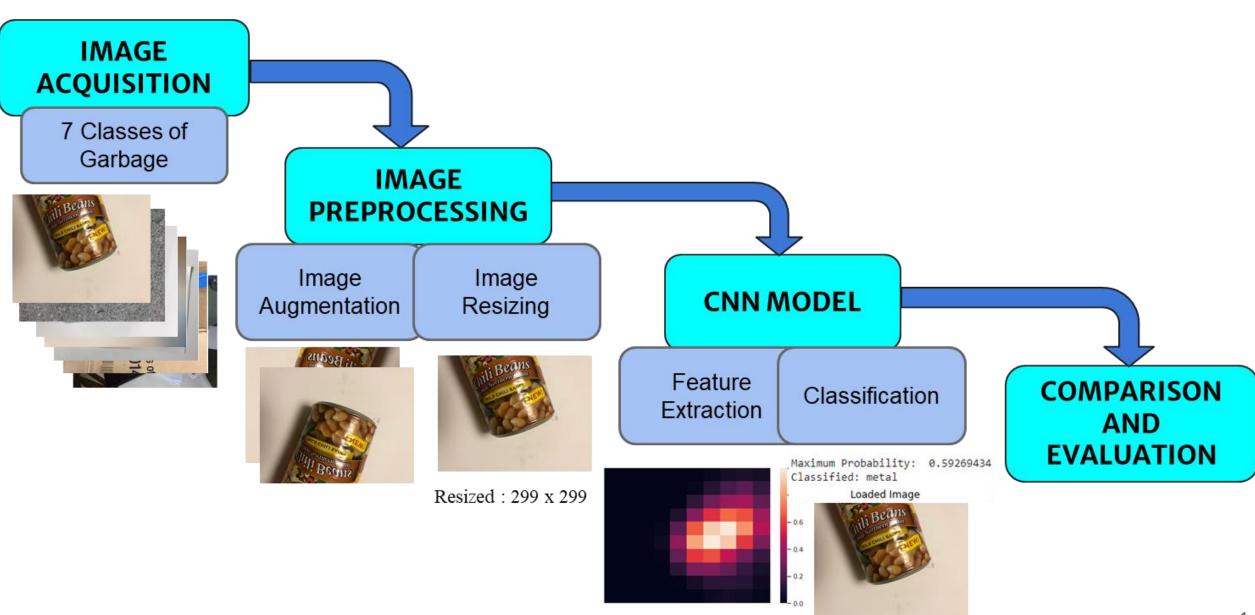
#### **BLOCK DIAGRAM**



# **IMAGE ACQUISITION**

- Images collected from online sources.
- Live Images have been obtained.
- Classification into 7 Different classes of Garbage Cardboard, Metal, Paper, Plastic, Glass, Trash, Ewaste.
- Image Format: jpg
- Image sizes: 4032x3024, 3024x4032, 3264x2448, 2448x3264

#### Sources:

- 1. <a href="https://github.com/garythung/trashnet">https://github.com/garythung/trashnet</a>
- 2. <a href="https://www.kaggle.com/kaustubh2402/ewaste-dataset">https://www.kaggle.com/kaustubh2402/ewaste-dataset</a>

#### IMAGE DATA AUGMENTATION

- Image Flip
- Image Shear
- Image Zoom
- Image Height Shift
- Image Width Shift
- Image Rotation

#### **IMAGE RESIZING**

- Resizing an image means changing the dimensions of it, be it width alone, height alone or changing both of them.
- The aspect ratio of the original image is also preserved in the resized image, so that the image does not appear distorted to the naked eye.
- To resize an image, OpenCV in python provides cv2.resize() function.
- Python, by default, uses a method called Bilinear Interpolation to resize the image.

#### FEATURE EXTRACTION

- Feature extraction involves reducing the number of resources required to describe a large set of data.
- Feature extraction refers to the process of transforming raw data into numerical features that can be processed while preserving information in the original dataset.

# Feature Extraction

**Features** 

#### FULLY CONNECTED LAYER

#### **Convolution Neural Network (CNN)** Input Output Pooling Pooling Pooling \_Horse -Zebra -Dog SoftMax Activation Convolution Convolution Convolution Function ReLU ReLU Kernel Flatten ReLU Layer Fully Connected -Feature Maps Layer Probabilistic Classification Feature Extraction Distribution

### MODEL TRAINING REQUIREMENTS

- Platform used: Google Colab
- Language: Python
- RAM Used: 2.38 GB
- GPU Used: 8.81 GB
- Total No. of Images in Dataset: 3344
- Images Split-up: Cardboard 403

E-waste - 406

Plastic - 482

Paper - 594

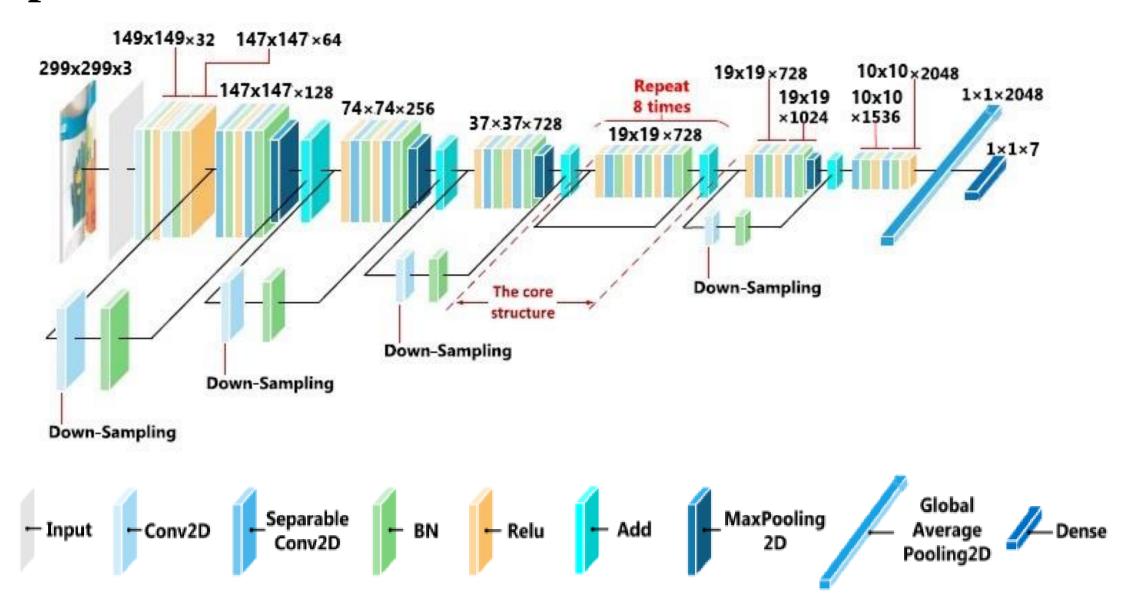
Metal - 410

Glass - 501

Trash - 548

- Training Split − 80% 2675
- Testing and Validation Split 20% 669

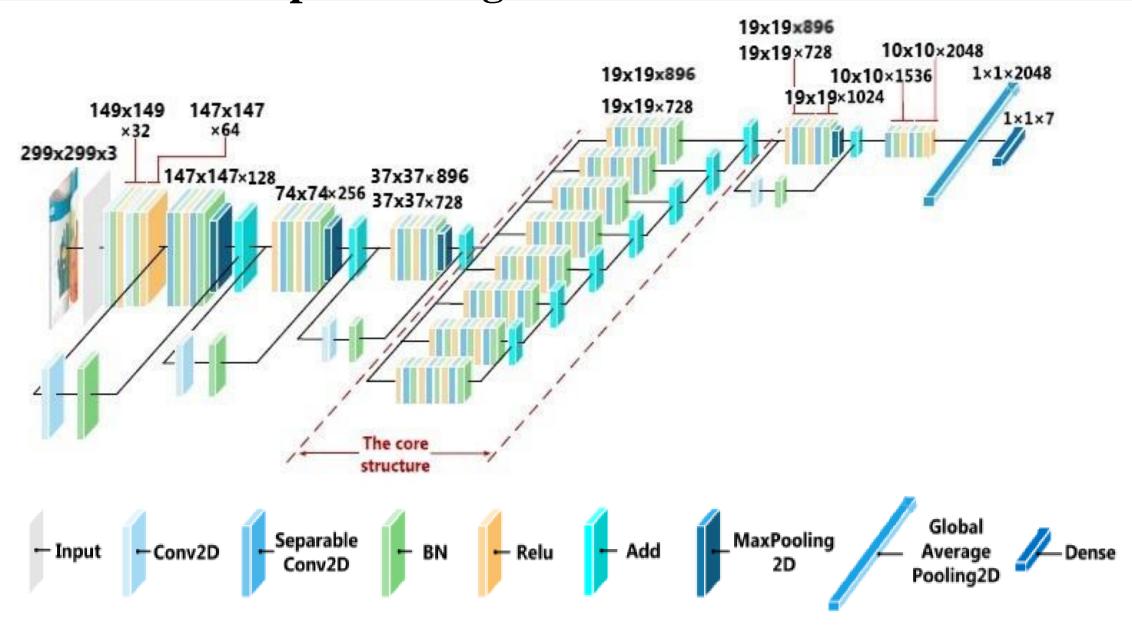
### **Xception Model**



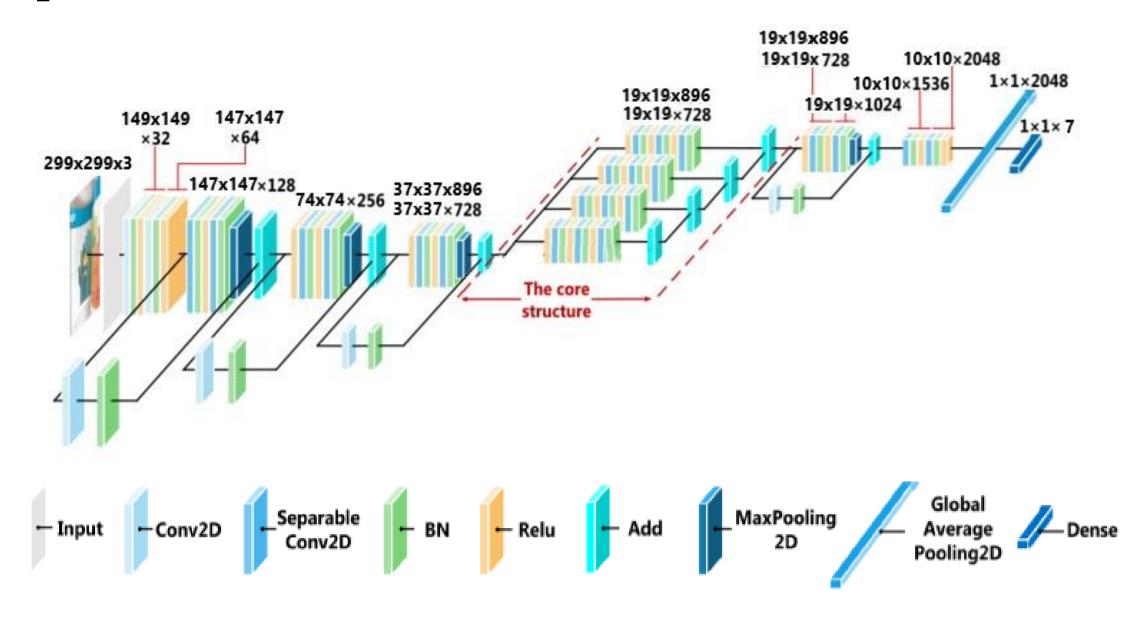
#### **XCEPTION FEATURES**

- Total number of Convolutional layers : 36
- Total number of Features extracted : 2048
- Xception uses Modified Depthwise Separable Convolution : Pointwise followed by Depthwise Convolution
- Optimizer Used: SGD (Stochastic Gradient Descent) Optimizer
- Loss function used: Categorical Cross Entropy

### Multi Branch Deep Learning Model



### **Proposed Model**



#### **CNN Model Parameters**

Models	Features Extracted	Input Size	Model Storage Size	No. of Epochs	Training Time per Epoch	Total Training Time
Xception	2048	299x299	80 MB	250	105 seconds	7.3 hours
Inception V3	2048	299x299	85 MB	500	67 seconds	9.3 hours
Resnet50	2048	229x229	91 MB	500	59 seconds	8.2 hours
VGG16	512	224x224	57 MB	500	50 seconds	6.9 hours
MobileNet	1024	224x224	13 MB	500	45 seconds	6.25 hours

THIRD REVIEW 12

#### **Confusion Matrix**

• Confusion Matrix gives us a matrix as output and describes the complete performance of a CNN model.

- There are 4 important terms :
  - True Positives
  - True Negatives
  - False Positives
  - False Negatives

	Predicted:	Predicted:
n=165	NO	YES
Actual:		
NO	50	10
Actual:		
YES	5	100

#### **Metrics for Model Evaluation**

1. 
$$Accuracy = \frac{True\ Positive + True\ Negative}{Total\ Samples}$$
 Range = 0 to 1

2.  $Precision = \frac{True\ Positive}{True\ Positive + False\ Positive}$  Range = 0 to 1

3.  $Recall = \frac{True\ Positive}{True\ Positive + False\ Negative}$  Range = 0 to 1

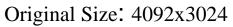
4.  $F1\ Score = \frac{True\ Positive}{True\ Positive + \frac{1}{2}(False\ Positive + False\ Negative)} = \frac{2}{\frac{1}{Precision} + \frac{1}{Recall}}$  Range = 0 to 1

5.  $MCC = \frac{TP * TN - FP * FN}{\sqrt{(TP + FP)(TP + FN)(TN + FP)(TN + FN)}}$  Range = -1 to 1

6.  $\kappa = \frac{po - pe}{1 - ne} = 1 - \frac{1 - po}{1 - ne}$  Range = -1 to 1

# Results







Original Size: 3264x2448



Original Size: 3264x2448



Resized: 299x299



Resized: 299x299



Resized: 299x299

# **IMAGE AUGMENTATION**



Original Image



Rotated Image



Sheared Image



Horizontal Flip

# **IMAGE AUGMENTATION**



Original Image



Height Shifted Image

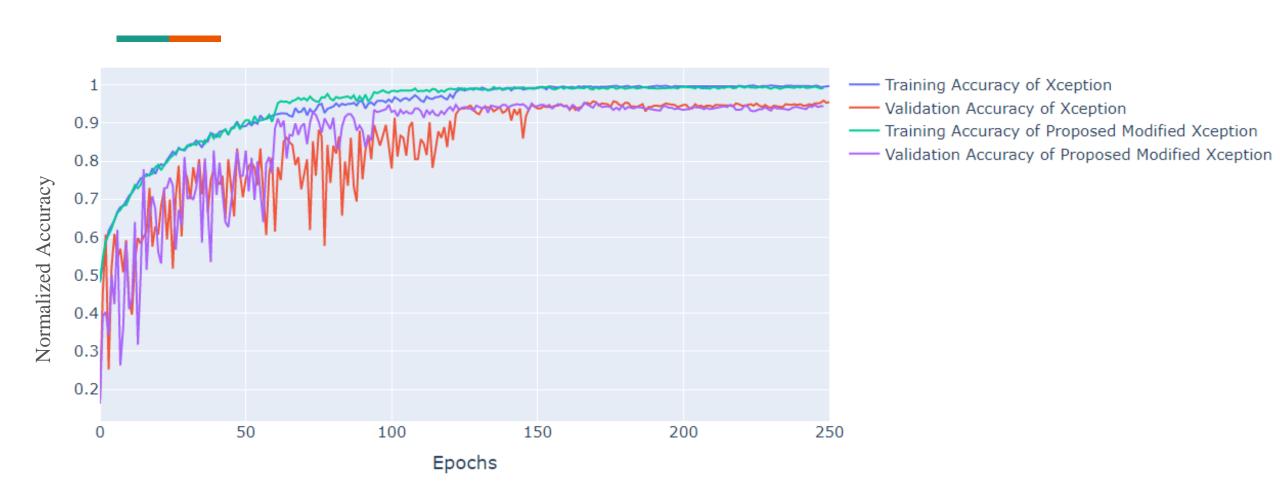


Zoomed Image

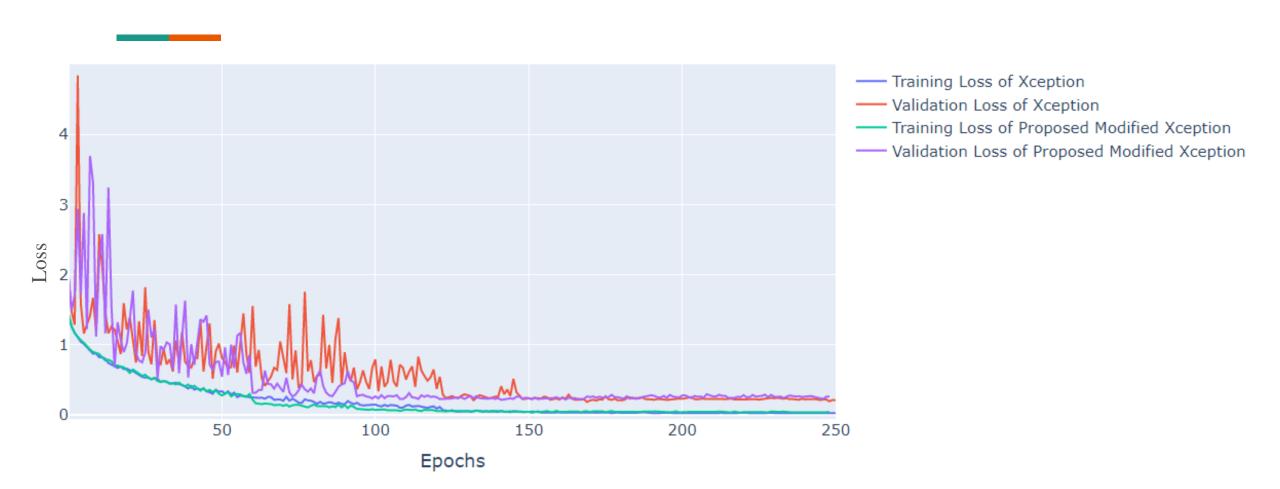


Width Shifted Image

# **Performance of Xception - Accuracy**



# **Performance of Xception - Loss**



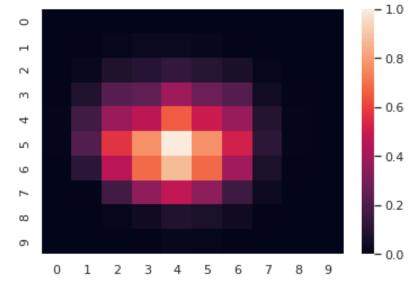
#### VISUALIZATION OF EXTRACTED FEATURES

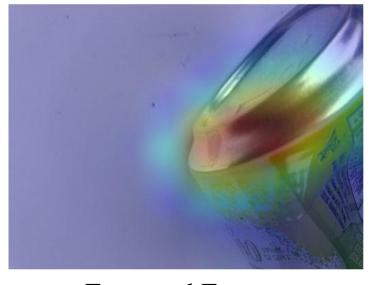


Original Image

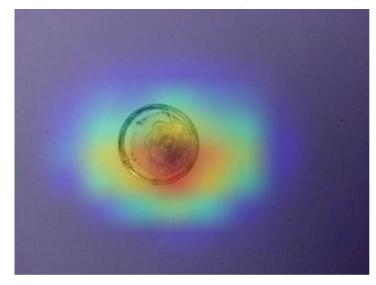


-1.0
-0.8
-0.6
-0.6
-0.4
-0.2
Extracted Features





Extracted Features
Visualized over Original Image



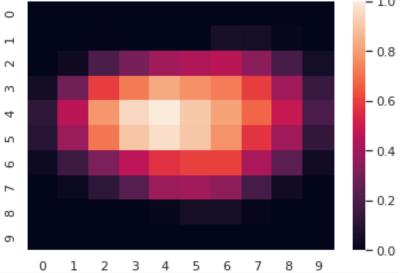


Original Image



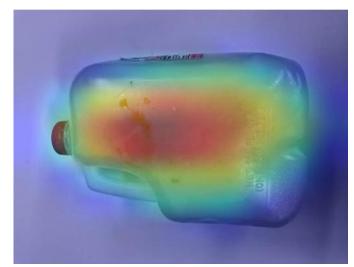
-1.0 -0.8 -0.6 -0.4 -0.2 0 1 2 3 4 5 6 7 8 9

**Extracted Features** 





Extracted Features Visualized over Original Image



# **CLASSIFICATION OUTPUT EXAMPLES**



Maximum Probability: 0.9979786

Predicted: Ewaste Truth: Ewaste



Maximum Probability: 0.9616854

Predicted: Glass Truth: Glass



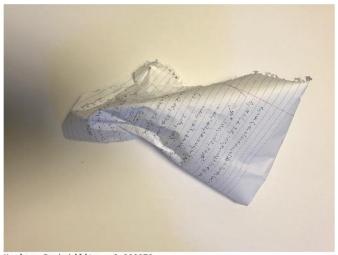
Maximum Probability: 0.9999882

Predicted: Cardboard Truth: Cardboard



Maximum Probability: 0.99974936

Predicted: Metal Truth: Metal



Maximum Probability: 0.999979

Predicted: Paper Truth: Paper



Maximum Probability: 0.999691

Predicted: Plastic Truth: Plastic

#### **CLASSIFICATION OUTPUT EXAMPLES**

Loaded Image	AND		
Predicted Class	Ewaste	Cardboard	Paper
Truth	Ewaste	Cardboard	Paper
Maximum Probability	0.9979786	0.9999882	0.999979

#### **CLASSIFICATION OUTPUT EXAMPLES**

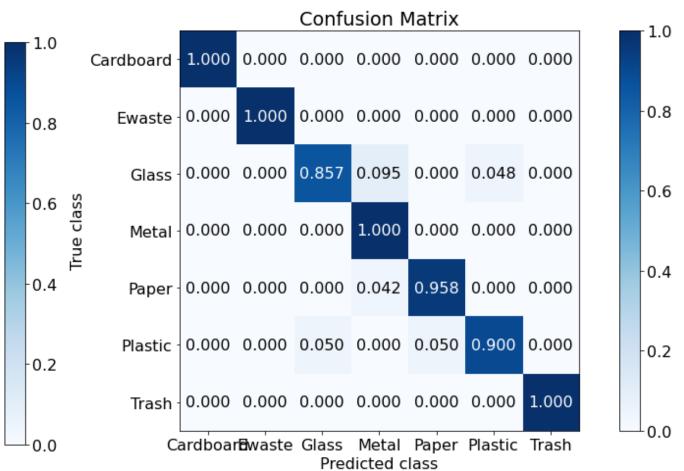
Loaded Image			
Predicted Class	Glass	Metal	Plastic
Truth	Glass	Metal	Plastic
Maximum Probability	0.9616854	0.99974936	0.999691

#### **Confusion Matrix**



# Confusion Matrix Cardboard 0.941 0.000 0.000 0.000 0.000 0.000 0.059 Ewaste 0.000 0.944 0.000 0.000 0.056 0.000 0.000 Glass 0.000 0.000 0.952 0.048 0.000 0.000 0.000 **True class** Metal 0.059 0.000 0.059 0.882 0.000 0.000 0.000 Paper 0.042 0.000 0.000 0.000 0.875 0.042 0.042 Plastic 0.000 0.000 0.050 0.000 0.000 0.950 0.000 Trash 0.000 0.000 0.000 0.000 0.000 1.000 Cardboar Bwaste Glass Metal Paper Plastic Trash Predicted class

#### **Proposed Modified Xception**

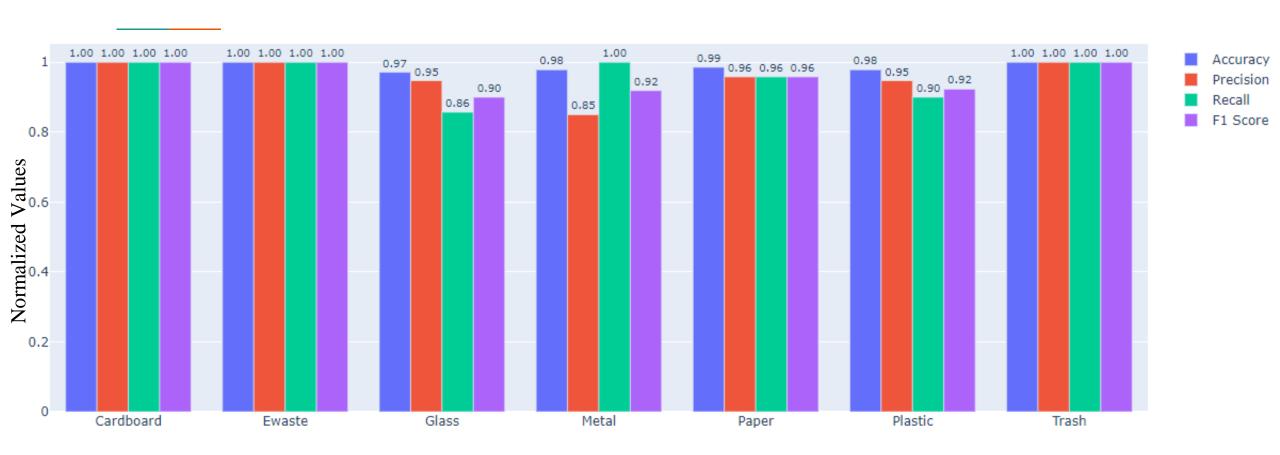


# Comparison

Model Metrics	Normalized Values		
Accuracy	0.9571		
MCC	0.9503		
Cohen's Kappa Coefficient	0.9499		

Class	Accuracy	Precision	Recall	F1 Score
Cardboard	1.0000	1.0000	1.0000	1.0000
Ewaste	1.0000	1.0000	1.0000	1.0000
Glass	0.9714	0.9474	0.8571	0.9714
Metal	0.9786	0.8500	1.0000	0.9786
Paper	0.9857	0.9583	0.9583	0.9857
Plastic	0.9786	0.9474	0.9000	0.9786
Trash	1.0000	1.0000	1.0000	1.0000

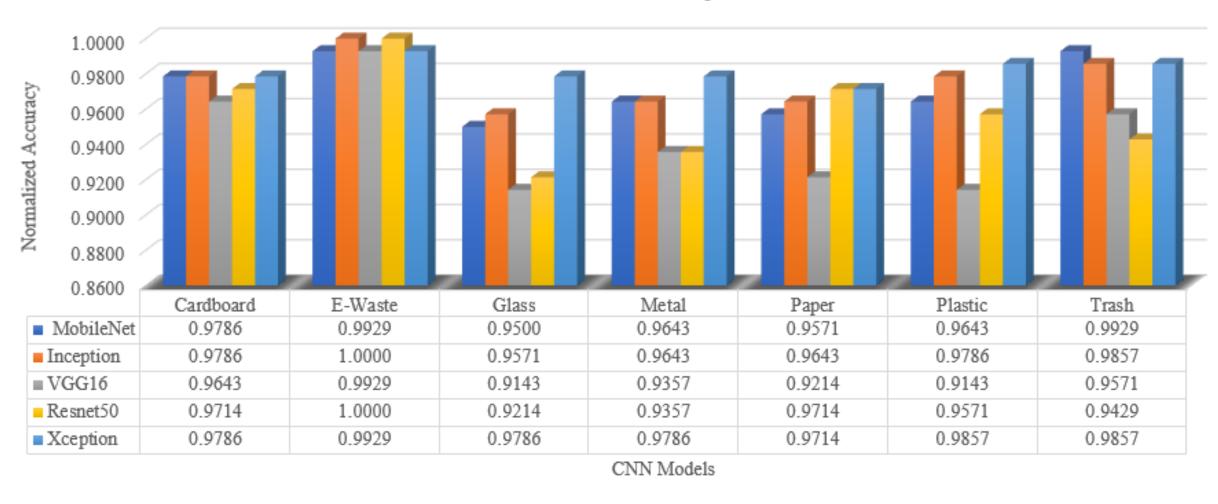
#### Accuracy, Recall, Precision and F1 Score of Proposed Model



Classes

#### **Class Accuracy**

#### Class Accuracy



■ VGG16 Resnet50

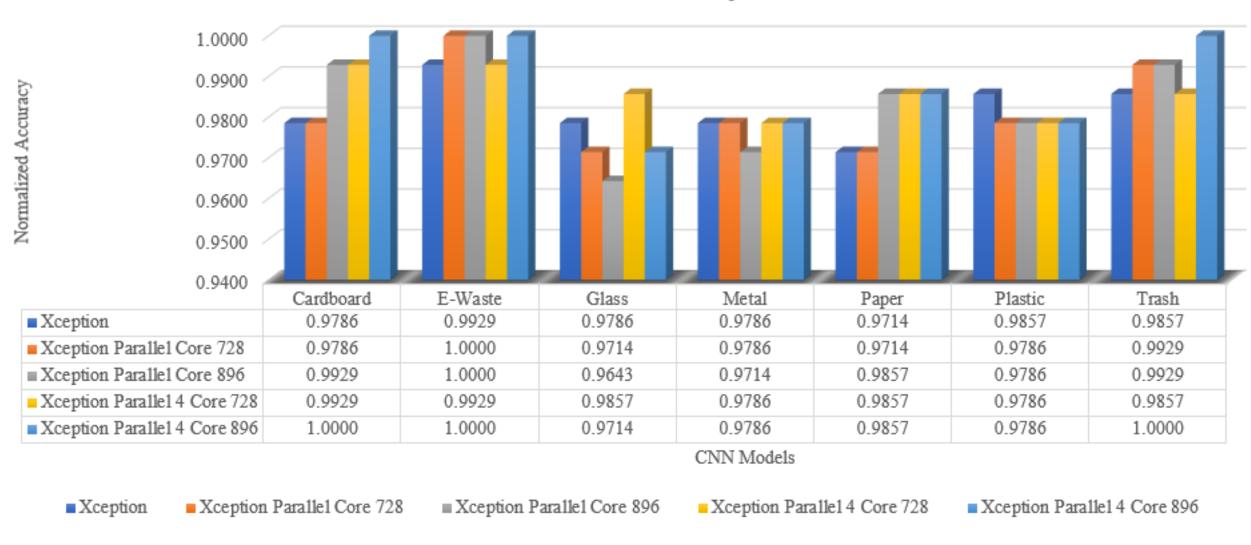
Xception

MobileNet

Inception

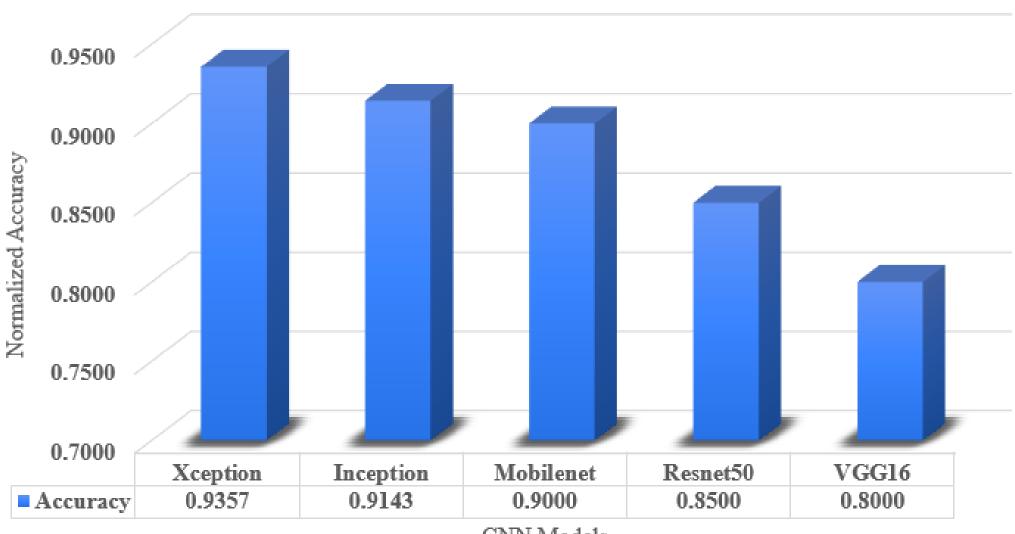
#### **Class Accuracy**

#### Class Accuracy



#### **Model Accuracy Comparison**

#### Accuracy



#### **Proposed Model Accuracy Comparison**

#### Accuracy

