
Code: 2023FS-COMP_SCI-5567-0001 Deep Learning

Mini_project-2

Name: Ramya Kumari Thambabattula

Student ID: 16342145

Analysis for trainNetwork usage

Name: net

Analysis date: 01-Dec-2023 17:30:27

143.6M

total learnables

47

layers

0

warnings

0

errors

ANALYSIS RESULT

	Name	Type	Activations	Learnable Proper...	St
1	input 224x224x3 images with 'zerocenter' nor...	Image Input	$224(S) \times 224(S) \times 3(C) \times 1(B)$	-	-
2	conv1_1 64 3x3x3 convolutions with stride [1 1] a...	2-D Convolution	$224(S) \times 224(S) \times 64(C) \times 1(B)$	Weig... $3 \times 3 \times 3 \dots$ Bias $1 \times 1 \times 64$	-
3	relu1_1 ReLU	ReLU	$224(S) \times 224(S) \times 64(C) \times 1(B)$	-	-
4	conv1_2 64 3x3x64 convolutions with stride [1 1] ...	2-D Convolution	$224(S) \times 224(S) \times 64(C) \times 1(B)$	Weig... $3 \times 3 \times 64\dots$ Bias $1 \times 1 \times 64$	-
5	relu1_2 ReLU	ReLU	$224(S) \times 224(S) \times 64(C) \times 1(B)$	-	-
6	pool1 2x2 max pooling with stride [2 2] and pa...	2-D Max Pooling	$112(S) \times 112(S) \times 64(C) \times 1(B)$	-	-
7	conv2_1 128 3x3x64 convolutions with stride [1 1]...	2-D Convolution	$112(S) \times 112(S) \times 128(C) \times 1(B)$	Weig... $3 \times 3 \times 64 \dots$ Bias $1 \times 1 \times 128$	-
8	relu2_1 ReLU	ReLU	$112(S) \times 112(S) \times 128(C) \times 1(B)$	-	-
9	conv2_2 128 3x3x128 convolutions with stride [1 ...	2-D Convolution	$112(S) \times 112(S) \times 128(C) \times 1(B)$	Weig... $3 \times 3 \times 128\dots$ Bias $1 \times 1 \times 128$	-
10	relu2_2 ReLU	ReLU	$112(S) \times 112(S) \times 128(C) \times 1(B)$	-	-
11	pool2 2x2 max pooling with stride [2 2] and pa...	2-D Max Pooling	$56(S) \times 56(S) \times 128(C) \times 1(B)$	-	-
12	conv3_1 256 3x3x128 convolutions with stride [1 ...	2-D Convolution	$56(S) \times 56(S) \times 256(C) \times 1(B)$	Weig... $3 \times 3 \times 128\dots$ Bias $1 \times 1 \times 256$	-
13	relu3_1 ReLU	ReLU	$56(S) \times 56(S) \times 256(C) \times 1(B)$	-	-
14	conv3_2 256 3x3x256 convolutions with stride [1 ...	2-D Convolution	$56(S) \times 56(S) \times 256(C) \times 1(B)$	Weig... $3 \times 3 \times 256\dots$ Bias $1 \times 1 \times 256$	-
15	relu3_2 ReLU	ReLU	$56(S) \times 56(S) \times 256(C) \times 1(B)$	-	-

input

conv1_1

relu1_1

conv1_2

relu1_2

pool1

conv2_1

relu2_1

conv2_2

relu2_2

pool2

conv3_1

relu3_1

conv3_2

relu3_2

conv3_3

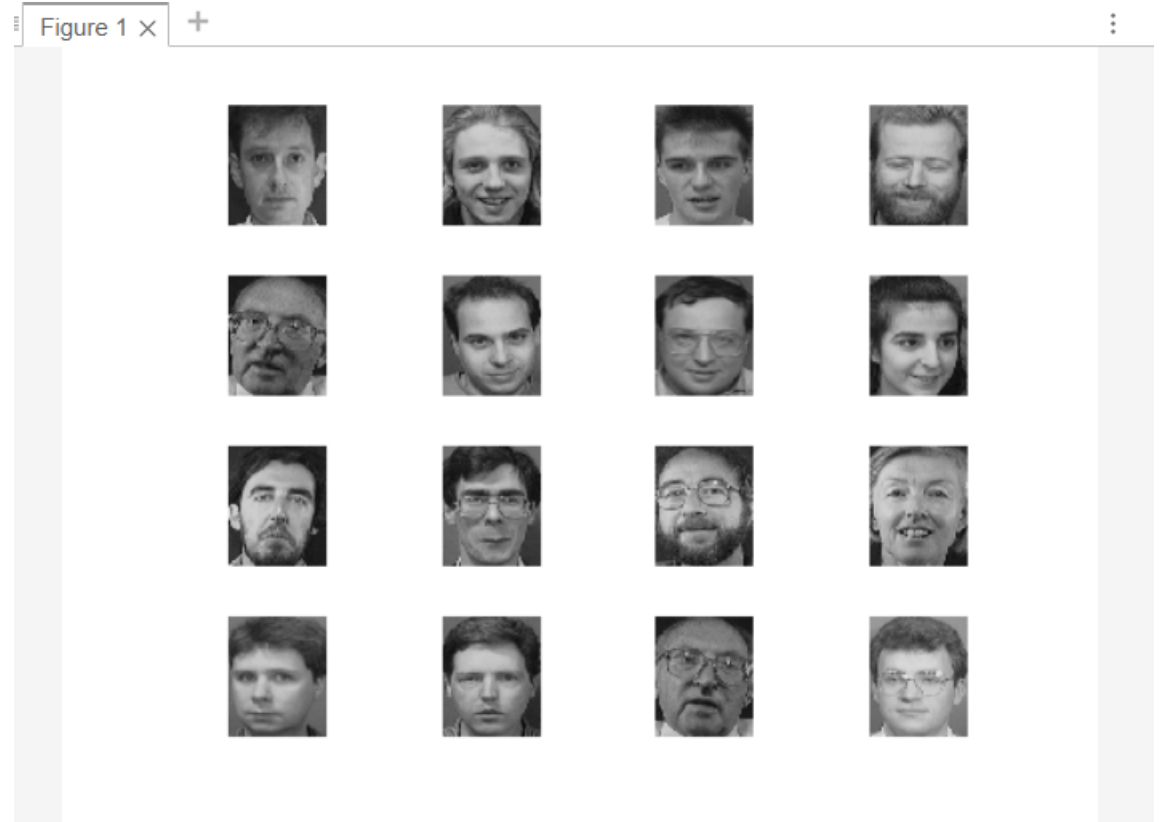
relu3_3

conv3_4

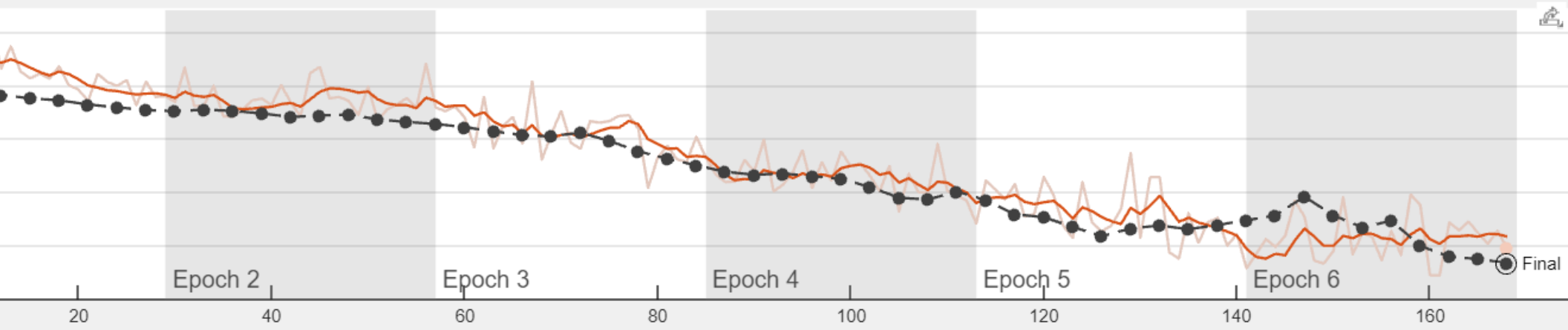
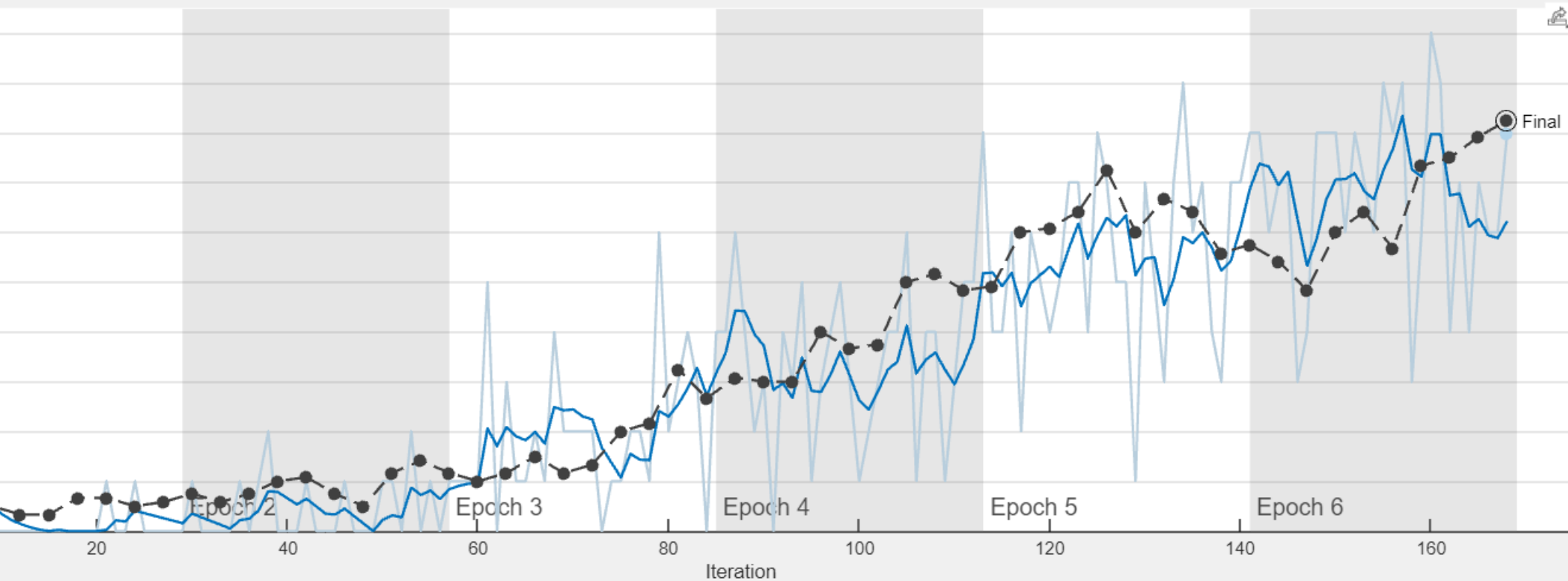
relu3_4

Download the Vgg19 On
Addons. Get Face
Recognition using AlexNet

```
'MiniBatchSize', 10, ...  
'MaxEpochs', 6, ...  
'InitialLearnRate', 1e-4, ...  
'Shuffle', 'every-epoch', ...  
'ValidationData',  
augimdsValidation, ...  
'ValidationFrequency', 3,  
...  
'Verbose', false, ...  
'Plots', 'training-  
progress');
```



Training Progress (01-Dec-2023 20:12:55)



Results

Validation accuracy: 82.50%

Training finished: Max ep

Training Time

Start time: 01-Dec-

Elapsed time: 20 min

Training Cycle

Epoch: 6 of 6

Iteration: 168 of 1

Iterations per epoch: 28

Maximum iterations: 168

Validation

Frequency: 3 iterati

Other Information

Hardware resource: Single C

Learning rate schedule: Const

Learning rate: 0.0001

[Learn more](#)

Accuracy

— Training (smooth

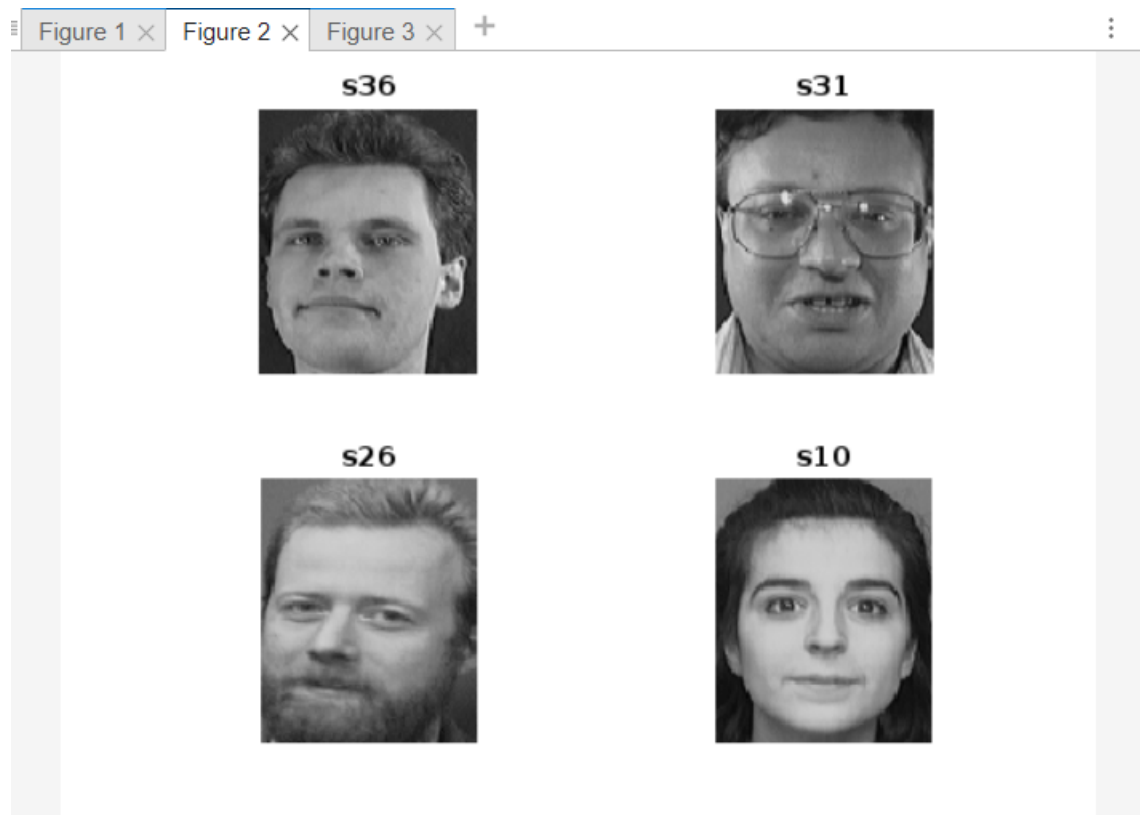
— Training

— Validation

Loss

— Training (smooth

Face Recognition using AlexNet: Validated Images



Results	
Validation accuracy:	82.50%
Training finished:	Max epochs completed
Training Time	
Start time:	01-Dec-2023 20:12:55
Elapsed time:	20 min 41 sec
Training Cycle	
Epoch:	6 of 6
Iteration:	168 of 168
Iterations per epoch:	28
Maximum iterations:	168
Validation	
Frequency:	3 iterations
Other Information	
Hardware resource:	Single CPU
Learning rate schedule:	Constant
Learning rate:	0.0001

Workspace

Files				
Workspace				
Name	Value	Size	Class	
Cos	0.4679	1x1	single	
Genuine1	1x40 single	1x40	single	
Genuine2	1x40 single	1x40	single	
I	112x92 uint8	112x92	uint8	
X	41x1 single	41x1	single	
Y	41x1 single	41x1	single	
YPred	120x1 cate...	120x1	categorical	
YValidat...	120x1 cate...	120x1	categorical	
a	1x4096 single	1x4096	single	
accuracy	0.8250	1x1	double	
augimd...	1x1 augme...	1x1	augmentedl...	
augimd...	1x1 augme...	1x1	augmentedl...	
features	120x4096 s...	120x4096	single	
gset1	1x40 logical	1x40	logical	
histogra...	1x1 Histogr...	1x1	matlab.grap...	
histogra...	1x1 Histogr...	1x1	matlab.grap...	
i	4	1x1	double	
idx	[88,75,55,4]	1x4	double	
imageA...	1x1 image...	1x1	imageData...	
imds	1x1 Image...	1x1	matlab.io.d...	
imdsTrain	1x1 Image...	1x1	matlab.io.d...	
imdsVali...	1x1 Image...	1x1	matlab.io.d...	

Name	Value	Size	Class	
imposter1	1x40 single	1x40	single	
imposter2	1x40 single	1x40	single	
inputSize	[224,224,3]	1x3	double	
label	s10	1x1	categorical	
layering	'fc7'	1x3	char	
layers	47x1 Layer	47x1	nnet.cnn.la...	
layersTr...	44x1 Layer	44x1	nnet.cnn.la...	
na	70.8954	1x1	single	
net	1x1 Series...	1x1	SeriesNetw...	
netTran...	1x1 Series...	1x1	SeriesNetw...	
numCla...	40	1x1	double	
numTrai...	280	1x1	double	
nz	65.1566	1x1	single	
options	1x1 Trainin...	1x1	nnet.cnn.Tr...	
pixelRa...	[-30,30]	1x2	double	
set1	40x4096 si...	40x4096	single	
set2	40x4096 si...	40x4096	single	
set3	40x4096 si...	40x4096	single	
t	0.7000	1x1	double	
threshold	0.8000	1x1	double	
z	1x4096 single	1x4096	single	
za	2.1613e+03	1x1	single	

Analysis Face Recognition using AlexNet. Training the network with modified data, Training Parameters 9 Epochs, Mini batch size 12

TurboVNC: ml-teaching-00-04: x +

ood.umkc.edu/pun/sys/dashboard/noVNC-1.3.0/vnc.html?autoconnect=true&path=rnode%2Fml-teaching-00-04%2F40969%2Fwebsocketify&resize=remote&password=P9Pdd5...

MATLAB R2023a - acad... Deep Learning Network... [Training Progress (30-...

01:52:40 AM

Deep Learning Network Analyzer

Analysis for trainNetwork usage

Name: net

Analysis date: 30-Nov-2023 01:50:42

60.9M total learnables 25 layers 0 warnings 0 errors

data

conv1

relu1

norm1

pool1

conv2

relu2

norm2

pool2

conv3

relu3

conv4

relu4

conv5

relu5

pool5

fc6

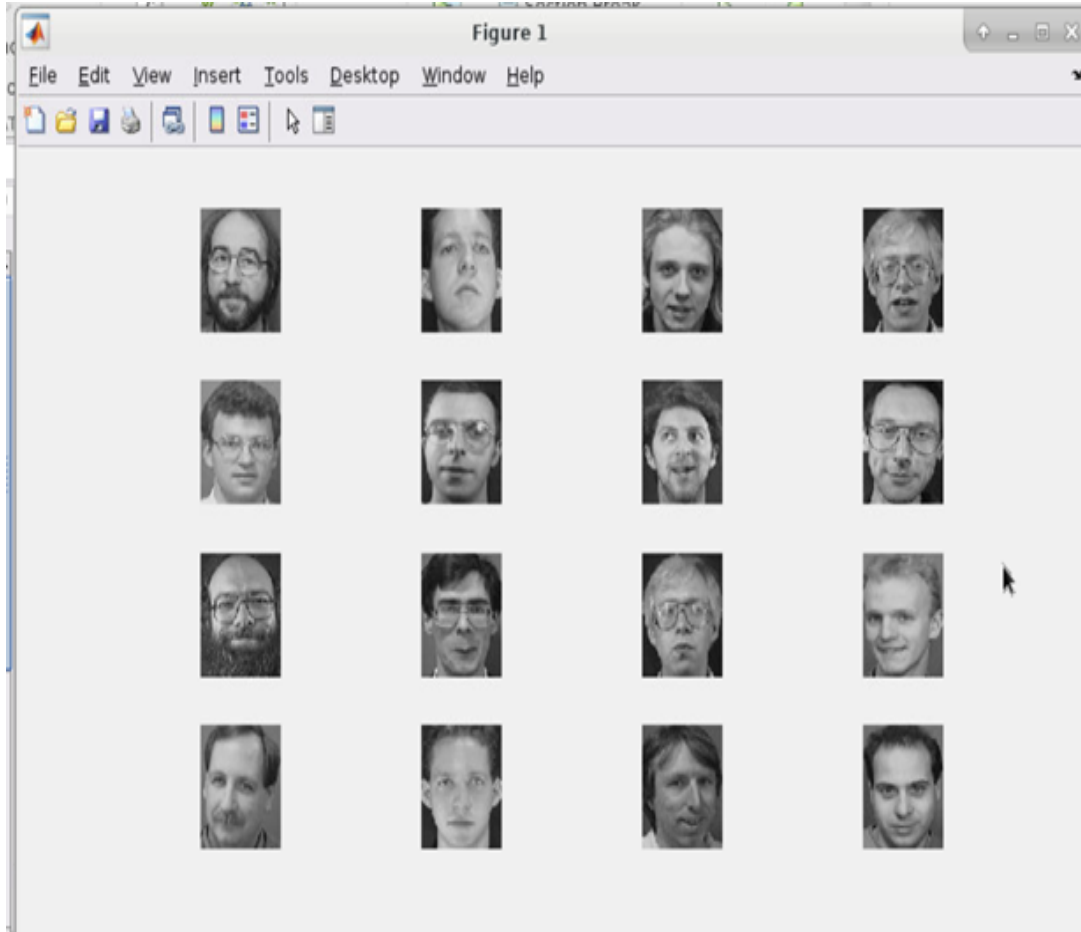
ANALYSIS RESULT

	Name	Type	Activations	Learnable Prope...	States
1	data 227×227×3 Images with 'zerocenter' nor...	Image Input	227(S) × 227(S) × 3(C) × 1(B)	-	-
2	conv1 96 11×11×3 convolutions with stride [4 4...	2-D Convolution	55(S) × 55(S) × 96(C) × 1(B)	Wei... 11 × 11 × ... Bias 1 × 1 × 96	-
3	relu1 ReLU	ReLU	55(S) × 55(S) × 96(C) × 1(B)	-	-
4	norm1 cross channel normalization with 5 chan...	Cross Channel Nor...	55(S) × 55(S) × 96(C) × 1(B)	-	-
5	pool1 3×3 max pooling with stride [2 2] and pa...	2-D Max Pooling	27(S) × 27(S) × 96(C) × 1(B)	-	-
6	conv2 2 groups of 128 5×5×48 convolutions wi...	2-D Grouped Conv...	27(S) × 27(S) × 256(C) × 1(B)	Wei... 5 × 5 × 48... Bi... 1 × 1 × 12...	-
7	relu2 ReLU	ReLU	27(S) × 27(S) × 256(C) × 1(B)	-	-
8	norm2 cross channel normalization with 5 chan...	Cross Channel Nor...	27(S) × 27(S) × 256(C) × 1(B)	-	-
9	pool2 3×3 max pooling with stride [2 2] and pa...	2-D Max Pooling	13(S) × 13(S) × 256(C) × 1(B)	-	-
10	conv3 384 3×3×256 convolutions with stride [1 ...	2-D Convolution	13(S) × 13(S) × 384(C) × 1(B)	Wei... 3 × 3 × 25... Bias 1 × 1 × 384	-
11	relu3 ReLU	ReLU	13(S) × 13(S) × 384(C) × 1(B)	-	-
12	conv4 2 groups of 192 3×3×192 convolutions ...	2-D Grouped Conv...	13(S) × 13(S) × 384(C) × 1(B)	Wei... 3 × 3 × 192... Bi... 1 × 1 × 192...	-
13	relu4 ReLU	ReLU	13(S) × 13(S) × 384(C) × 1(B)	-	-

8°C Mostly cloudy

Search

01:52 30-11-2023

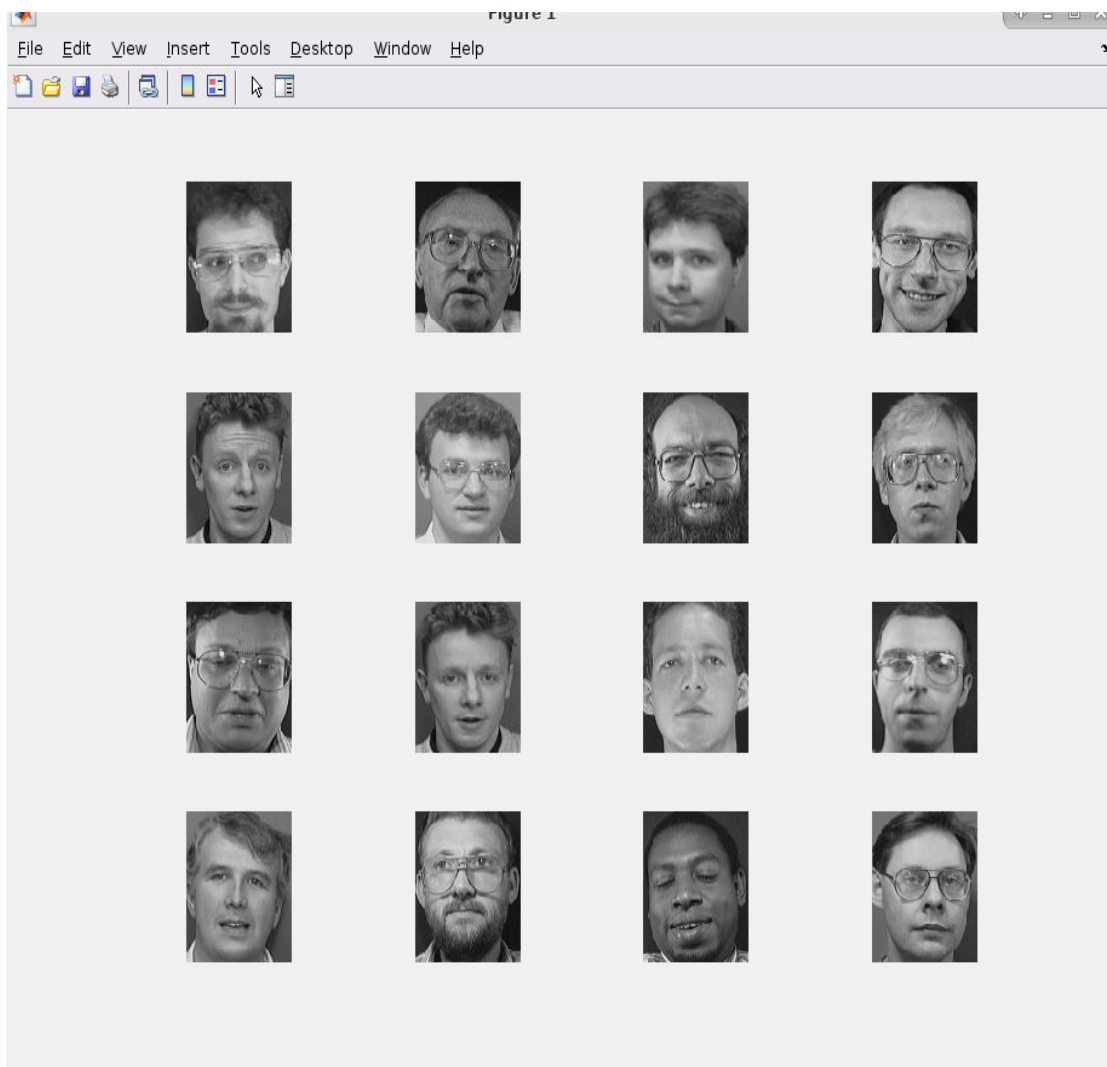


Training the network with
modified settings

Epochs: 6 ; Batch size: 10; Validation accuracy: 71.67%. And default settings indicate the model achieved 71.67% accuracy on the validation set with limited training epochs and a moderate batch size.

Modified data Epochs: 9 ; Batch size: 12; Validation accuracy: 85%

Changing the settings to 9 epochs and a batch size of 12 resulted in a substantial increase in validation accuracy to 85%. The improvement from 71.67% to 85% in validation accuracy highlights the effectiveness of the changes.



Second Part: VGG19 Face Recognition

Analysis for trainNetwork usage

Name: net

Analysis date: 01-Dec-2023 21:41:56

143.6M

total learnables

47

layers

0

warnings

0

errors

ANALYSIS RESULT

	Name	Type	Activations	Learnable Proper...	St
1	input 224x224x3 images with 'zerocenter' nor...	Image Input	224(S) × 224(S) × 3(C) × 1(B)	-	-
2	conv1_1 64 3x3x3 convolutions with stride [1 1] a...	2-D Convolution	224(S) × 224(S) × 64(C) × 1(B)	Weig... 3 × 3 × 3 ... Bias 1 × 1 × 64	-
3	relu1_1 ReLU	ReLU	224(S) × 224(S) × 64(C) × 1(B)	-	-
4	conv1_2 64 3x3x64 convolutions with stride [1 1] ...	2-D Convolution	224(S) × 224(S) × 64(C) × 1(B)	Weig... 3 × 3 × 64... Bias 1 × 1 × 64	-
5	relu1_2 ReLU	ReLU	224(S) × 224(S) × 64(C) × 1(B)	-	-
6	pool1 2x2 max pooling with stride [2 2] and pa...	2-D Max Pooling	112(S) × 112(S) × 64(C) × 1(B)	-	-
7	conv2_1 128 3x3x64 convolutions with stride [1 1]...	2-D Convolution	112(S) × 112(S) × 128(C) × 1(B)	Weig... 3 × 3 × 64 ... Bias 1 × 1 × 128	-
8	relu2_1 ReLU	ReLU	112(S) × 112(S) × 128(C) × 1(B)	-	-
9	conv2_2 128 3x3x128 convolutions with stride [1 ...	2-D Convolution	112(S) × 112(S) × 128(C) × 1(B)	Weig... 3 × 3 × 128... Bias 1 × 1 × 128	-
10	relu2_2 ReLU	ReLU	112(S) × 112(S) × 128(C) × 1(B)	-	-
11	pool2 2x2 max pooling with stride [2 2] and pa...	2-D Max Pooling	56(S) × 56(S) × 128(C) × 1(B)	-	-
12	conv3_1 256 3x3x128 convolutions with stride [1 ...	2-D Convolution	56(S) × 56(S) × 256(C) × 1(B)	Weig... 3 × 3 × 128... Bias 1 × 1 × 256	-
13	relu3_1 ReLU	ReLU	56(S) × 56(S) × 256(C) × 1(B)	-	-
14	conv3_2 256 3x3x256 convolutions with stride [1 ...	2-D Convolution	56(S) × 56(S) × 256(C) × 1(B)	Weig... 3 × 3 × 256... Bias 1 × 1 × 256	-

input

conv1_1

relu1_1

conv1_2

relu1_2

pool1

conv2_1

relu2_1

conv2_2

relu2_2

pool2

conv3_1

relu3_1

conv3_2

relu3_2

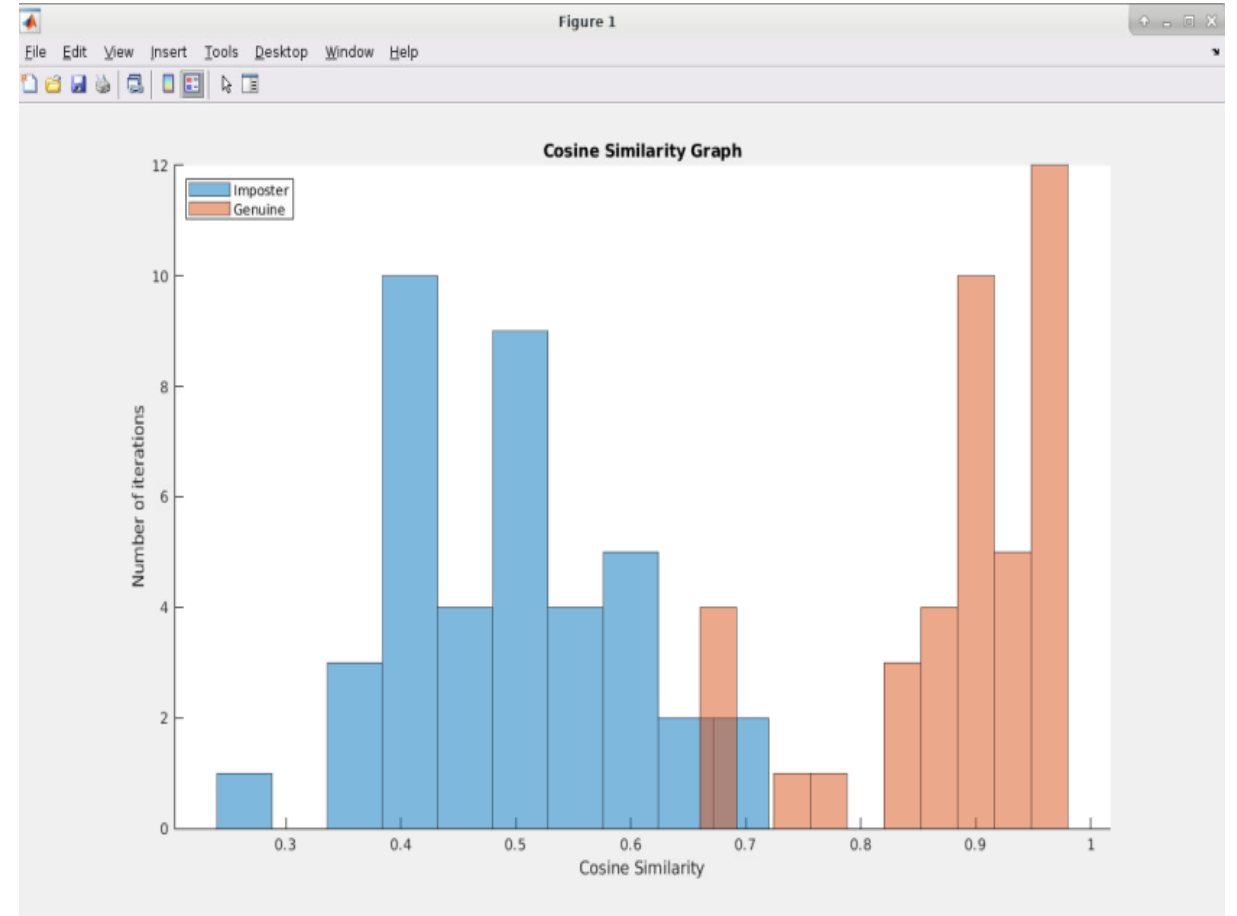
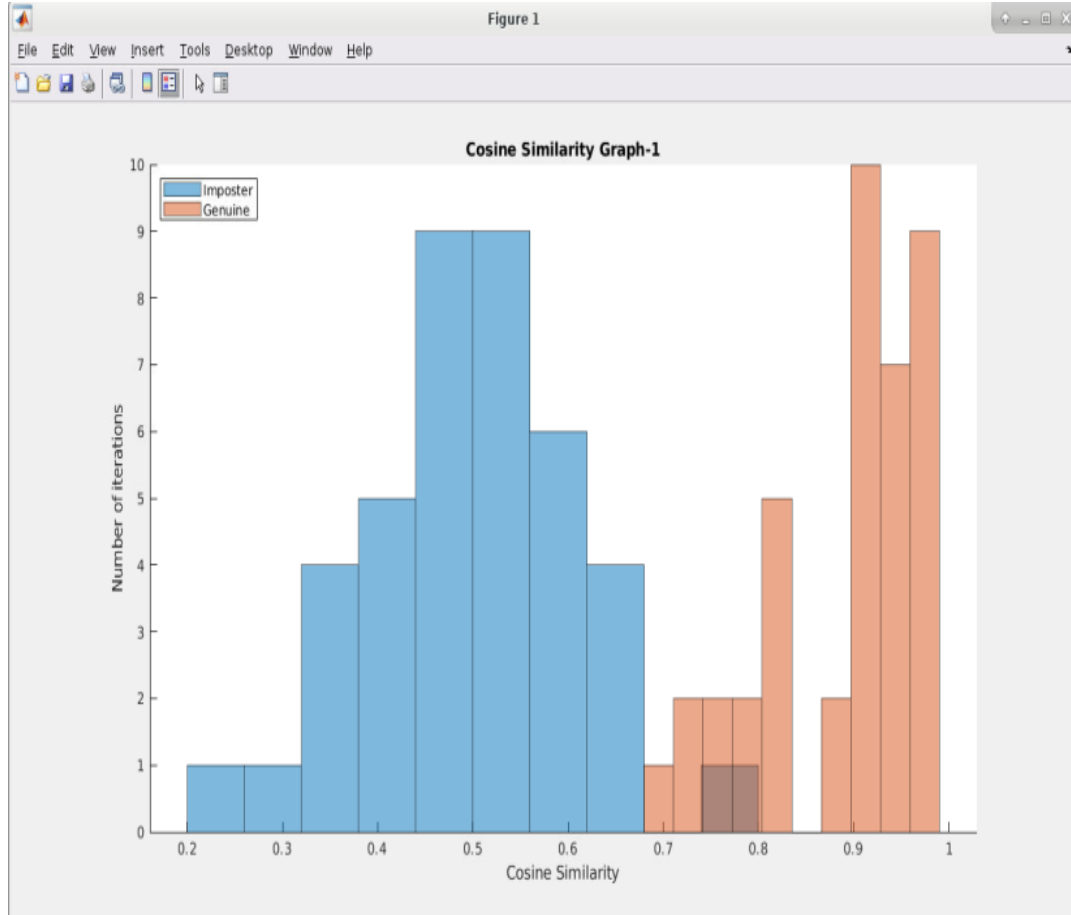
conv3_3

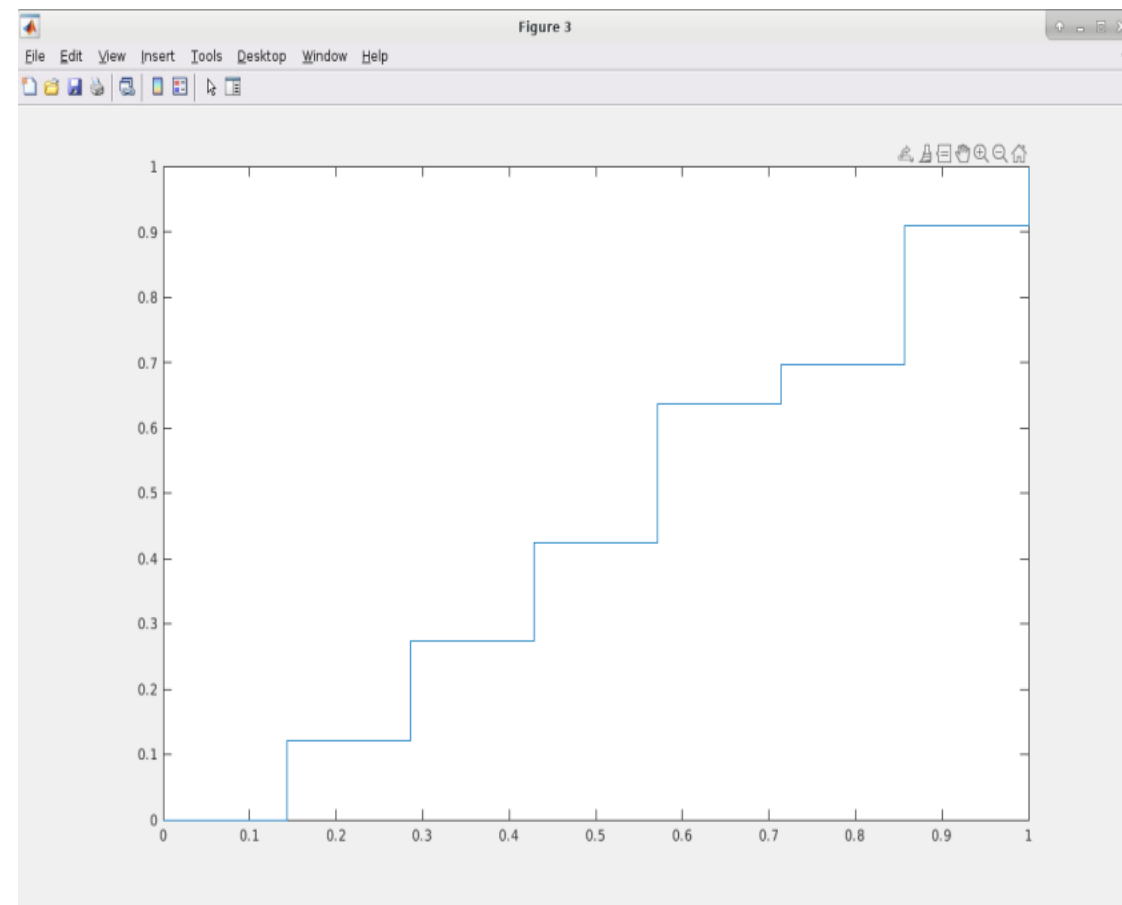
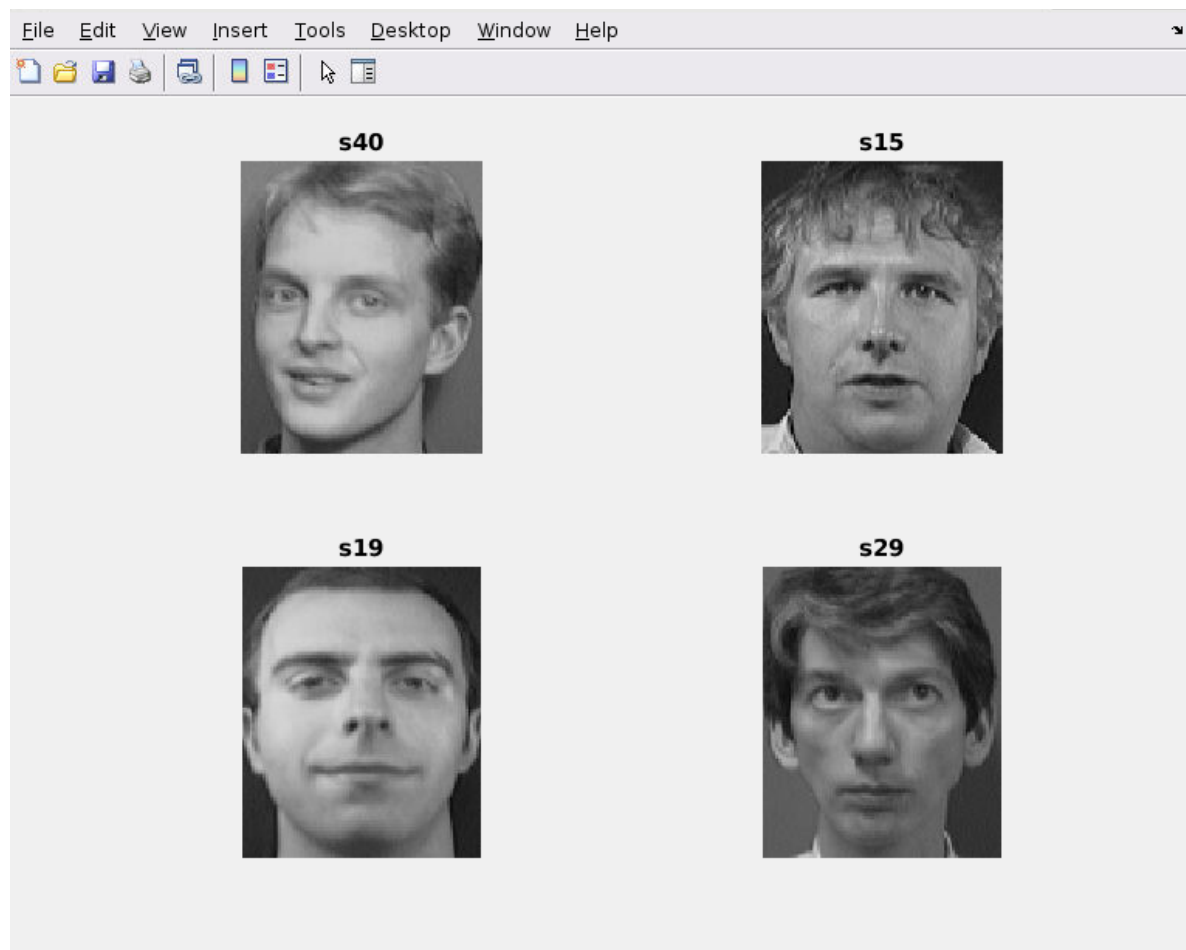
relu3_3

conv3_4

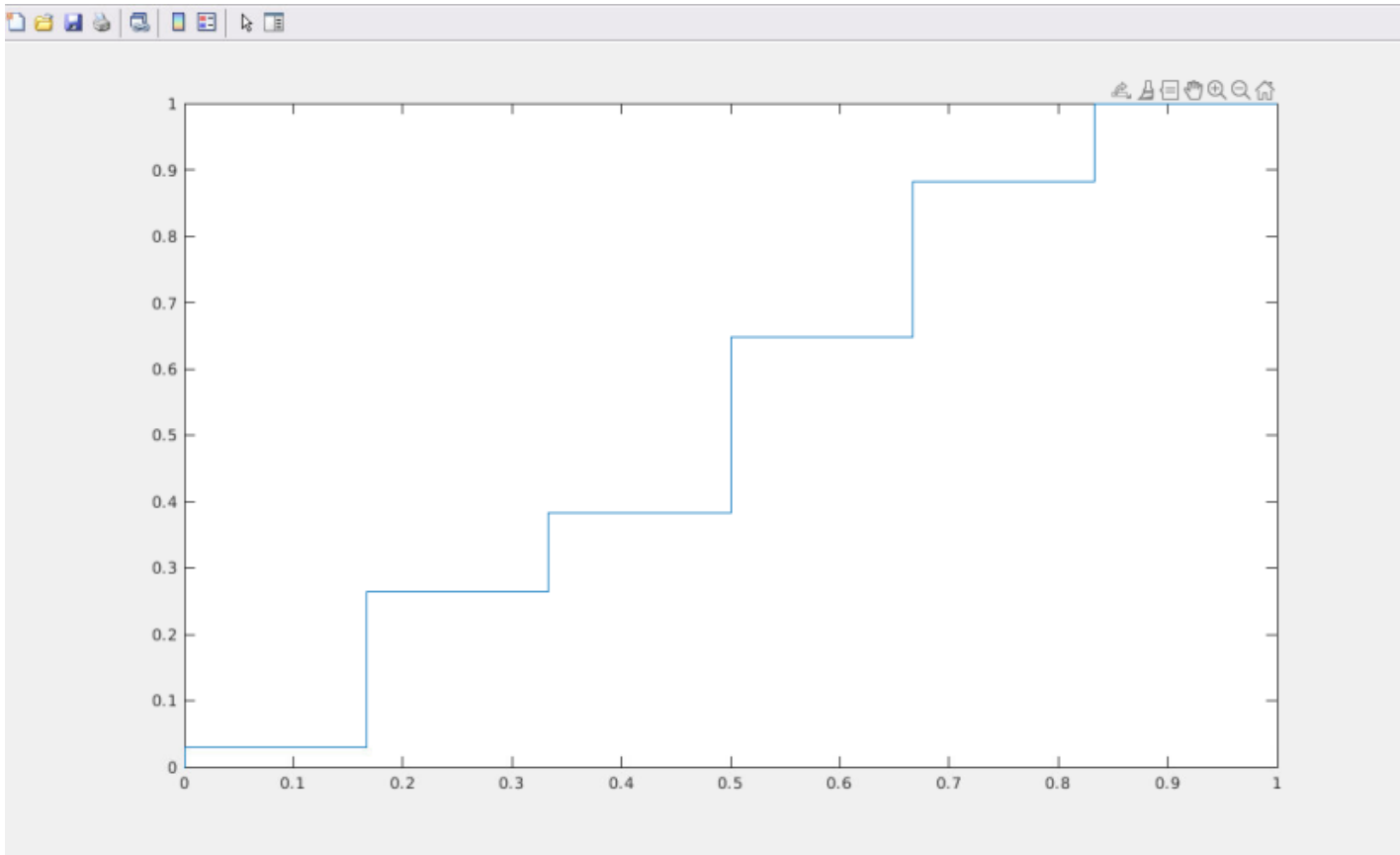
ings
work
ameters
nes ; Mini
10

Histogram





Roc curve verificatgion



Connect VGG-19 network ;Max epochs 6;
Mini batch size 10; Validation accuracy:
89.16%; VGG-19 network achieved an
89.16% validation accuracy.

The Analyzing histograms: The cosine similarity between genuine and imposter scores for certain image pairs and showed more overlap compared to others.

The ROC curve observations: Among different verification scenarios, the ROC curve for the second verification exhibited the highest AUC score when compared to the ROC curve of the first verification