# AGE ESTIMATION FROM FACIAL PARTS USING COMPACT MULTI STREAM CONVOLUTIONAL NEURAL NETWORK

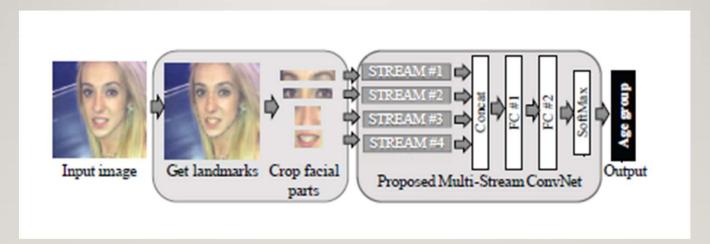
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### INTRODUCTION

- Age plays a key role in many real-world applications such as preventing purchase of alcohol and tobacco by minors(under 16 yr), soft biometrics, electronic customer relationship and as age synthesis to find lost people.
- Face age estimation is defined as the possibility of labelling a face image automatically with the exact age or the age group (e.g., young, adult) of the individual face".
- The proposed approach uses deep learning for age estimation, adopts facial parts as input and a compact multi stream CNN as architecture.

### **OBJECTIVE**

• The main goal of our work is to propose and evaluate a compact multi-stream CNN architecture to explore preprocessed facial parts in order to estimate human age from a single image.



### source:-

https://openaccess.thecvf.com/content\_ICCVW\_2019/papers/CEFRL/Angeloni\_Age\_Estimation\_From\_Facial\_Parts\_Using\_Compact\_Multi-Stream\_Convolutional\_Neural\_ICCVW\_2019\_paper.pdf

### **PROCEDURE**

- 1. A single RGB image is input to the system
- 2. A face detector is applied followed by a 2D facial landmarks estimator.
- 3. Based on the landmark coordinates, the facial parts of interest are preprocessed and cropped.
- 4. Each facial part feeds a specific stream of CNN, whose outputs are concatenated and processed by a sequence of fully connected layers.
- 5. Finally, a softmax function returns the probabilities of the person belonging to which age group.

### PRE-PROCESSING

Dlib face landmarks (68 landmarks)

Face contour: 1 - 17 (Blue)

Eyebrow: 18 - 27 (cyan)

Nose: 27 - 36(red)

Eye: 36 - 48 (green)

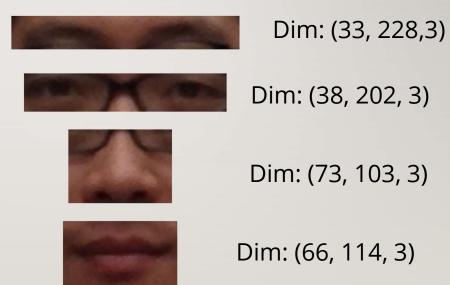
Mouth: 49-68 (yellow)



# CONTD..

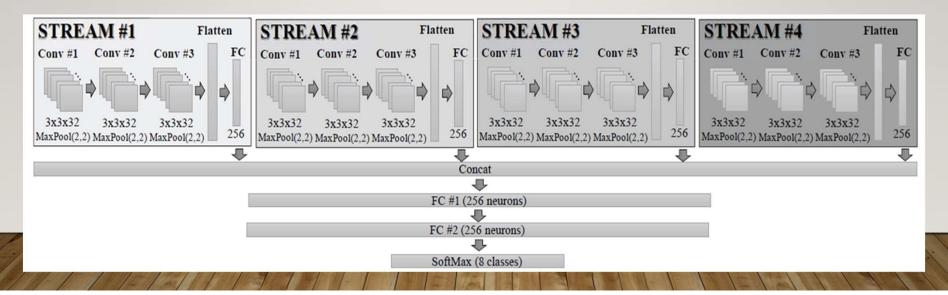
Cropped feature of face using landmarks

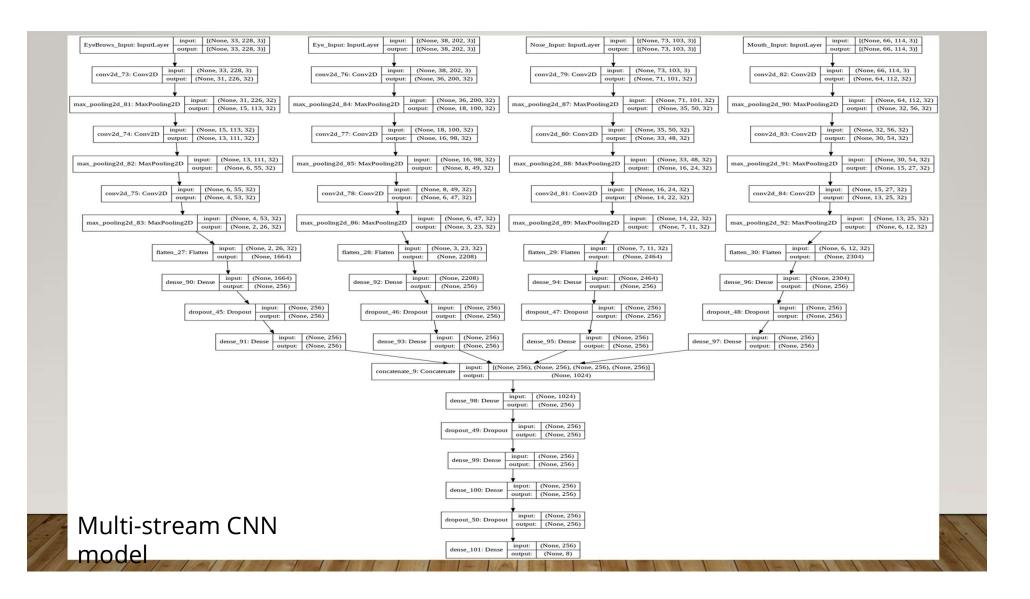




### PROPOSED MULTI STREAM CNN

- Each facial part is process by an independent CNN stream prior to concatenation with other parts.
- The feature learning occurs before concatenating each facial part.
- Output of each dense layer is used to perform the classification of each stream individually.





### **IMPLEMENTATION**

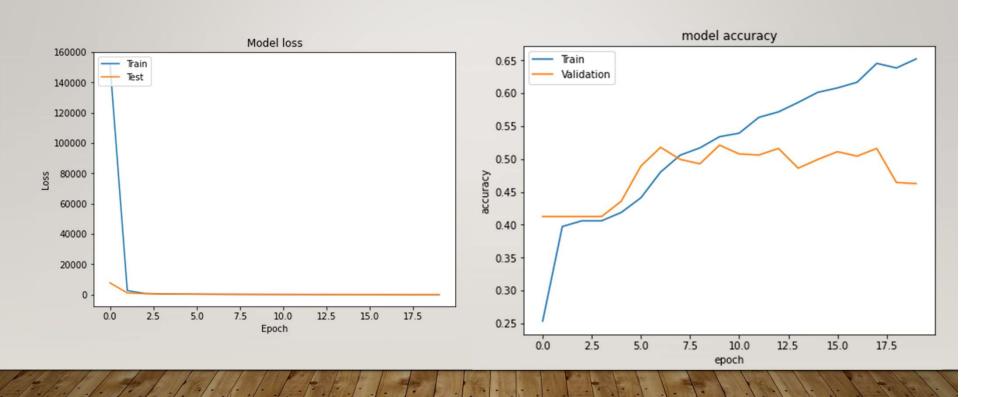
### Training:

- 1. The proposed multi-stream CNNs were implemented using Tensorflow and keras libraries.
- 2. use 20 epochs
- 3. Batch size to 32
- 4. Categorical Cross Entropy was chosen as a loss function
- 5. Adam algorithm was chosen as optimizer with Ir = 0.001

### Dataset:

- 1. Adience benchmark dataset, which was basically designed for age and gender classification.
- 2. Adience data set consists of images uploaded from smartphone devices.
- 3. Data set includes 26580 images with eight unbalanced age group classes.

## **RESULT**



### **REFERENCE**

- https://openaccess.thecvf.com/content\_ICCVW\_2019/papers/CEFRL/Angeloni\_Age\_Estimat ion\_From\_Facial\_Parts\_Using\_Compact\_Multi-Stream\_Convolutional\_Neural\_ICCVW\_2019\_paper.pdf
- http://dlib.net/files/data/dlib\_face\_detection\_dataset-2016-09-30.tar.gz

# **THANK YOU**