

FAISAL KALEMA

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# A.G.E DETECTION

## (Age, Gender, Ethnicity)

— Neural Networks —

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# Agenda

- Data Collection
  - Features
- Data Cleaning
- Labeling
- Model
  - Model architecture
  - Customized Model
  - Training
- Model Evaluation
- Conclusion
- Further Improvements

## Main Objectives

- Classification of face images in three categories; Age, Gender, Ethnicity
- Leveraging pretrained model using transfer learning to enhance accuracy

# Data collection

Wikipedia



- ❑ 62, 328 face images from Wikipedia dataset with age and gender labels.
- ❑ Features include; face\_score, second\_face\_score, face\_location, name, gender, full\_path, photo\_taken, dob.

gettyimages® | 25 YEARS



- ❑ About 60,000 face images from gettyimages.com
- ❑ 100 pages
- ❑ Window (Before COVID-19)
- ❑ 1 person
- ❑ Headshot

# Data Cleaning

## Wiki Dataset

- ❑ Convert dob from Matlab serial date number format to Python datetime format.

## Drop observations with:

- ❑ wrong dob
- ❑ no faces (face\_score = inf)
- ❑ Second\_face\_score
- ❑ Face\_score < 3
- ❑ No gender information
- ❑ Age more than 100
- ❑ Age below 0
- ❑ Unnecessary columns

## Gettyimages Dataset

## Delete images with:

- ❑ No face
- ❑ Facemasks
- ❑ Black and white images
- ❑ More than 1 face
- ❑ Unclear face
- ❑ Eye glasses

# Labeling

## Wiki Dataset

### Age labels

- ☐ 0 to 100

### Gender labels

- ☐ 0 for Female
- ☐ 1 for Male

## Gettyimages Dataset

### 8 classes of Ethnicity

- ☐ Asian
- ☐ Black
- ☐ Caucasian
- ☐ Hispanic/Latino
- ☐ Mixed race
- ☐ Multi Ethnic group
- ☐ Native American
- ☐ Pacific islander

# Model

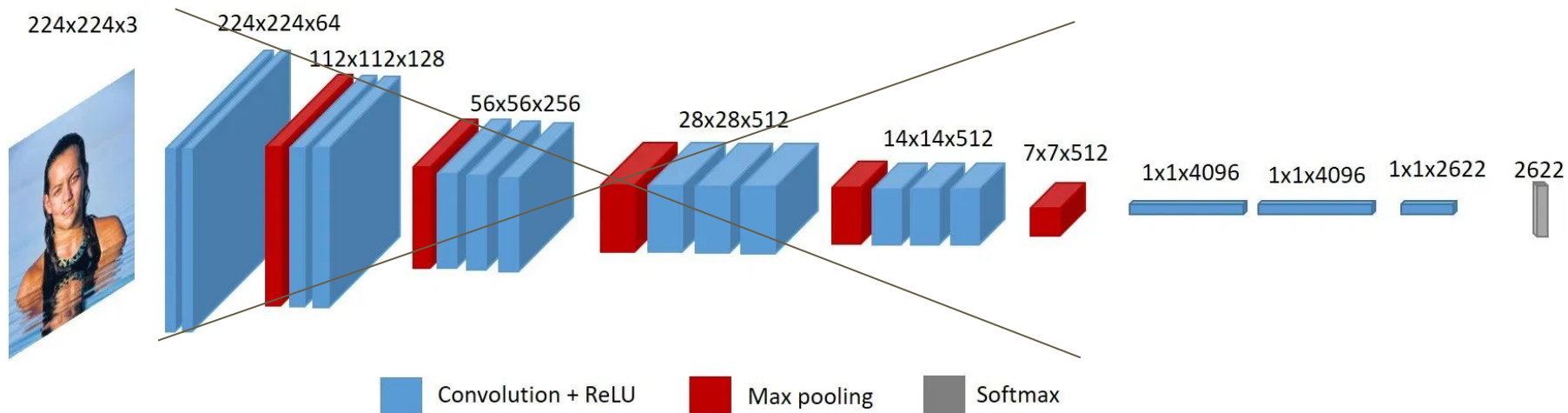
VGG-Face Model (22 layers) pre-trained on ImageNet

Pre\_trained weights

By Oxford Visual Geometry Group



# Customized Model



# Model Architecture

1. Input Image



2. Face Detection



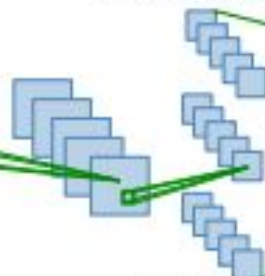
Mathias et al. detector

3. Cropped face



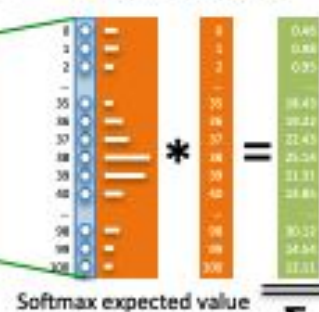
+ 40% margin

4. Feature Extraction



VGG-16 architecture

5. Prediction





# Training

	Classification type	Loss Function	Optimizer	Activation
<b>Age</b>	Multi-class	Categorical cross-entropy	Adam	Softmax
<b>Gender</b>	Binary	Binary cross-entropy	Adam	Sigmoid
<b>Ethnicity</b>	Multi-class	Categorical cross-entropy	Adam	Softmax

# Model Evaluation

	Validation loss	Accuracy
Age	3.30	0.07
Gender	0.09	0.98
Ethnicity	1.23	0.55

**Age**

**MAE = 5.4**

## Evaluation of Model on gender

	Predict Female	Predict Male
Actual Female	1283	27
Actual Male	32	3086

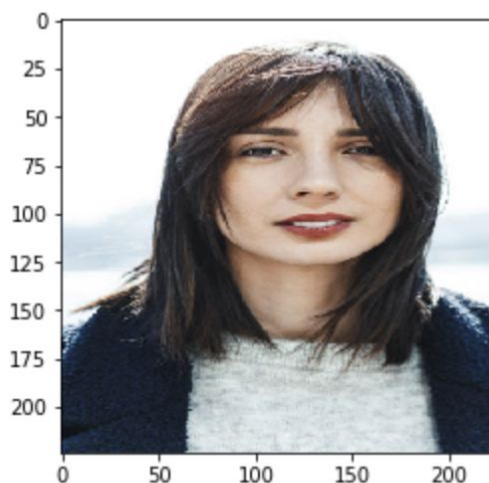
Precision: 0.979

Recall : 0.975

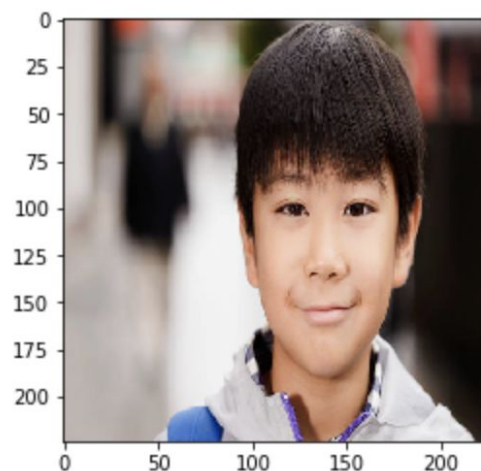
## Ethnicity results



Actual: asian  
Predicted: asian

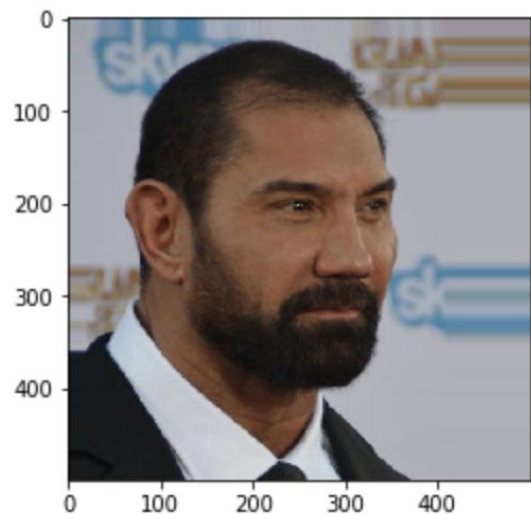


Actual: asian  
Predicted: caucasian

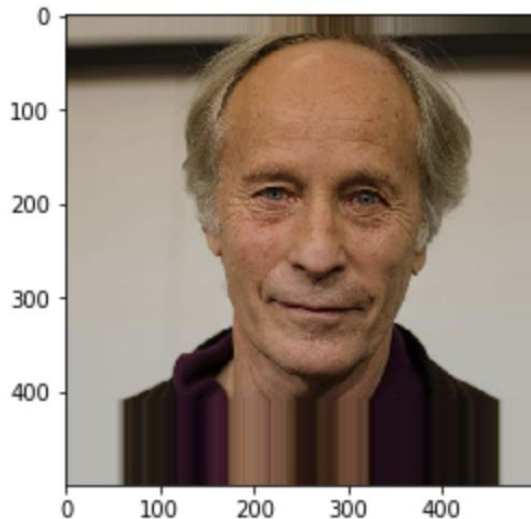


Actual: asian  
Predicted: asian

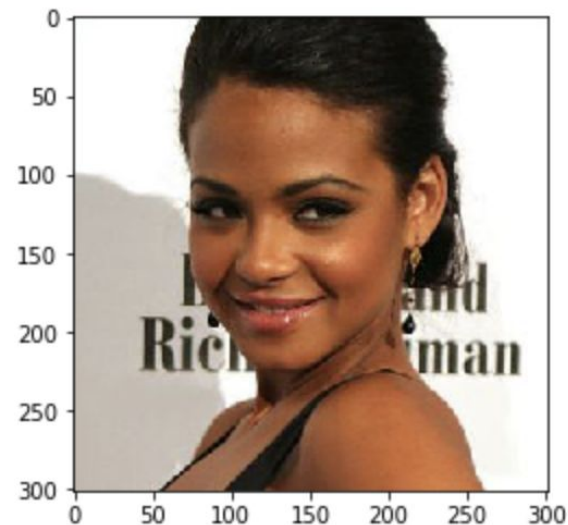
## Age results



Predicted : 44  
Actual Age: 44

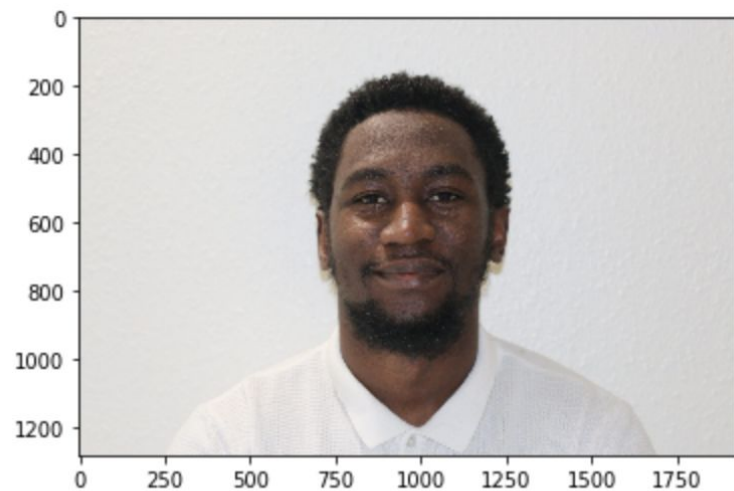


Predicted : 63  
Actual Age: 62



Predicted : 25  
Actual Age: 28

## Gender results



gender: Male



gender: Male

# Further Improvements

- More data to avoid unbalanced classes
- Build a Multi-Output Model