



Faculty of Computers and Artificial Intelligence
Computer Science Department
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# CS 396 Selected Topics in CS-2 Research Project

Report Submitted for Fulfillment of the Requirements and ILO's for Selected Topics in CS-2 course for Fall 2021

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#### Delivered to:

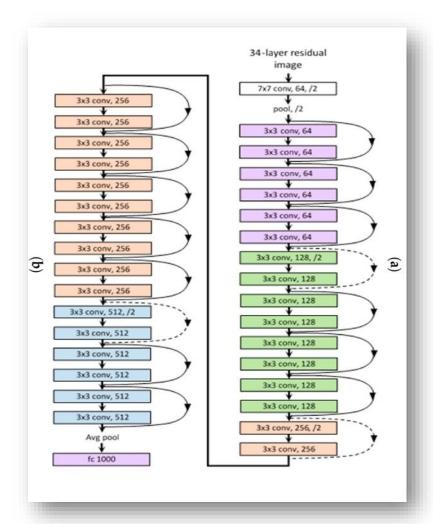
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## **Architecture used in the paper: -**

- 1-The paper used a residual neural network(resnet34).
- 2-A ResNet can be called an upgraded version of the VGG architecture, with the difference between them being the skip connections used in ResNets. In the figure below, we can see the architecture of the VGG as well as the 34 layer ResNet.
- 4-You might wonder why we have taken padding as 'same' only for all Convolution layers. The reason behind this is, we have to maintain the shape of our input until we add it to the residue. If the shape of the input gets changed, we will get a Numpy Error saying- "Two arrays with different shapes cannot be added".
- 5-Some of the points to note in this convolution block are, the residue is not directly added to the output but is passed through a Convolution Layer
- 3-The image size 224 x 224 pixels.



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

 $n_{in}$ : number of input features  $n_{out}$ : number of output features

k: convolution kernel size

p: convolution padding size

s: convolution stride size

## Algorithm for Convolutional Block

X\_skip = Input

Convolutional Layer (3X3) (Strides = 2) (Filters = f) (Padding = 'same') →(Input)

Batch Normalisation →(Input)

Relu Activation →(Input)

Convolutional Layer (3X3) (Filters = f) (Padding = 'same') →(Input)

Batch Normalisation →(Input)

Convolutional Layer (1X1) (Filters = f) (Strides = 2) →(X\_skip)

Add (Input + X\_skip)

Relu Activation

# Algorithm for Identity Block

X\_skip = Input

Convolutional Layer (3X3) (Padding='same') (Filters = f) →(Input)

Batch Normalisation →(Input)

Relu Activation →(Input)

Convolutional Layer (3X3) (Padding = 'same') (Filters = f) →(Input)

Batch Normalisation →(Input)

Add (Input + X\_skip)

Relu Activation

## ResNet-34 Structure and Code

| layer name | 34-layer  | 50-layer   | 101-layer   |  |  |
|------------|---|--|---|--|--|
| conv1      |   | 7 × 7,64, stride 2   |   |  |  |
|            | 3 × 3 max pool, stride 2  |  |   |  |  |
| conv2_x    | $\begin{bmatrix} 3 \times 3,64 \\ 3 \times 3,64 \end{bmatrix} \times 3$   | $\begin{bmatrix} 1 \times 1,64 \\ 3 \times 3,64 \\ 1 \times 1,256 \end{bmatrix} \times 3$    | $\begin{bmatrix} 1 \times 1,64 \\ 3 \times 3,64 \\ 1 \times 1,256 \end{bmatrix} \times 3$     |  |  |
| conv3_x    | $\begin{bmatrix} 3 \times 3,128 \\ 3 \times 3,128 \end{bmatrix} \times 4$ | $\begin{bmatrix} 1 \times 1,128 \\ 3 \times 3,128 \\ 1 \times 1,512 \end{bmatrix} \times 4$  | $\begin{bmatrix} 1 \times 1,128 \\ 3 \times 3,128 \\ 1 \times 1,512 \end{bmatrix} \times 4$   |  |  |
| conv4_x    | $\begin{bmatrix} 3 \times 3,256 \\ 3 \times 3,256 \end{bmatrix} \times 6$ | $\begin{bmatrix} 1 \times 1,256 \\ 3 \times 3,256 \\ 1 \times 1,1024 \end{bmatrix} \times 6$ | $\begin{bmatrix} 1 \times 1,256 \\ 3 \times 3,256 \\ 1 \times 1,1024 \end{bmatrix} \times 23$ |  |  |
| conv5_x    | $\begin{bmatrix} 3 \times 3,512 \\ 3 \times 3,512 \end{bmatrix} \times 3$ | $\begin{bmatrix} 1 \times 1,512 \\ 3 \times 3,512 \\ 1 \times 1,2048 \end{bmatrix} \times 3$ | $\begin{bmatrix} 1 \times 1,512 \\ 3 \times 3,512 \\ 1 \times 1,2048 \end{bmatrix} \times 3$  |  |  |
|            | average pool,2048-d fc  |  |   |  |  |

## **Dataset details: -**

- 1-this project use an Audience data set created to use to Predict age's Person,
- 2-the Dataset contain image of Person's Faces,
- 3-the Dataset Contain 18000 image between women, men, Childs, adults, youth, babies, elderly.
- 4-classify the dataset into 18 Class
- 5-the Dataset contain people of age range (1-100)
- 6-The images cover large variation in pose, facial expression, illumination, occlusion, resolution



### Labels:

The labels of each face image are embedded in the file name.

• [age] is an integer from 1 to 100, indicating the age

## **Implementation details: -**

We have implemented the code using tensorflow library & the tensorflow pipeline to preprocess the date then we used the famous resnet 34 architecture that have been published in 2015 then started building the model then we compiled it using the adam optimizer with initial learning rate of .0002, sparse categorical cross entropy, 50 epoch & verbose value of 2 to provide info about our model through its training process ( use the link provided here to see for yourself )

https://colab.research.google.com/drive/1bD\_EjirKmbR HIP-C1AtohMoKj3tIm7G\_?usp=sharing