

## Facial Features Based Human Age, Gender And Ethnicity Identification System

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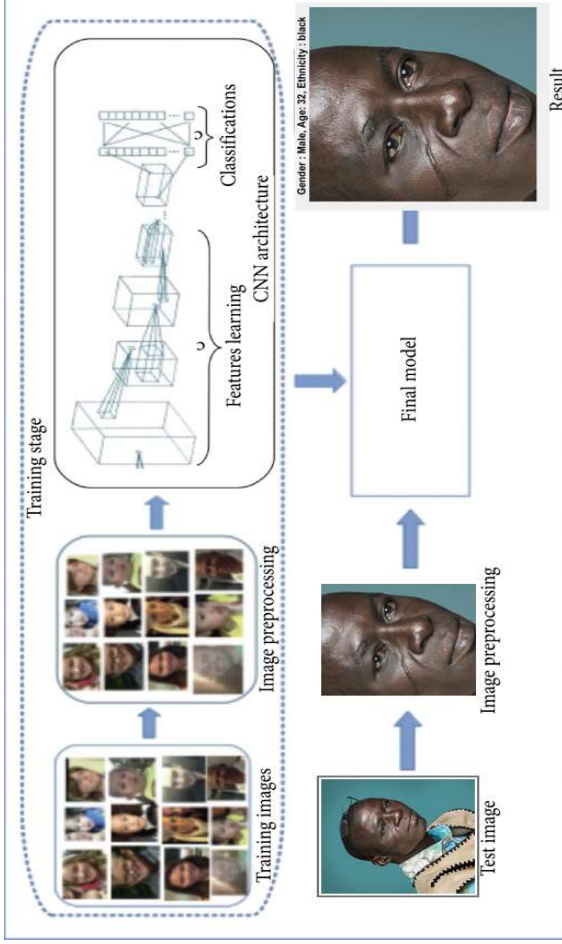
UG FINAL YEAR PROJECT  
JUNE 2022

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### Abstract

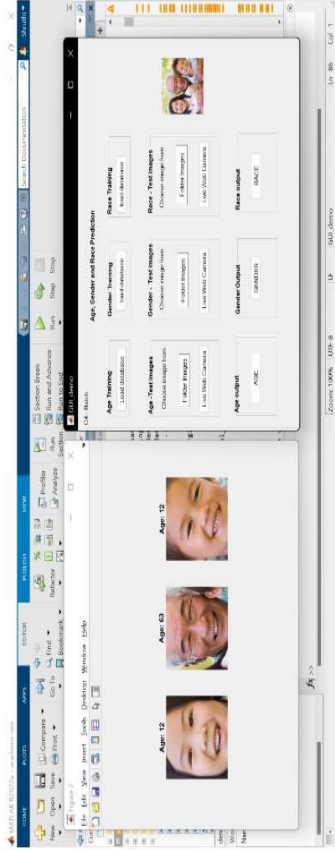
- Since the rise of social platforms and social media, automatic age, ethnicity, and gender classification have been relevant to a growing number of applications.
- Nevertheless, the performance of existing methods on real-world images is still significantly lacking.
- As a result, we show that learning representations using deep-convolution neural networks (CNN) can result in a considerable improvement in performance on **these** tasks.
- For image-based gender, age, and ethnicity estimates, deep neural networks with pre-trained weights are used.
- VGG is used to investigate transfer learning.
- Finally, a hierarchy of deep CNNs is explored, which first classifies participants by gender and then predicts age and ethnicity using separate models.

### Project Workflow

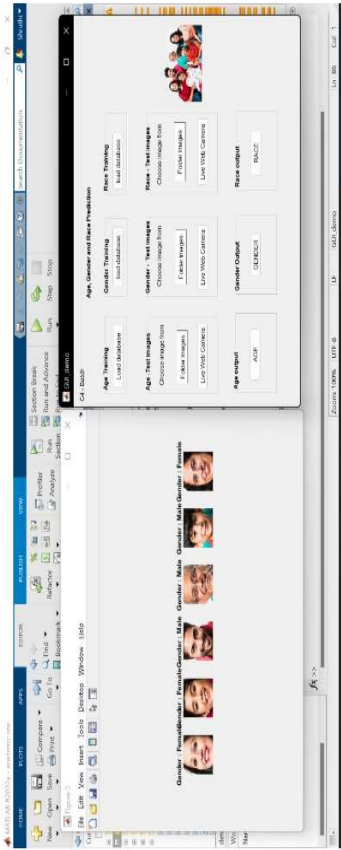


### Implementations

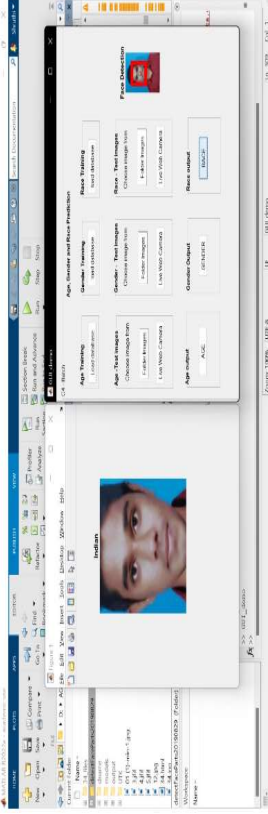
#### AGE



#### GENDER



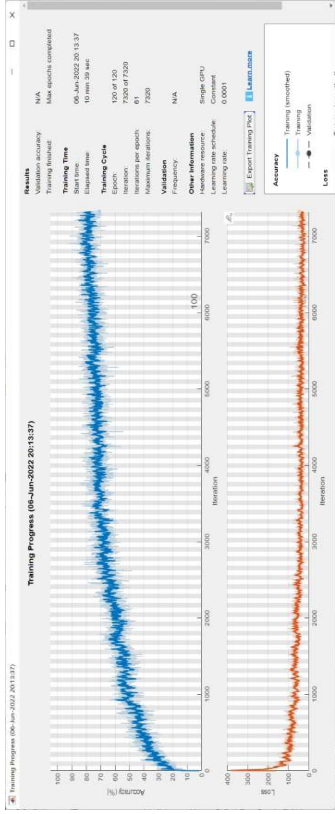
#### RACE



### Graphs

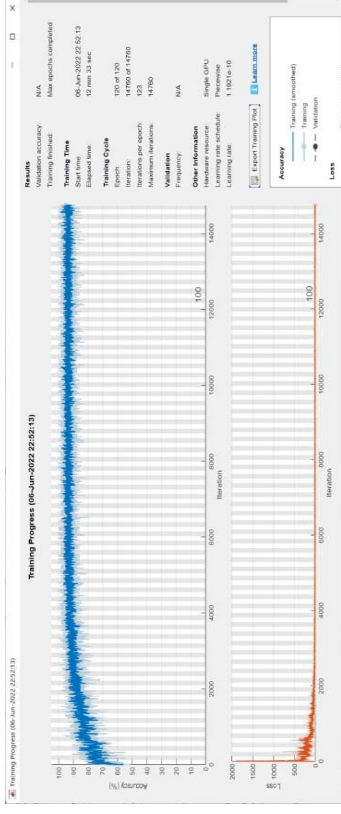
#### AGE

Training accuracy - 81%  
Validation error - 33.13%



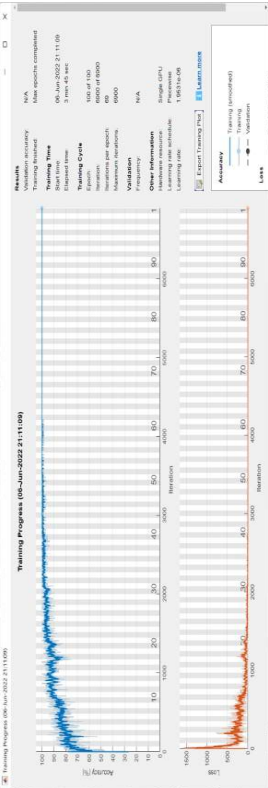
#### GENDER

Training accuracy - 91%  
Validation error - 8.16%



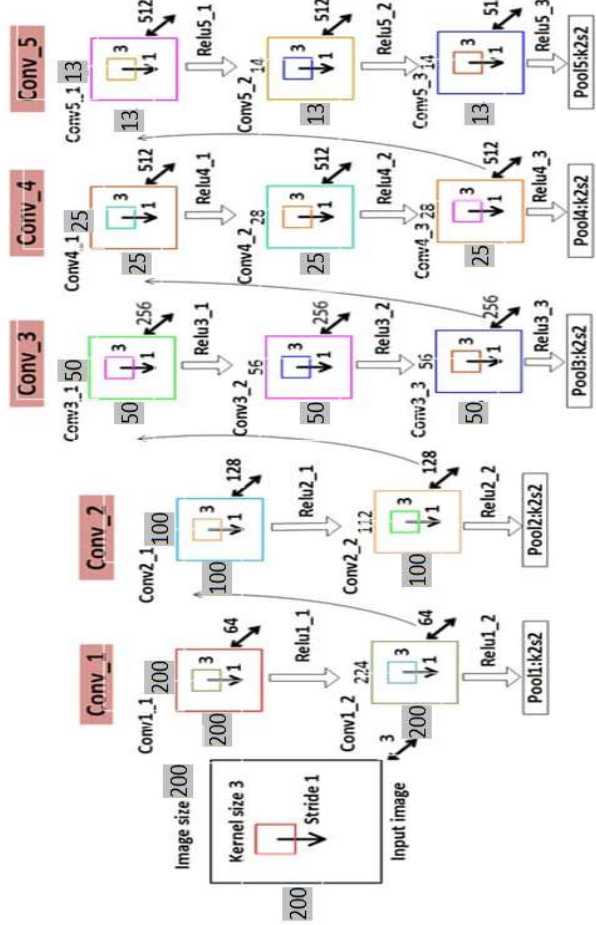
#### RACE

Training accuracy -98 %  
Validation error - 3.06%



### Proposed Architecture Diagram

VGG16 is a convolution neural net (CNN) architecture which is trained from scratch for facial recognition using VGG Face Dataset.



Here we use, Convolution layer of 3x3 kernel ,stride 1x1 and padding(1),starting with 200x200 input features. Max pool layer of 2x2 pool size and stride 2x2 and has no zero padding of the system.

We use the below formula to determine the number of output features from each layer.

$$n_{out} = \left\lfloor \frac{n_{in} + 2p - k}{s} \right\rfloor + 1$$

$n_{in}$ : number of input features

$n_{out}$ : number of output features

$k$ : convolution kernel size

$p$ : convolution padding size

$s$ : convolution stride size

### Conclusion and Future works

- The proposed approach utilizes a superior system performance in order to increase the estimation accuracy of demographic variables.
- It demonstrates that the suggested facial features-based Gender, Age, and Ethnicity estimation outperforms the majority of state-of-the-art techniques in UTK databases.
- In addition, the proposed method performed admirably in the difficult databases.
- In future different optimization and regularisation functions can be incorporated to the proposed system to increase the accuracy and also decrease the false prediction rate.

### References

- N. Kumar, A. Berg, P. Belhumeur, and S. Nayar. (2021), "Describable visual attributes for face verification and image search." IEEE Trans. PAMI, 33(10):1962–1977, 2021.
- Zhuang, Z., Landsittel, D., Benson, S., Roberge, R., Shaffer, R. (2019), "Facial anthropometric differences among gender, ethnicity, and age groups." Annals of Occupational Hygiene (2019)