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Course: Deep Learning for Computer Vision

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Homework 02

Problem 1:

Part 1.1:

Which data augmentation techniques being used:

- Applied random horizontal flips with probability of 0.5.
- \bullet Applied random rotation in between -30° to 30° with probability of 0.5.

How data normalized:

• I used mean and standard deviation of the ImageNet to normalize the dataset.

Part 1.2.1 and Part 1.2.2:

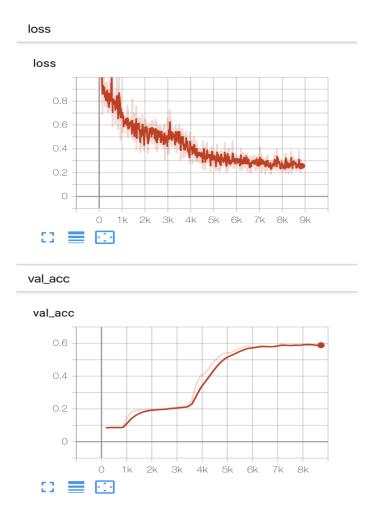


Figure 1: First training: ResNet18 parameters are freezed

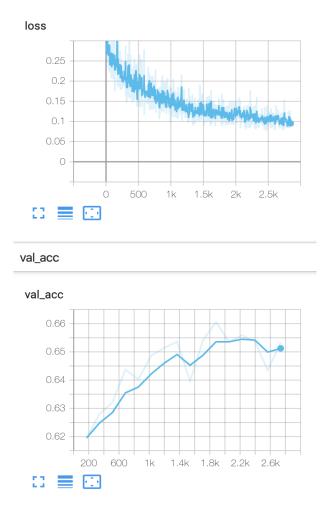


Figure 2: Second Training: Un-freezing the ResNet18 parameters and Fine Tune

Part 1.3:



Figure 3: Class 1: Person

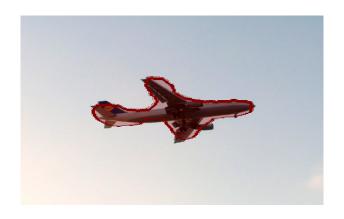


Figure 4: Class 2: Aeroplane



Figure 5: Class 3: Bus

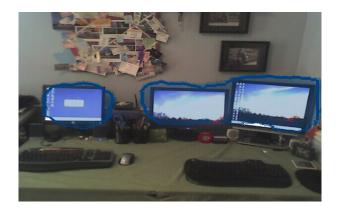


Figure 6: Class 4: TV/Monitor

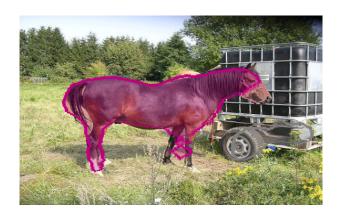


Figure 7: Class 5: Horse



Figure 8: Class 6: Dog

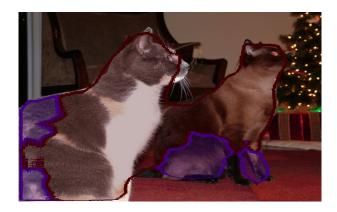


Figure 9: Class 7: Cat



Figure 10: Class 8: Car

Part 1.4:

```
==> prepare data loader ...

class #0 : 0.90130

class #1 : 0.73386

class #2 : 0.66845

class #3 : 0.70397

class #4 : 0.38061

class #5 : 0.51795

class #6 : 0.66064

class #7 : 0.74053

class #8 : 0.70904

mean iou: 0.668482
```

Testing Accuracy: 0.6684824296634783

Figure 11: mIoU scores of the Baseline Model

The class 0 which is the class belonging background has the highest IoU score. On the other hand, the class 4 which is the class belonging to the tv/monitor has the lowest IoU score.

The reason behind this result can be related with the content of the dataset given in the hw2 and the ImageNet dataset. The ResNet18 classifiers are pre-trained with the ImageNet dataset so they learnt detecting the objects which are most common in the ImageNet dataset. Additionally, the top feature extractors of the baseline model are trained with the dataset given in the hw2. Most part of the images in the dataset contains pixels belonging the background class and there are fewer pixels which labelled as tv/monitor class.

Part 2.1:

```
BaselineModel(
  (resnet34): Sequential(
  (conv transpose1): ConvTranspose2d(512, 256,
kernel\_size=(4, 4), stride=(2, 2), padding=(1, 1),
bias=False)
  (relu1): ReLU(inplace)
  (conv transpose2): ConvTranspose2d(256, 128,
kernel\_size=(4, 4), stride=(2, 2), padding=(1, 1),
bias=False)
  (relu2): ReLU(inplace)
  (conv transpose3): ConvTranspose2d(128, 64,
kernel_size=(4, 4), stride=(2, 2), padding=(1, 1),
bias=False)
  (relu3): ReLU(inplace)
  (conv transpose4): ConvTranspose2d(64, 32,
kernel size=(4, 4), stride=(2, 2), padding=(1, 1),
bias=False)
  (relu4): ReLU(inplace)
  (conv transpose5): ConvTranspose2d(32, 16,
kernel size=(4, 4), stride=(2, 2), padding=(1, 1),
bias=False)
  (relu5): ReLU(inplace)
  (conv1): Conv2d(16, 9, kernel size=(1, 1),
stride=(1, 1))
```

Figure 12: Improved Baseline Model Architecture

Part 2.2:

According to the training information belonging to the Baseline Model, the model converges after 7k epochs which is pushed forward by unfreezing and fine tuning the pretrained Resnet18 parameters. The fine tuned network also seems to converge after 2.5k epochs from start of the new training. The validation accuracy does not go down as I trained further the network which is a good indication of the model does not start to over-fit the data yet. So I decided to choose a deeper neural network architecture rather than using regularization techniques such as Dropout. I choose ResNet34 because it's demonstrated performance on benchmark datasets outperforms ResNet18. I keep architecture of the other layers same.

Part 2.3:

mean_iou: 0.723660

Testing Accuracy: 0.7236596400012784

Figure 13: Validation Set mIoU Score of the improved model



Figure 14: Improved Model Prediction: Class Dog



Figure 15: Baseline Model Prediction: Class Dog



Figure 16: Improved Model Prediction: Class Bus



Figure 17: Baseline Model Prediction: Class Bus



Figure 18: Improved Model Prediction: Class People



Figure 19: Baseline Model Prediction: Class People

References

I used Google Colab's GPU: Tesla K80

 $1.\ https://pytorch.org/docs/stable/torchvision/$