

# **Building Damage Detection based on Post-Hurricane Satellite Imagery using Transfer Learning and Convolutional Neural Networks**

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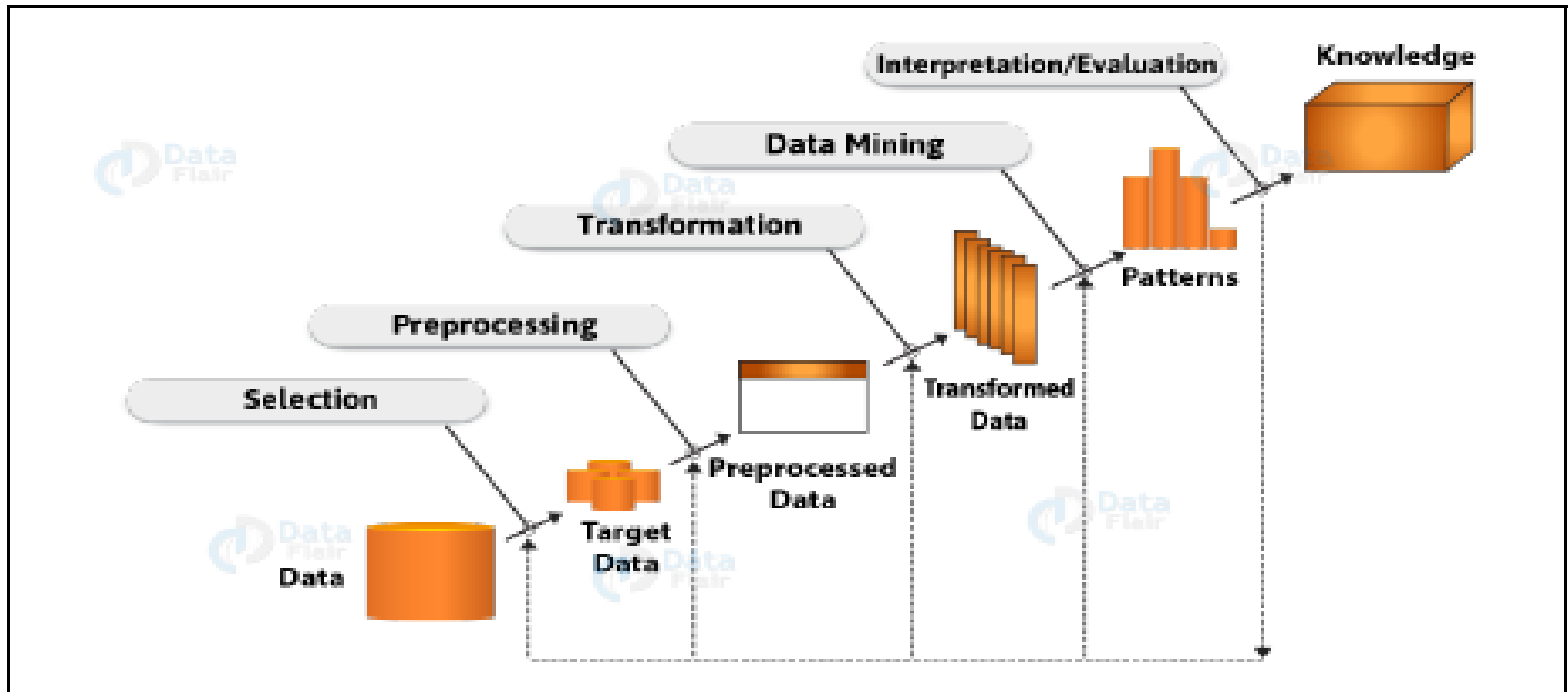
# Objective

The objective of this study is to identify buildings that have been impacted by a natural disaster using satellite images of the areas impacted by the disaster.

# Related Works

After critically reviewing the key literature in the domain, we can conclude that different pre-trained deep learning architectures like VGG, Inception V3, Resnet have performed exceptionally well in the satellite imagery classification tasks. They were even able to outperform the state-of-the-art models in some cases. Therefore, for the purpose of our study, we will also be using a pre-trained VGG16 architecture for the damage annotation in the optical satellite imagery. We will also design two different custom architectures using data augmentation and Leaky ReLU activation function and compare them based on different metrics like test and validation accuracy, ROC curves, precision, etc. The next section will discuss the methodology implemented for the study.

# METHODOLOGY



# Data Selection

Dataset Source

IEEE data portal



<https://ieee-dataport.org/open-access/detecting-damaged-buildings-post-hurricane-satellite-imagery-based-customized>)

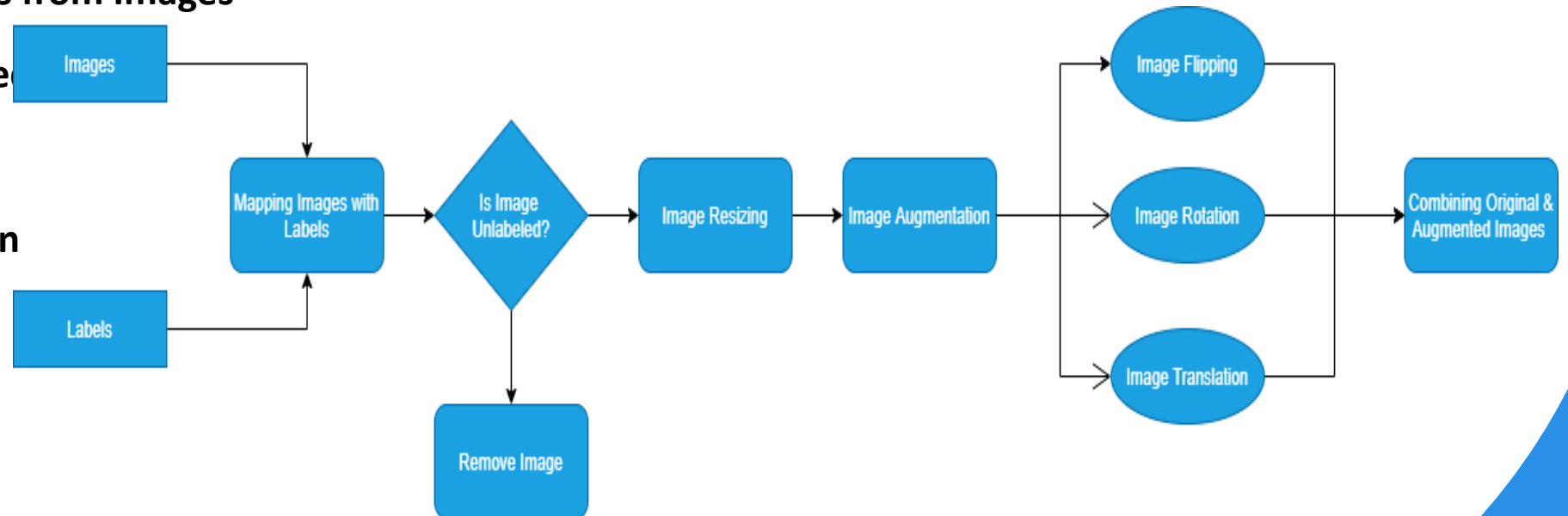
# Ethical Concerns

*NO MEANS NO!*

# Data Pre-Processing & Transformation

- Importing data into Google Colab
- Extracting Labels from Images
- Extracting Features from Images
- Handling Unlabelled
- Image Resizing
- Data Augmentation

- ☐ Image Flipping
- ☐ Image Rotation
- ☐ Data Translation



# Implementation and Evaluation

## VGG16 (Transfer Learning)

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
vgg16 (Functional)	(None, 4, 4, 512)	14714688
global_average_pooling2d (G1	(None, 512)	0
dense (Dense)	(None, 1)	513
=====		

Total params: 14,715,201

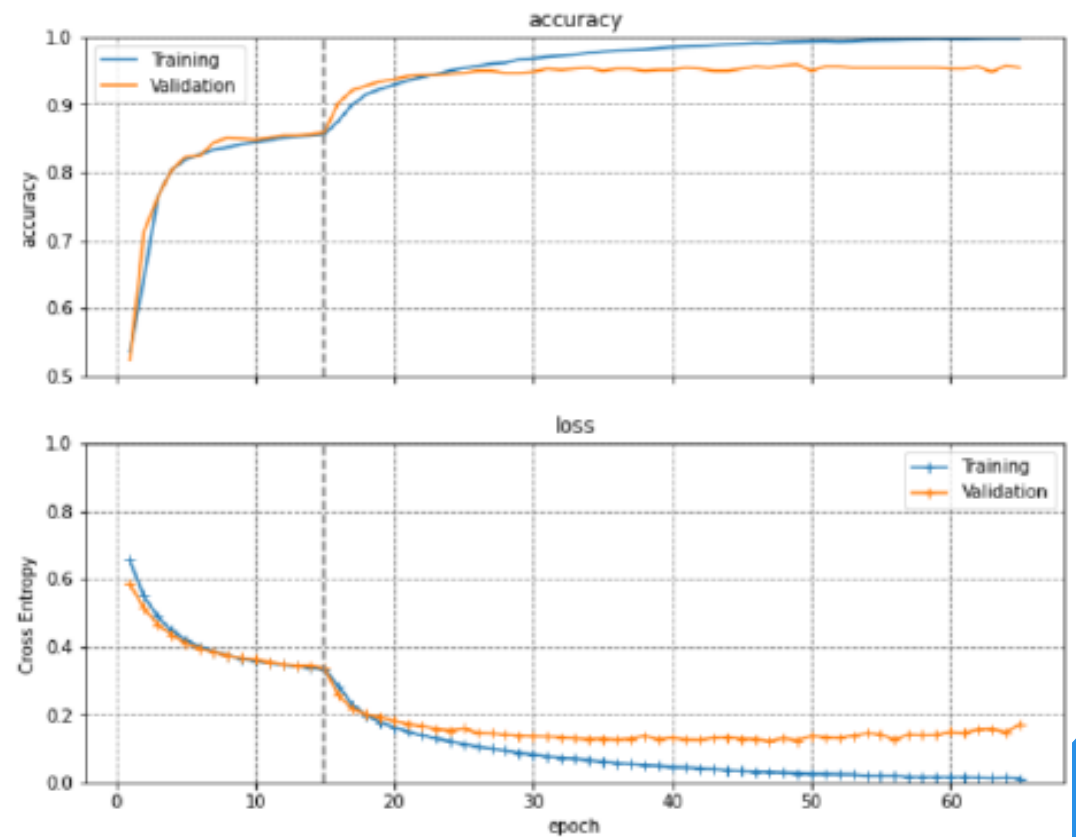
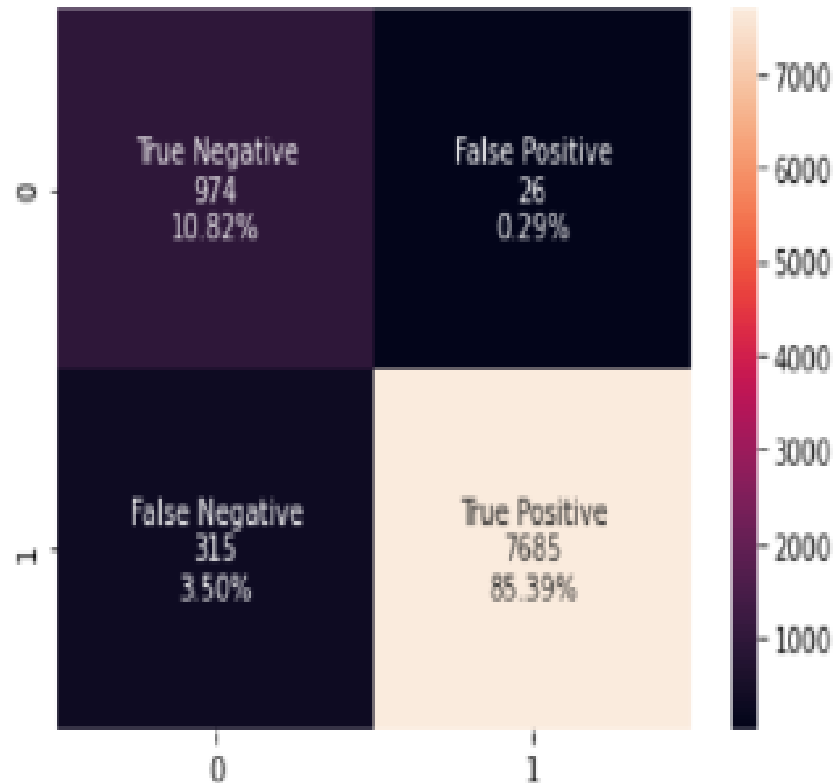
Trainable params: 7,079,937

Non-trainable params: 7,635,264



# VGG16 Model Evaluation

- Accuracy - 96.21 %



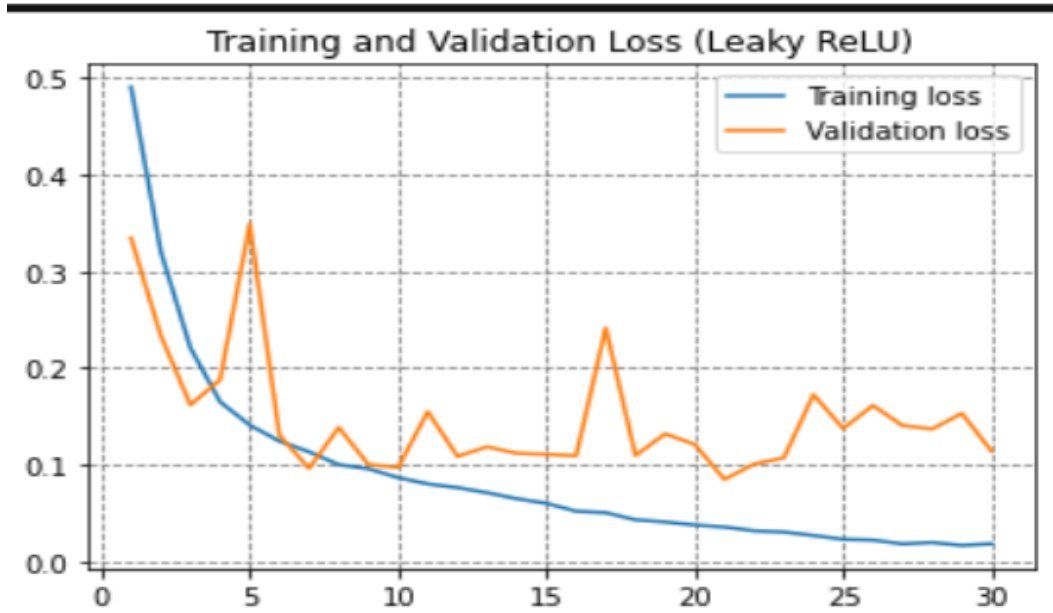
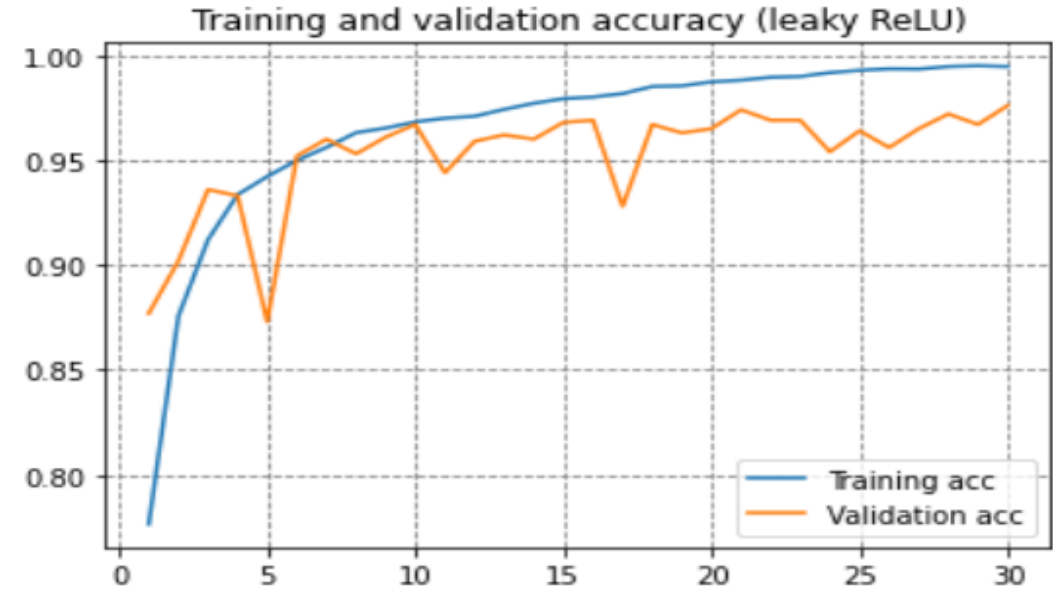
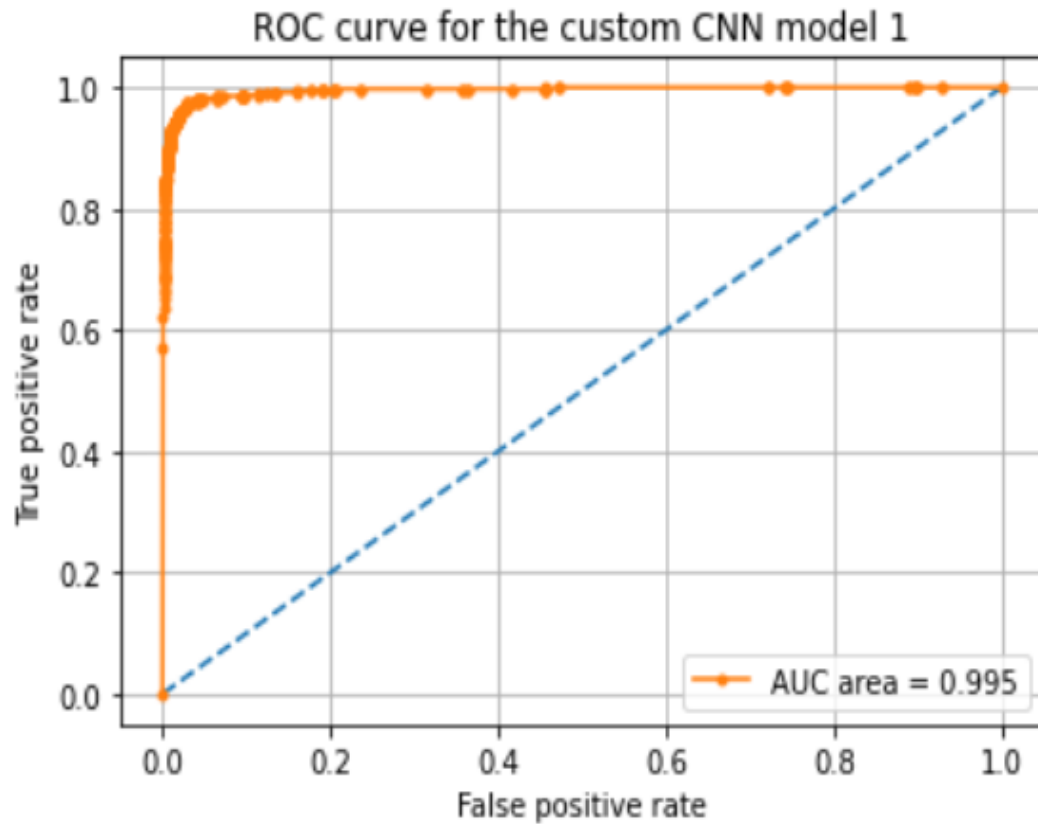
# Custom Convolutional Neural Network 1

Model: "sequential\_4"

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 148, 148, 32)	896
leaky_re_lu_1 (LeakyReLU)	(None, 148, 148, 32)	0
max_pooling2d (MaxPooling2D)	(None, 74, 74, 32)	0
conv2d_2 (Conv2D)	(None, 72, 72, 64)	18496
leaky_re_lu_2 (LeakyReLU)	(None, 72, 72, 64)	0
max_pooling2d_1 (MaxPooling2D)	(None, 36, 36, 64)	0
conv2d_3 (Conv2D)	(None, 34, 34, 128)	73856
leaky_re_lu_3 (LeakyReLU)	(None, 34, 34, 128)	0
max_pooling2d_2 (MaxPooling2D)	(None, 17, 17, 128)	0
conv2d_4 (Conv2D)	(None, 15, 15, 128)	147584
leaky_re_lu_4 (LeakyReLU)	(None, 15, 15, 128)	0
max_pooling2d_3 (MaxPooling2D)	(None, 7, 7, 128)	0
Flatten (Flatten)	(None, 6272)	0
dense_1 (Dense)	(None, 512)	3211776
leaky_re_lu_5 (LeakyReLU)	(None, 512)	0
dense_2 (Dense)	(None, 1)	513
Total params: 3,453,121		
Trainable params: 3,453,121		
Non-trainable params: 0		

# Custom CNN 1 : Evaluation

- Accuracy : 89%



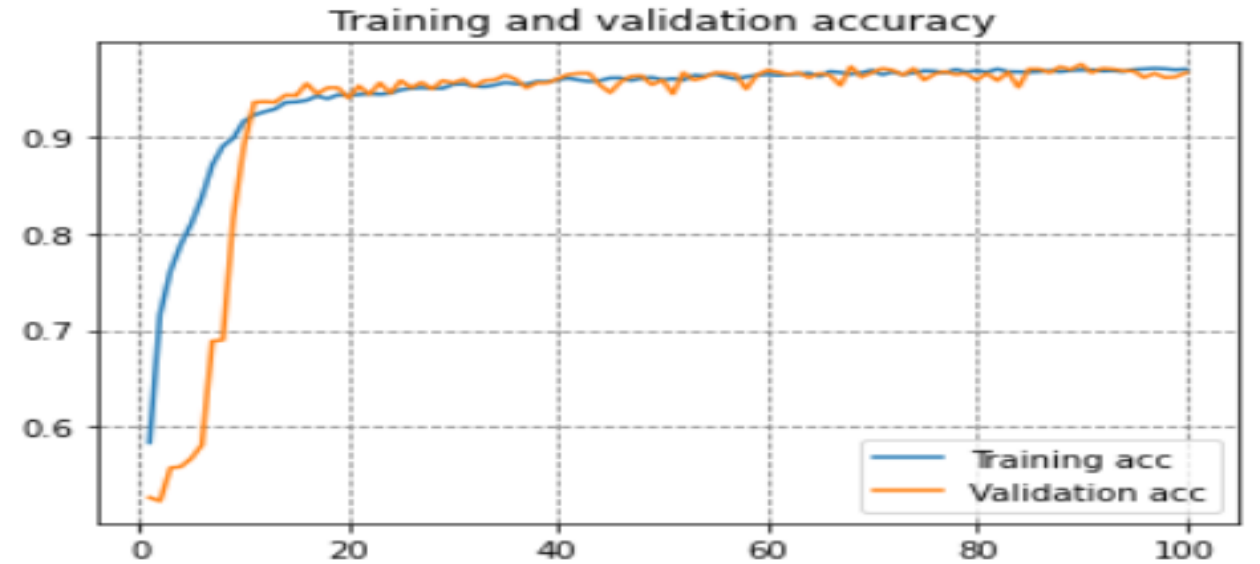
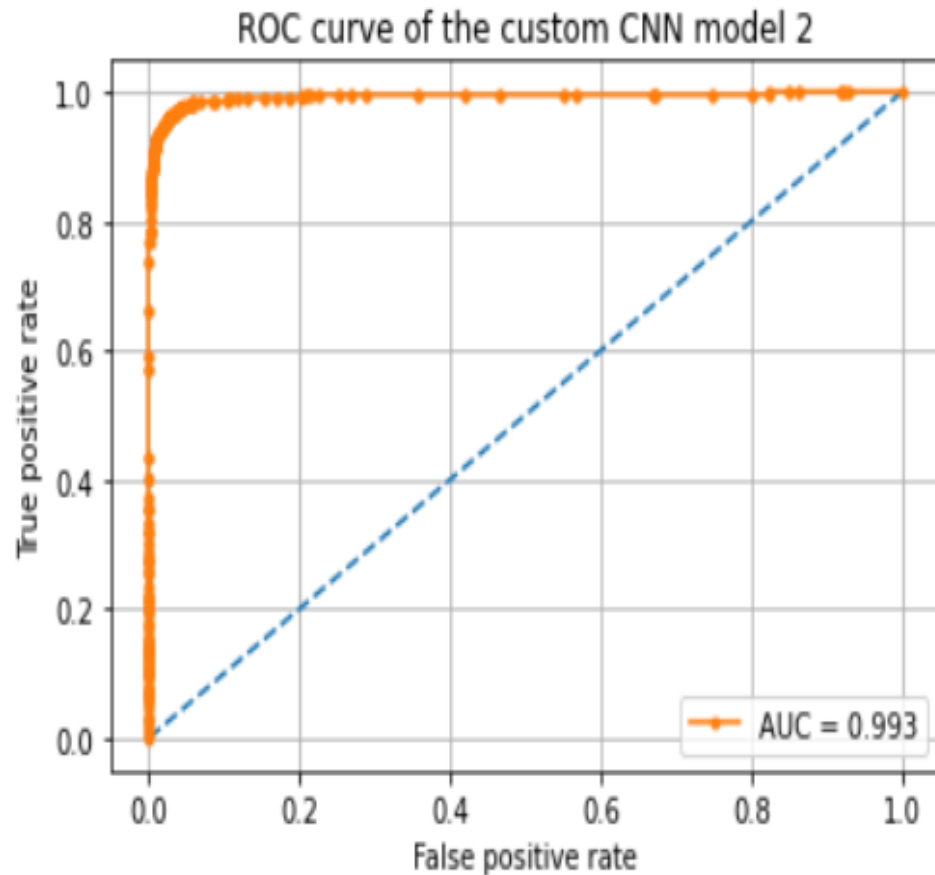
# Custom Convolutional Neural Network 2

Model: "sequential\_3"

Layer (type)	Output Shape	Param #
conv2d_8 (Conv2D)	(None, 148, 148, 32)	896
leaky_re_lu_10 (LeakyReLU)	(None, 148, 148, 32)	0
max_pooling2d_8 (MaxPooling2D)	(None, 74, 74, 32)	0
dropout_10 (Dropout)	(None, 74, 74, 32)	0
conv2d_9 (Conv2D)	(None, 72, 72, 64)	18496
leaky_re_lu_11 (LeakyReLU)	(None, 72, 72, 64)	0
max_pooling2d_9 (MaxPooling2D)	(None, 36, 36, 64)	0
dropout_11 (Dropout)	(None, 36, 36, 64)	0
conv2d_10 (Conv2D)	(None, 34, 34, 128)	73856
leaky_re_lu_12 (LeakyReLU)	(None, 34, 34, 128)	0
max_pooling2d_10 (MaxPooling2D)	(None, 17, 17, 128)	0
dropout_12 (Dropout)	(None, 17, 17, 128)	0
conv2d_11 (Conv2D)	(None, 15, 15, 128)	147584
leaky_re_lu_13 (LeakyReLU)	(None, 15, 15, 128)	0
max_pooling2d_11 (MaxPooling2D)	(None, 7, 7, 128)	0
dropout_13 (Dropout)	(None, 7, 7, 128)	0
flatten_2 (Flatten)	(None, 6272)	0
dropout_14 (Dropout)	(None, 6272)	0
dense_4 (Dense)	(None, 512)	3211776
leaky_re_lu_14 (LeakyReLU)	(None, 512)	0
dense_5 (Dense)	(None, 1)	513
Total params: 3,453,121		
Trainable params: 3,453,121		
Non-trainable params: 0		

# Custom CNN 2 : Evaluation

- Accuracy : 97.91



# Conclusion

The performance of Custom CNN 2 is better than VGG16 and Custom CNN 1 with an accuracy of 97.91.

