lua-5.3.4/`
 4. `make linux install`
 5. `luarocks install dpnn`
 6. `luarocks install nn`
 7. `luarocks install image`

4) Installing torch

- 1) In terminal, run commands WITHOUT sudo
- 2) `git clone https://github.com/torch/distro.git ~/torch --recursive`
- 3) `cd ~/torch; bash install-deps;`
- 4) `./install.sh`

1. Face detection

Two methods to detect the faces were enclosed. If pictures (.jpeg or .jpg format) are in directory of a file system in local machine then, to visualize how face detection is working, run `face_detect.py faces`(Testing purpose).

To run code : `python face_detect.py`

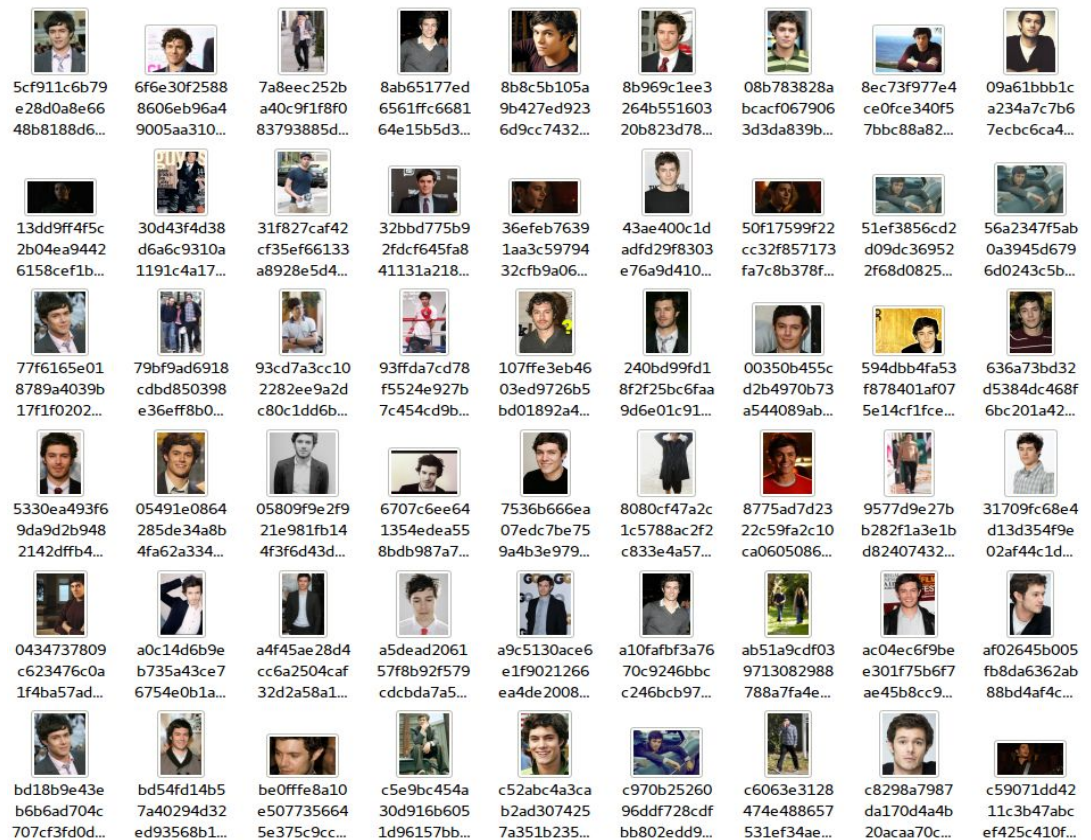
Sample output:



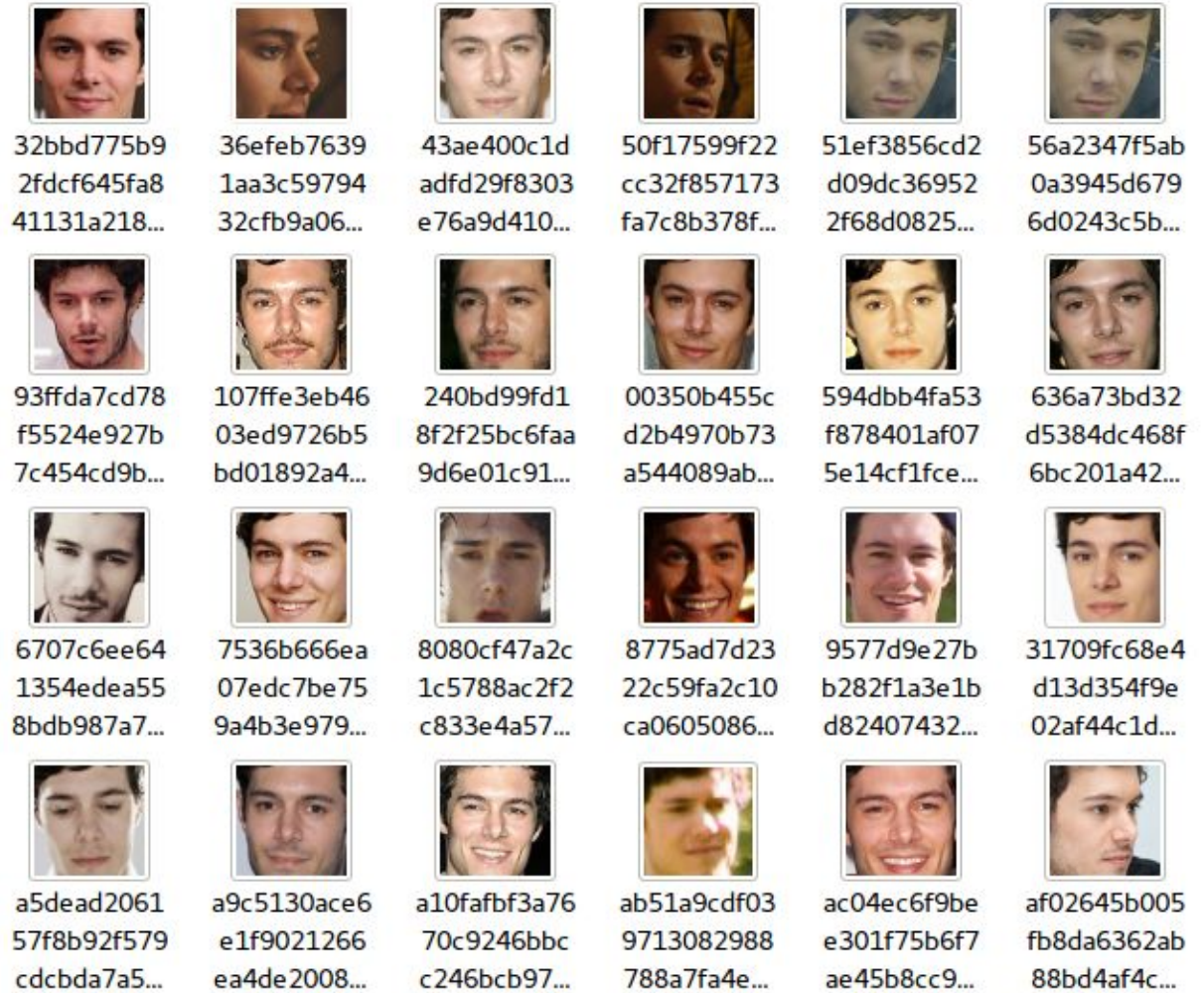
Dataset for face recognition is created by cropping facial images detected in parent picture (parent picture can also be group photo) and then feeded into neural network. Reverse mapping has to be done, when images are projected at phone (from cropped faces to its parent pictures - Database)

Method 1 : Run face_crop2.py to automatically detect faces and crop it to new directory in parent directory (Note: Haarcascade_frontalface_default.xml should be in parent directory while running face_crop2.py script). Haar cascade and dlib libraries are used to detect faces from the images (for more technical information refer code).

Sample : Parent directory containing single person photo or group photo



After cropping, new folder will be created in same directory with cropped faces, these faces should be transformed into vectors for clustering.



Method 2 : Run script download.py to download from url (Note : create uri as per sample files given (file1.txt and file2.txt), then pass file name in list of download.py , all files will be downloaded in path mentioned), example: Run script download.py to get hollywood actresses and actors face pictures.

Creation and Cleaning dataset

Feed the cropped faces into neural network, which will provide output of 128 dimensional vector for each faces. These are considered as features of faces.

a) Run lua script as below (make sure that 'nn4.small2.v1.t7' file should be in directory where lua script is located),

```
th load_images.lua -d path_to_directory_containing_only_faces >> output_file.txt
```

This script will load images in buffer and resize according to specific size (96x96x3x1), this is default input size of images to be feeded into neural network.

As result, all face images in mentioned directory as parameter in above command will be converted into 128 vector value. Those data has to be filtered from default strings. (Note : This step was performed manually finally it should be automated, by opening every "output_file.txt" and removed unwanted strings refer file from_NN.txt (file which stores default output from NN) and refer "to_next_step.txt" (file which is modified) to check the modifications).

Convert "to_next_step.txt" file to csv file to load data frame in python for clustering algorithms. Run convert_csv.py as "python convert_csv.py" to convert "to_next_step.txt" file into "output.csv" file which will result in n x 128 array. Sample file is attached as output.csv

Semi supervised learning model: (Implementation logic)

Problem statement : Building unsupervised model for face recognition.

We have pre trained model of Convolutional neural network which is capable of recognizing faces with 90 to 95% accuracy in case of supervised classification. (Open face - Developed from paper FaceNet <https://arxiv.org/pdf/1503.03832.pdf>)

Initial setup includes clustering of labeled data , so that centroids and farthest point from each centroid (radial distance of each cluster) were stored as pairs in csv file.

Validation techniques :

Once faces are detected from parent image which are cropped into proper size(1,3,96,96) and fed to existing neural network and output is sliced from 25th layer (Convolutional neural network is of 26 layers and because last layer is fully connected which will not map to pre trained faces or nodes we are slicing data from 25th layer). The result of 1 face was 128 vector value which is compared with centroids and farthest point from csv to validate whether result belongs to any of existing clustered faces (Clustered in Initial setup).

Above case is trivial , so that this result is better accuracy. But if result does not belongs to any of the clusters then added to new list, This result is considered as centroid with the minimal radial distance as mean of radial distances of all existing centroids.

For next coming facial images to be recognized will be compared with this newly created centroid in above step. Algorithm is performed recursively.

Once data were clustered and enough data for single person is collected , we can train CNN again to create node in final layer (26th layer) of NN. This technique is called semi supervised learning.

Detailed methodology

Clustering 128 Dimensional data:

Clustering includes two phases, Training (clustering known faces) and validation. In clustering mechanism known faces are clustered (Training phase) for example , If there were 10 unique person faces cropped which are converted into feature vectors loaded as csv file (as many instance of unique person faces are required minimum 1:20 ratio) K means clustering algorithm is performed , which results in 10 clusters, it is mandatory to mention number of person at particular instance of time at training.

Validation : Once clusters are formed their centroid values are stored in csv file externally, where in other hand dataset for validation is passed to be compared with the clusters, here euclidean distance is taken between the validation data point to all other clusters(centroid and farthest point) , if data is nearer to the centroid of any cluster and less than the distance between farthest point from centroid of cluster , then data is added to that cluster. If not ,then it is added to new list of centroids with minimal radial distance as mean of radial distance of all other existing clusters. Further points comes for validation will be checked with newly formed centroid and its radial distance.

Pseudo code

TRAINING :

1. K Means Clustering (Dataset of K persons images) , K = Number of unique person.
2. Result : $\langle C_1, RD_1 \rangle, \dots, \langle C_k, RD_k \rangle$ stored in csv.

VALIDATION:

1. Start timer from 0 mins, each time validation starts
2. For each point in validation data set(V_1 to V_n), perform below :

IF (TIMER < 60 mins){

FOR (i \rightarrow 1 to K) :

Distance , Centroid = FUNCTION (MIN(Euclidean distance (C_i, V_1)) , C_i)

IF (RD(Centroid) > Distance)

Add V_1 to C(Centroid)

ELSE

Add V_1 existing Result list as $\langle C_{new} = V_1 , RD_{new} = \text{MEAN}(\text{Result list } \langle RD_i \rangle)$

K = K+1;

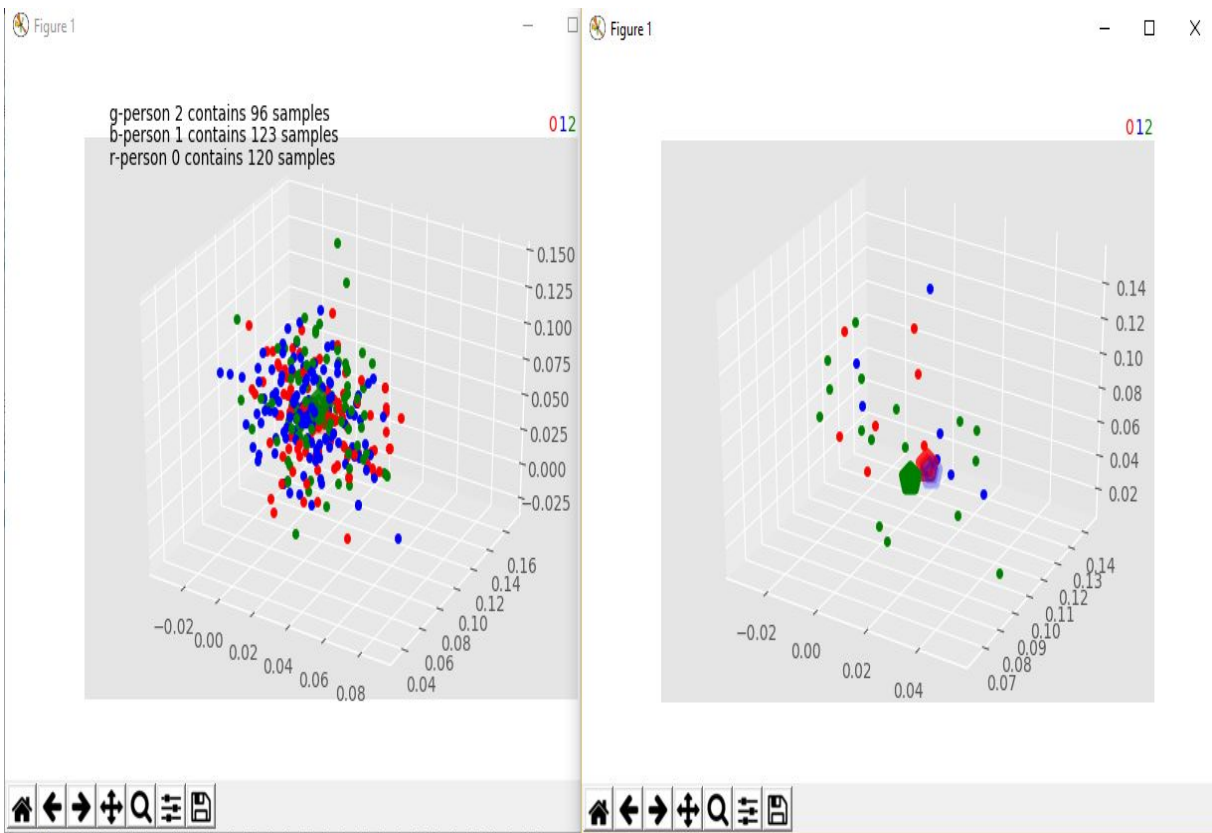
}

ELSE TRAINING with updated K value.

Experiments performed

K means ++ : According to observation on k means algorithm , it will take local optimum value as its centroid so, results are inaccurate, So k means ++ clustering is opted which produces more effective results but not scalable for larger data set with accuracy of 60%.If number of unique person at particular instance is mentioned then accuracy can be increased.

Training and Validation results of k means ++ algorithm :

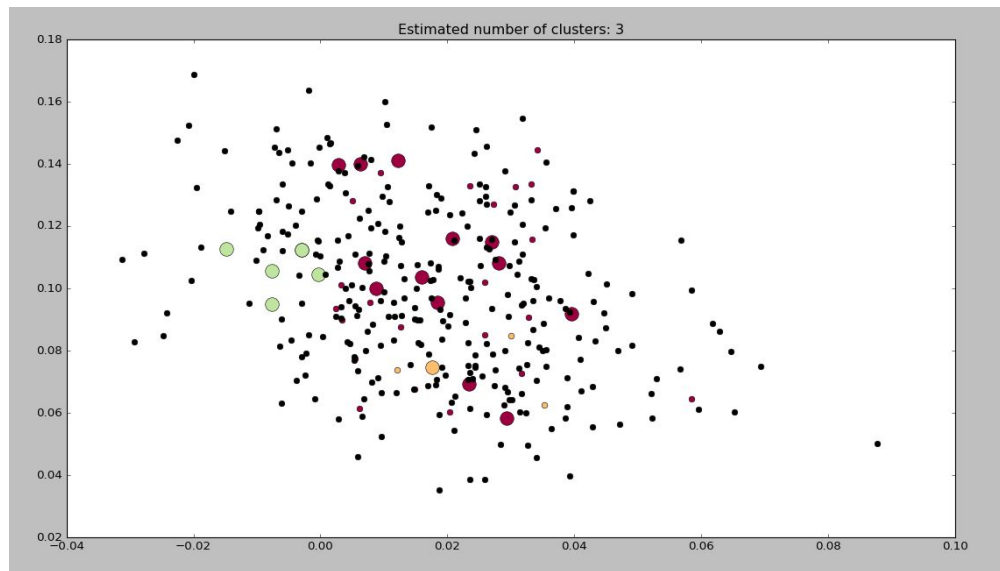


kNN algorithm : Implementation Kd tree and ball tree algorithm were giving more effective results than k means algorithm but for kNN classifier we need to mention number of classes at

particular instance of time. So the problem again falls into supervised learning so we dropped this approach. Here classes unique person faces.

DbScan algorithm : Density based Clustering algorithm is not much efficient due to curse of dimensionality , which resulting in clusters of irrelevant sizes which cannot be processed further

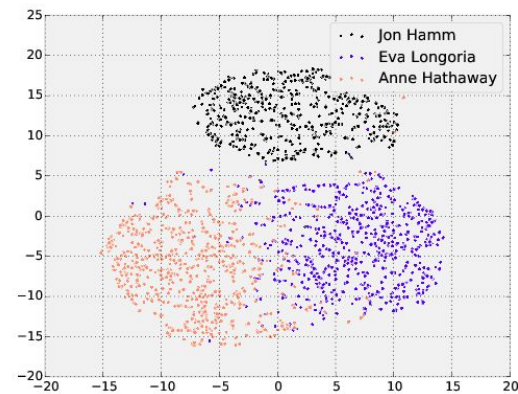
efficiently.



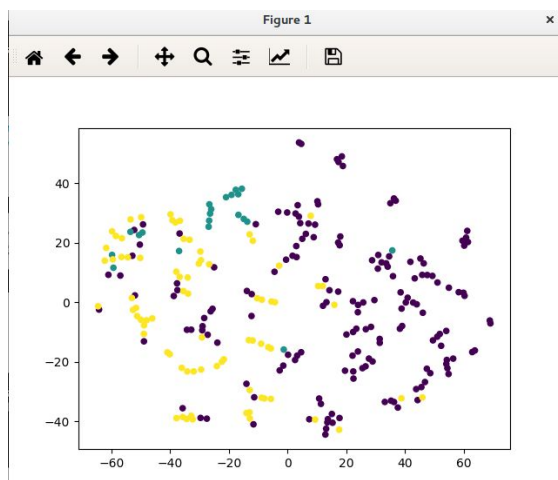
Validation Accuracy using t-SNE :

It is a non linear dimensionality reduction technique that is particularly well-suited for embedding high-dimensional data into a space of two or three dimensions, which can then be

visualized in a scatter plot. Specifically, it models each high-dimensional object by a two- or three-dimensional point in such a way that similar objects are modeled by nearby points and



dissimilar objects are modeled by distant points.



Current work :

1. Experimentation with k means parallel algorithm for better scalability.

2. Local Binary Patterns Histograms for Face Recognition in unsupervised environment.

<http://bitsearch.blogspot.in/2013/02/unsupervised-face-clustering-with-opencv.html>

Results :

Unsupervised model cannot be developed with 95% accuracy provided with very less data, time and computational power.

“ Google help thread on 2016 states - Google’s unsupervised model for 2000 images which was uploaded in photos app in one instance, takes 15 to 20 days to cluster similar faces using unsupervised technique, this feature called as “Group by faces” only available in specific countries.

Face Recognition in Unsupervised Environment is open research problem , Google has api for face detection only. Google Face recognition algorithm is not open sourced.

Google Brain team might be using “Variational Auto encoders” and “Generative Adversarial Neural Networks” to extract features of faces which requires exponentially large dataset of unlabeled faces and high computational power for better accuracies. These concepts are under research.

Papers published on 2017 under review , exactly suits our requirements

<https://arxiv.org/pdf/1611.02648.pdf>

<https://arxiv.org/pdf/1606.05579.pdf>

