

# **ELEC5690 Assignment 1**

## **Problem 1: Skin Lesion Image Classification**

Model used: ResNet50

Loss functions used:

- Cross-entropy loss
- Cross-entropy + focal loss

### **Model Parameters**

| Model                    | Cross-entropy loss only | Cross-entropy + focal loss |
|--------------------------|-------------------------|----------------------------|
| Learning Rate            | 0.001                   | 0.001                      |
| Number of epochs trained | 15                      | 10                         |
| Optimizer function       | Adam                    | Adam                       |
| Pre-trained              | Yes                     | Yes                        |

### **Overall Metrics**

| Loss Function              | Training Loss | Validation Loss | Training Accuracy | Validation Accuracy | Testing Accuracy |
|----------------------------|---------------|-----------------|-------------------|---------------------|------------------|
| Cross-entropy loss only    | 0.3188        | 0.7262          | 88.33%            | 78.76%              | 71.16%           |
| Cross-entropy + focal loss | 0.0681        | 0.3373          | 92.97%            | 79.27%              | 73.94%           |

### **Metrics on Test Dataset**

#### **1. Cross-entropy loss only**

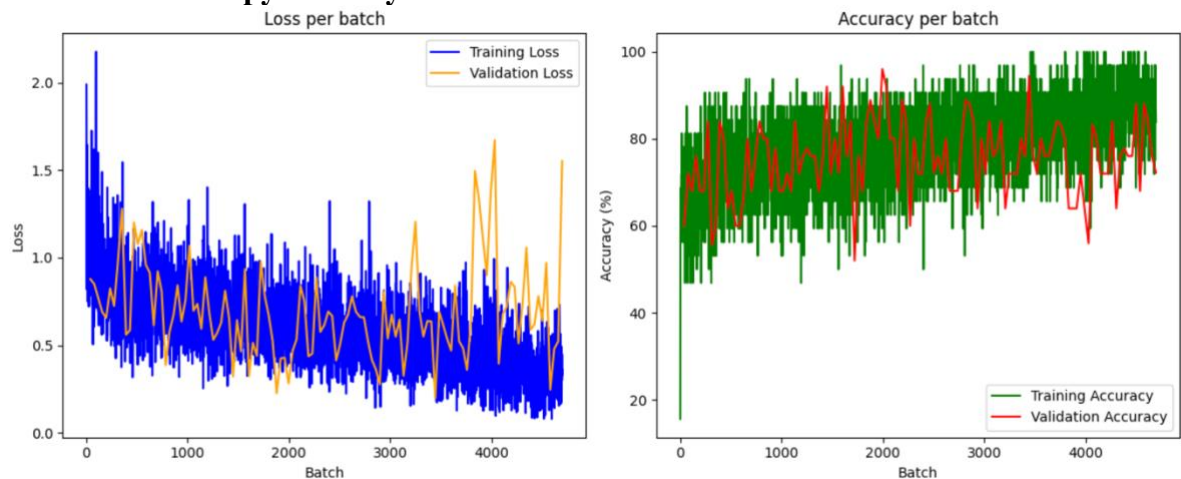
| Class            | Precision | Recall | F1-score | Support |
|------------------|-----------|--------|----------|---------|
| MEL              | 0.55      | 0.44   | 0.49     | 171     |
| NV               | 0.78      | 0.92   | 0.85     | 909     |
| BCC              | 0.44      | 0.48   | 0.46     | 93      |
| AKIEC            | 0.32      | 0.21   | 0.25     | 43      |
| BKL              | 0.64      | 0.35   | 0.46     | 217     |
| DF               | 0.70      | 0.16   | 0.26     | 44      |
| VASC             | 0.57      | 0.74   | 0.64     | 35      |
| Accuracy         |           |        | 0.71     | 1512    |
| Macro Average    | 0.57      | 0.47   | 0.49     | 1512    |
| Weighted Average | 0.69      | 0.71   | 0.69     | 1512    |

## 2. Cross-entropy + Focal loss

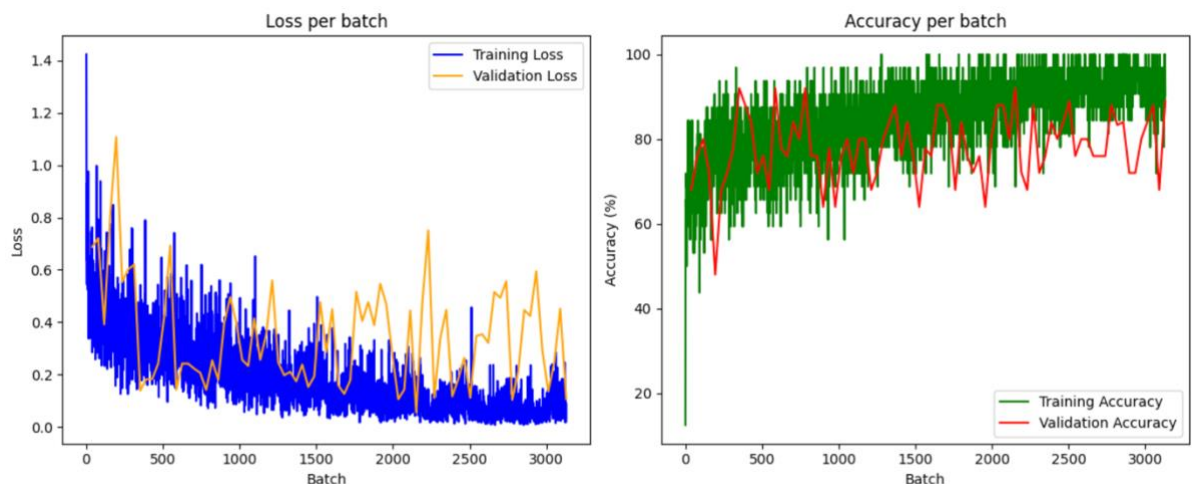
| Class                   | Precision | Recall | F1-score | Support |
|-------------------------|-----------|--------|----------|---------|
| MEL                     | 0.66      | 0.23   | 0.34     | 171     |
| NV                      | 0.79      | 0.95   | 0.86     | 909     |
| BCC                     | 0.57      | 0.63   | 0.60     | 93      |
| AKIEC                   | 0.35      | 0.21   | 0.26     | 43      |
| BKL                     | 0.62      | 0.58   | 0.60     | 217     |
| DF                      | 1.00      | 0.09   | 0.17     | 44      |
| VASC                    | 0.68      | 0.60   | 0.64     | 35      |
| <b>Accuracy</b>         |           |        | 0.74     | 1512    |
| <b>Macro Average</b>    | 0.67      | 0.47   | 0.50     | 1512    |
| <b>Weighted Average</b> | 0.73      | 0.74   | 0.71     | 1512    |

## Training and Validation curves

### 1. Cross-entropy loss only

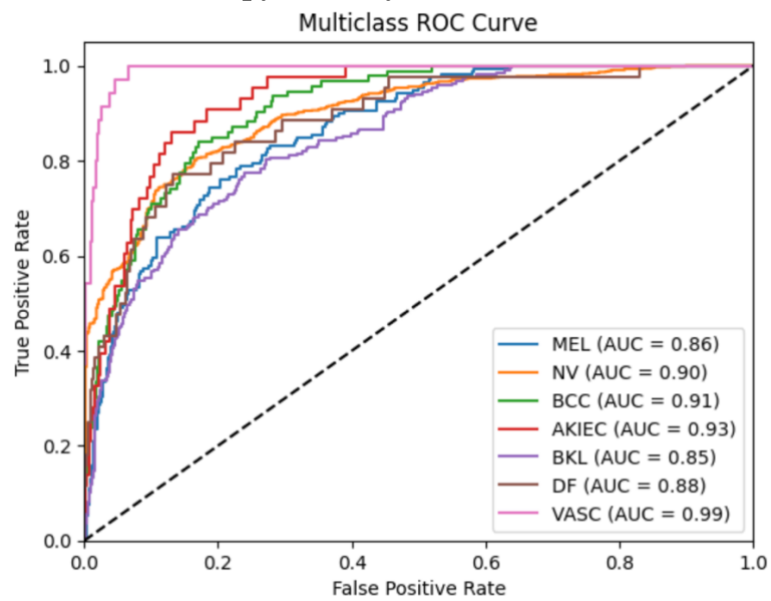


### 2. Cross-entropy + Focal loss

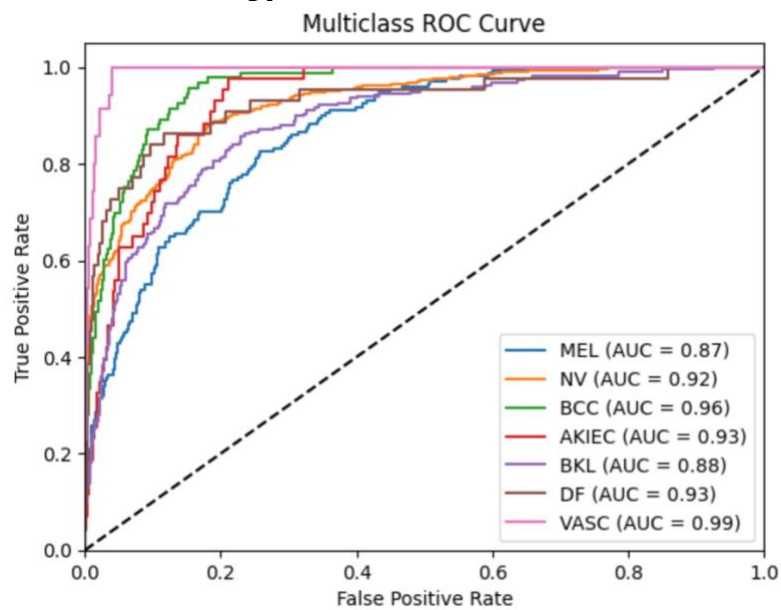


## ROC curves

### 1. Cross-entropy loss only



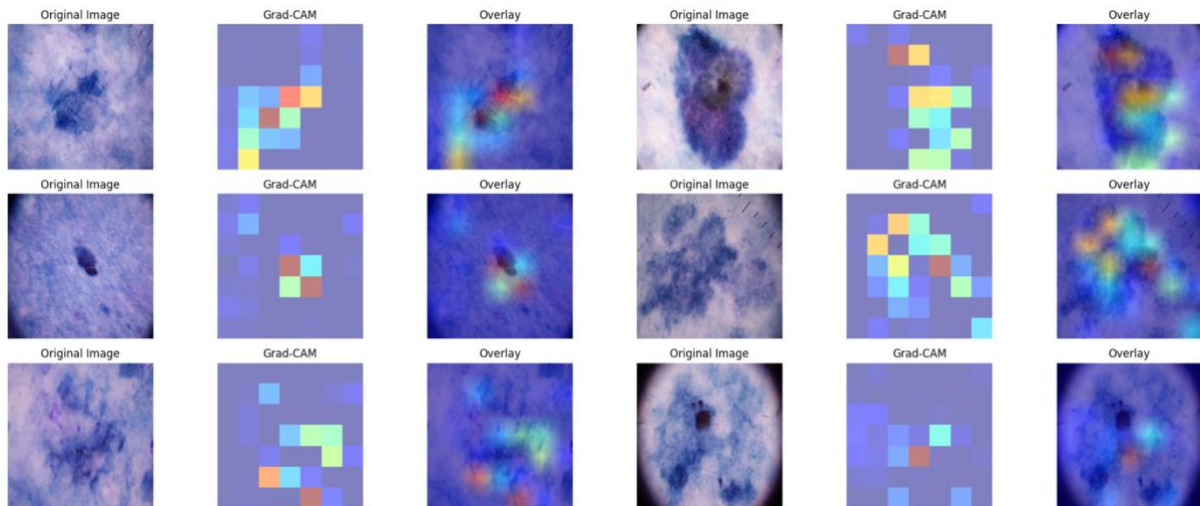
### 2. Cross-entropy + Focal loss



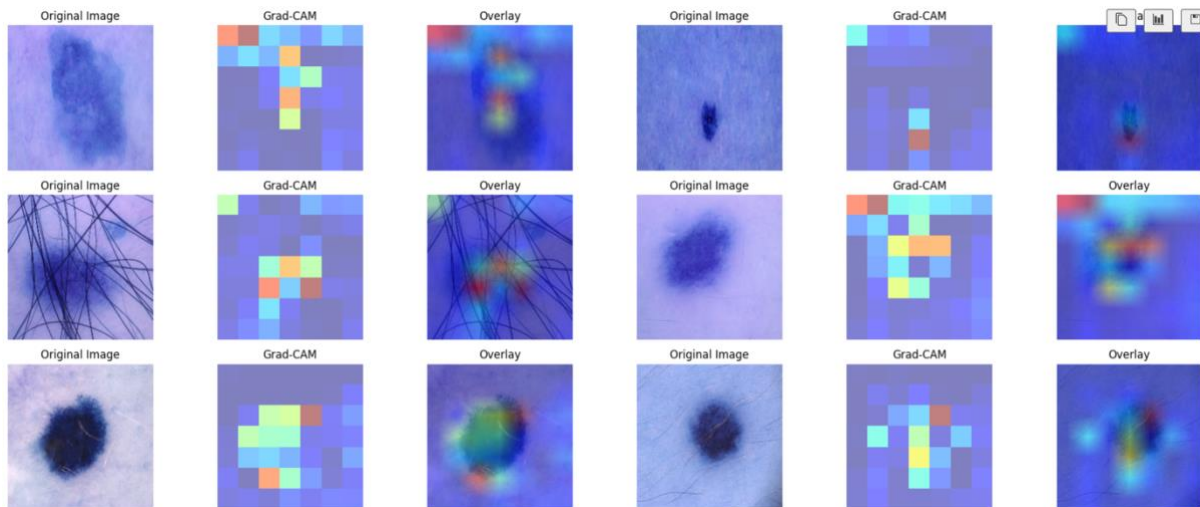
*According to the above metrics on test dataset, it is evident that the second model is better, hence we use that model for the below examples.*

### **Gradient class-activation map ( Using Cross-entropy + focal loss model )**

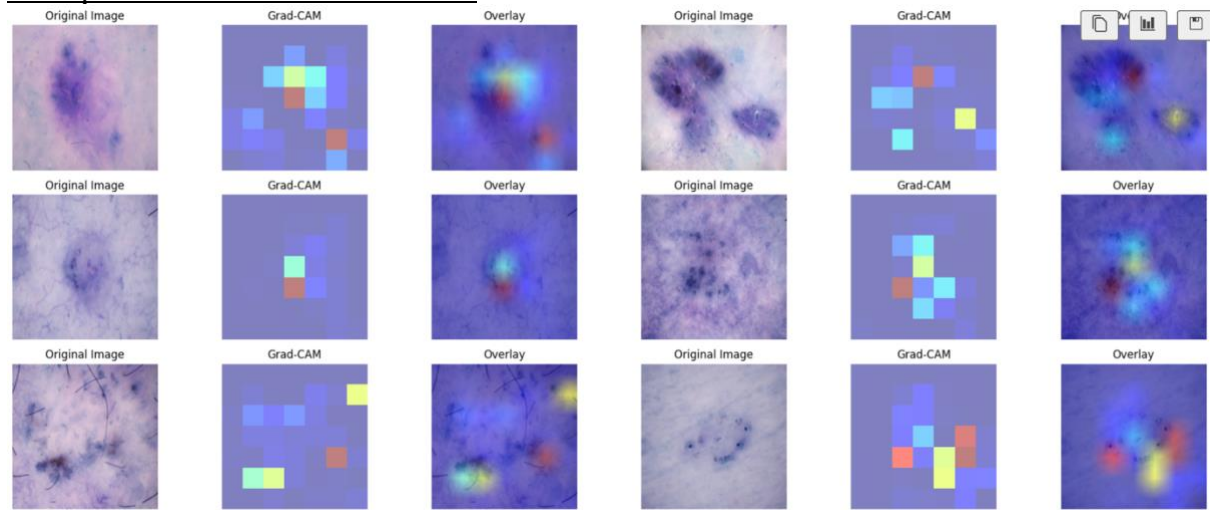
Examples for the first Class – MEL:



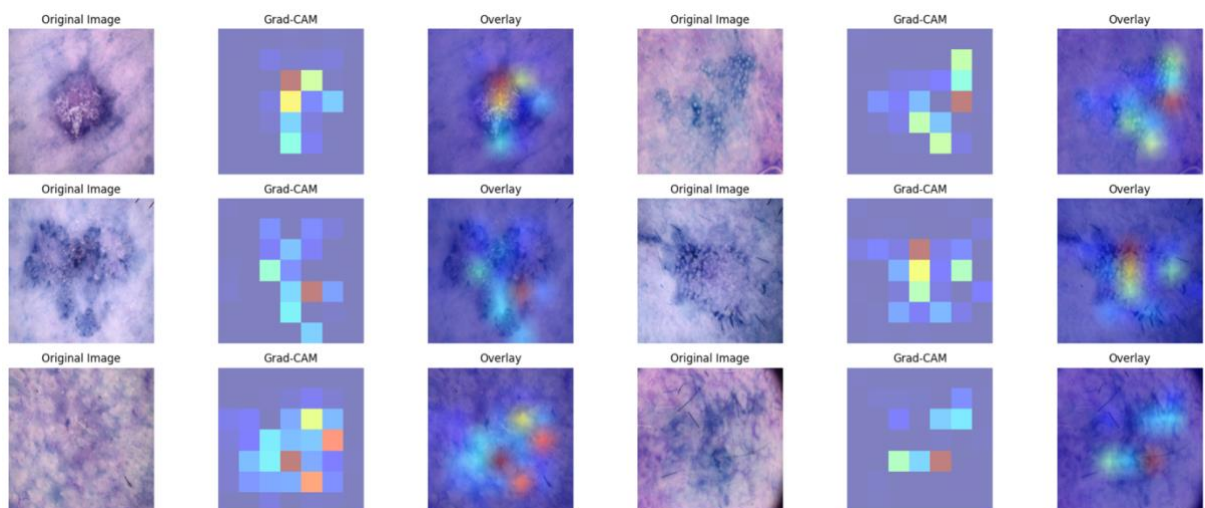
Examples for the second Class – NV:



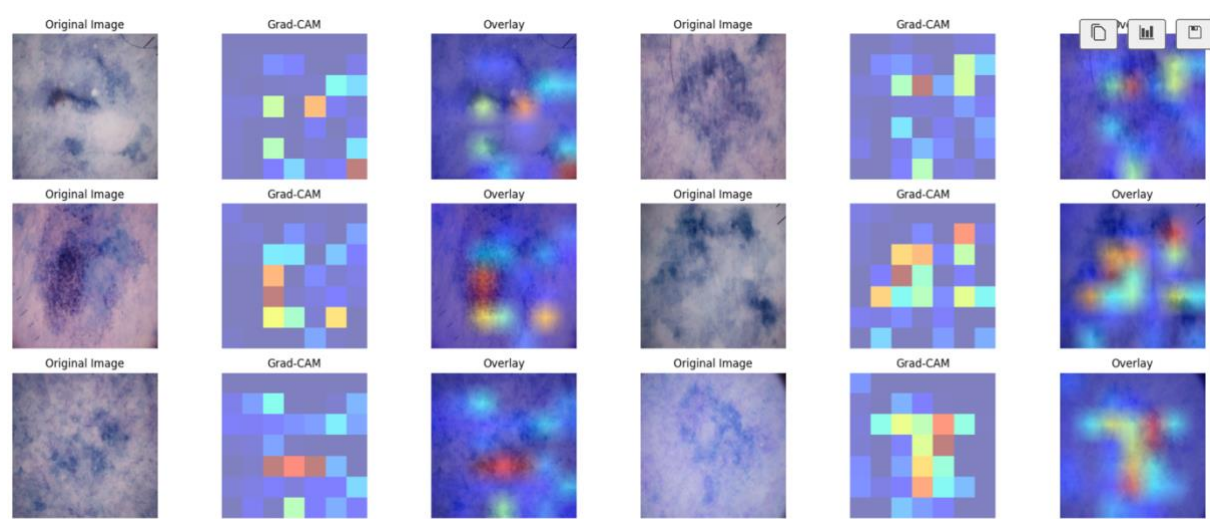
### Examples for the third Class – BCC:



### Examples for the fourth Class – AKIEC:

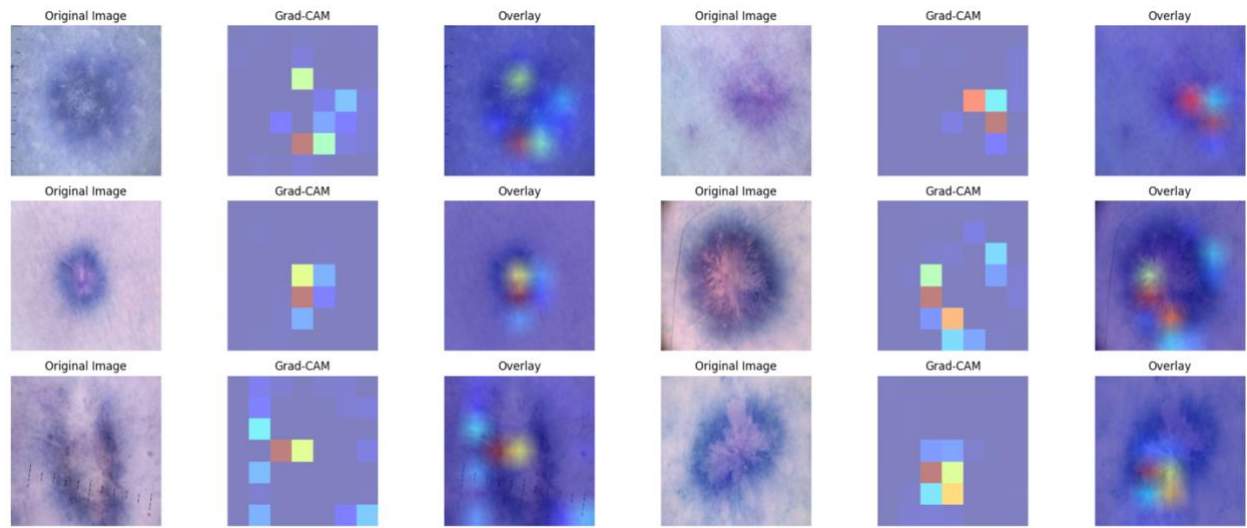


### Examples for the fifth Class – BKL:

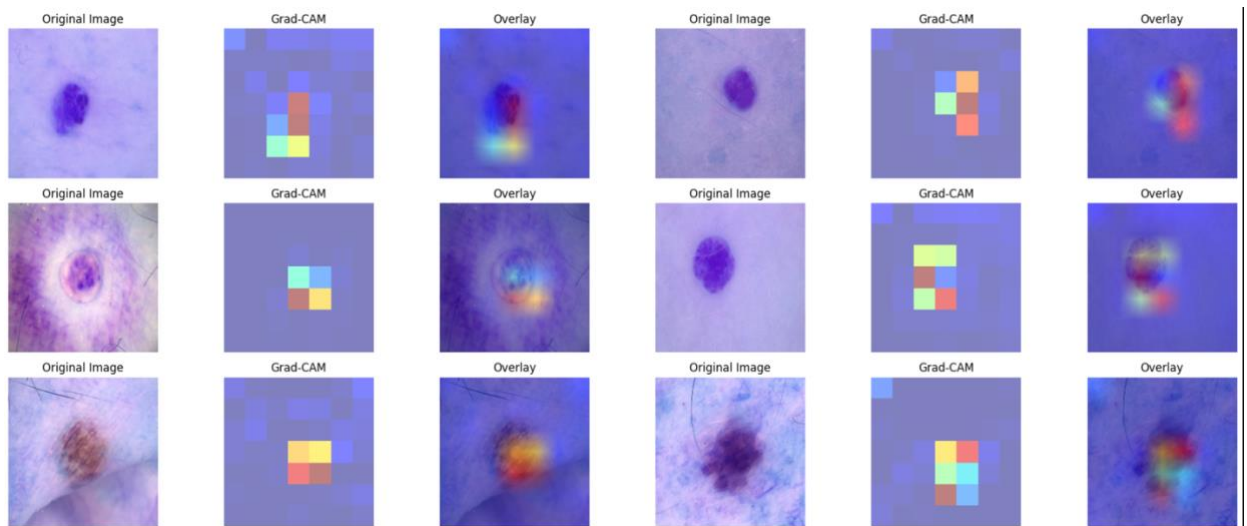




Examples for the sixth Class – DF:



Examples for the seventh Class – VASC:



## **Problem 2: 3D medical image segmentation**

**Model used: 3D Unet**

*Note: Metrics are rounded to 3 decimal places*

### **First Training Configuration:**

Learning rate = 0.001

Optimizer = Adam

Loss function = Binary Cross entropy

Data Augmentations = None

Batch Size = 4

Final Training loss = 0.311

Epoch trained = 50

#### **Metrics on test dataset:**

Dice Coefficient: 0.758

Jaccard Index: 0.639

ASD: 1.945

95HD: 8.606

### **Second Training Configuration:**

Learning rate = 0.00001

Optimizer = Adam

Loss function = Dice Loss

Data Augmentations = None

Batch Size = 2

Final Training loss = 0.359

Epochs trained = 75

#### **Metrics on test dataset:**

Dice Coefficient: 0.810

Jaccard Index: 0.686

ASD: 0.930

95HD: 6.087

### **Third Training Configuration:**

Learning rate = 0.00001

Optimizer = Adam

Loss function = Dice Loss

Data Augmentations = Random rotate, Random flip

Batch Size = 2

Final Training loss = 0.387

Epochs trained = 75

#### **Metrics on test dataset:**

Dice Coefficient: 0.737

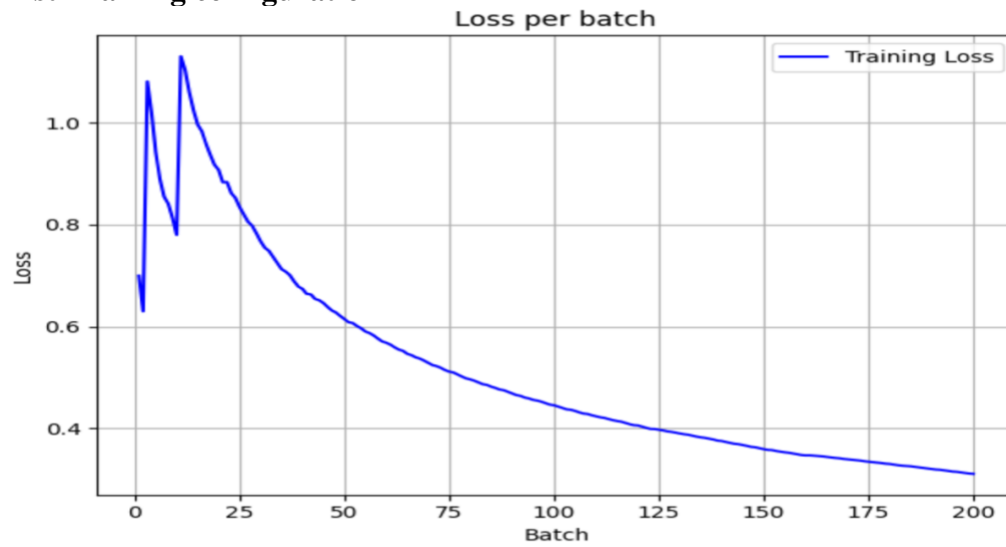
Jaccard Index: 0.593

ASD: 1.612

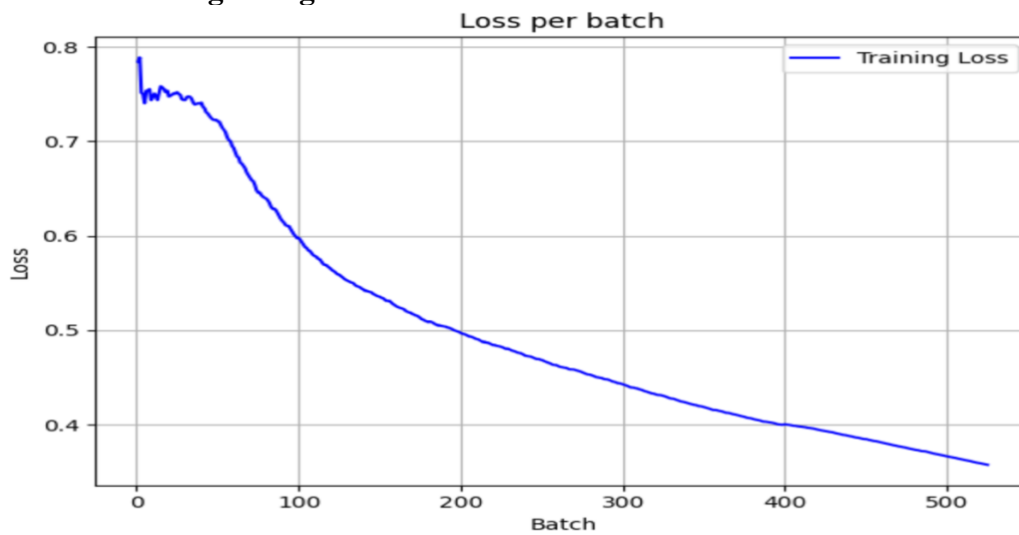
95HD: 9.047

## Training Loss curves

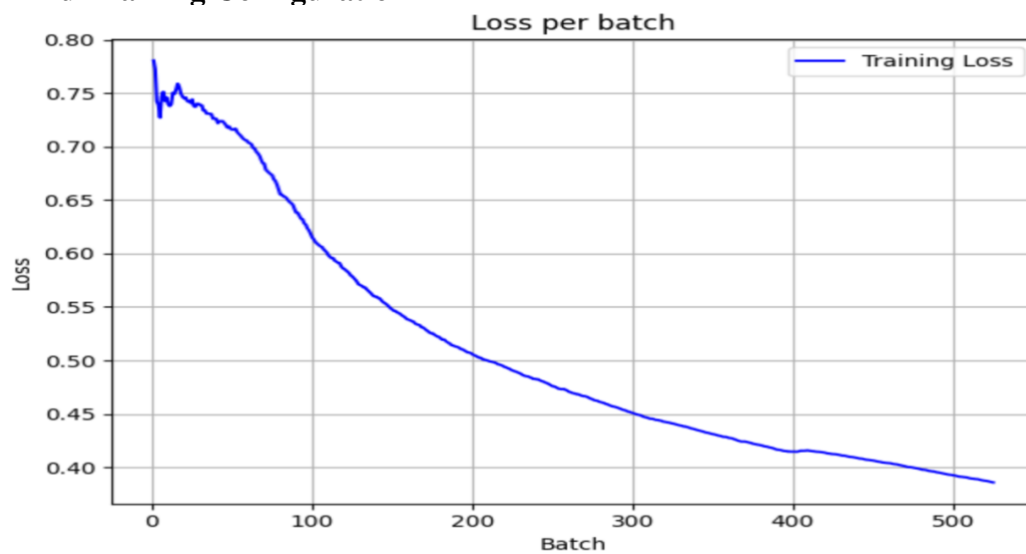
First Training configuration



Second Training configuration



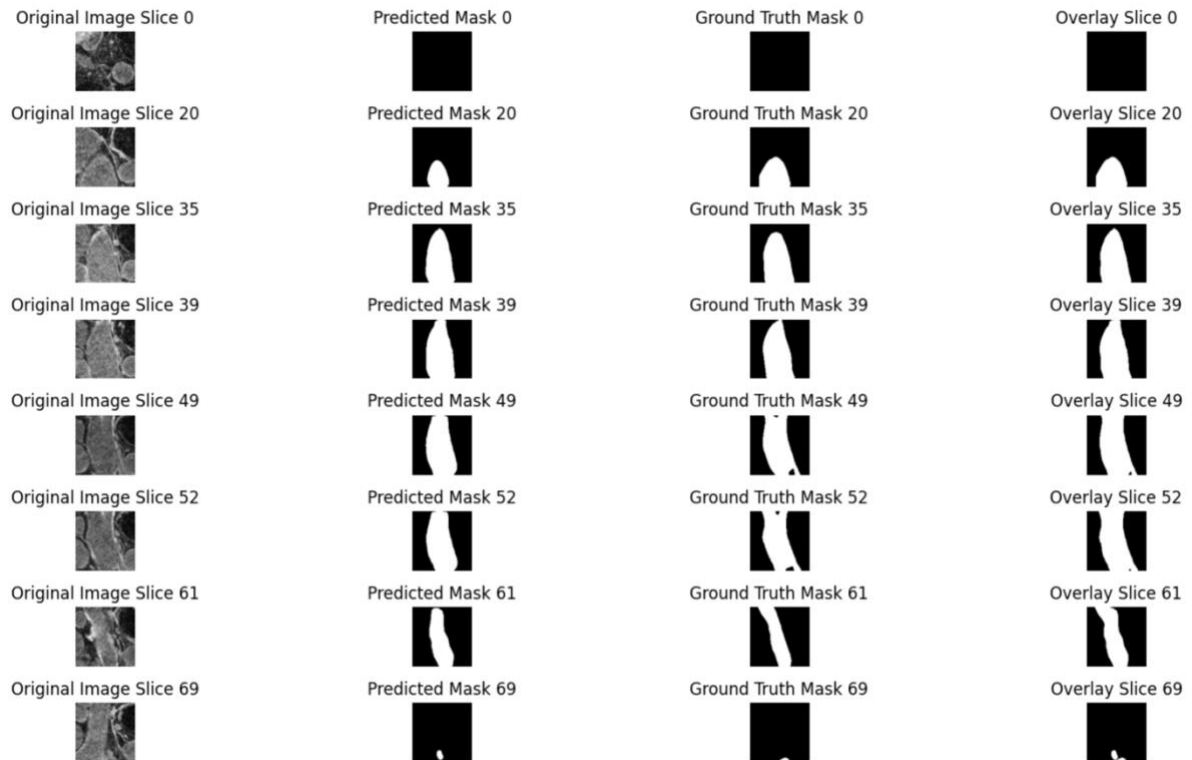
Third Training Configuration





*According to the above metrics on test dataset, it is evident that the second training configuration provides the best model, hence we use that model for the below example segmentations.*

## 2D segmentation results compared with ground truth label



## Visualising prediction masks in 3D

