CS 6476 Project 6

Cameron Potter cgpotter@gatech.edu cpotter8 903465425

Parts 4 & 5: mIoU of different models

Add each of the following (keeping the changes as you move to the next row):

	Training mIoU	Validation mIoU
Simple Segmentation Net (no pretrained weights)	0.3838	0.3504
+ ImageNet-Pretrained backbone	0.5307	0.5096
+ Data augmentation	0.4147	0.4448
ImageNet-Pretrained PSPNet w/ Data Aug. without PPM	0.5325	0.5432
+ PSPNet with PPM	0.5362	0.5362
+ PSPNet with auxiliary loss	0.5189	0.5386

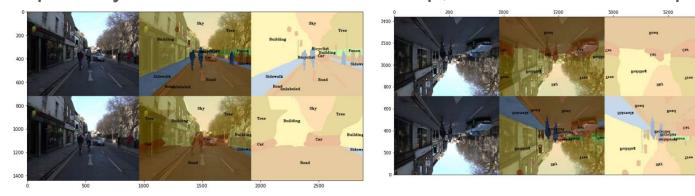
Parts 4 & 5: Per class IoUs

Report your model's IoU for the 11 Camvid classes (you can find the order they are listed in at dataset_lists/camvid-11/camvid-11_names.txt):

Class Index	Class name	Simple Segmentation Net Class IoU	PSPNet Class IoU
0	Building	0.5452	0.5452
1	Tree	0.6363	0.6363
2	Sky	0.8519	0.8519
3	Car	0.2300	0.2300
4	SignSymbol	0.0000	0.0000
5	Road	0.7198	0.7198
6	Pedestrian	0.0000	0.0000
7	Fence	0.0000	0.0000
8	Column_Pole	0.0000	0.0000
9	Sidewalk	0.0625	0.0625
10	Bicyclist	0.0000	0.0000

Parts 4 & 5: Most difficult classes

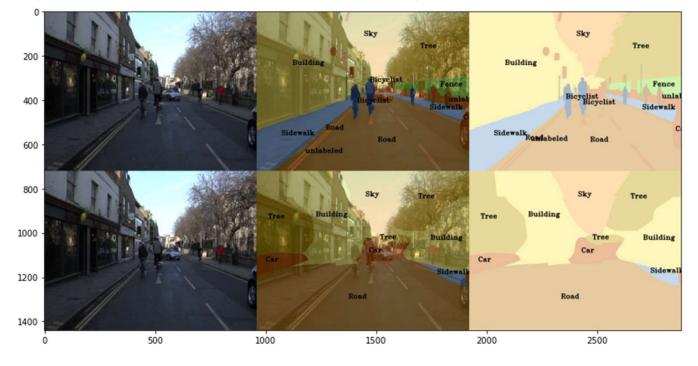
[Which classes have the lowest mIoU? Why might they be the most difficult? Provide an example RGB image from Camvid that illustrates your point] SignSymbol, Pedestrian, Fence, Column_Pole, and Bicyclist. These are the most difficult as they are smaller, specific features in an image, compared to things like Road or Sky, which are large areas of the image. Differentiating these or noticing them in the middle of one of the larger classes will inherently be more difficult, especially with a smaller feature map, as shown in these plots.



Part 4: Simple segmentation net qualitative results

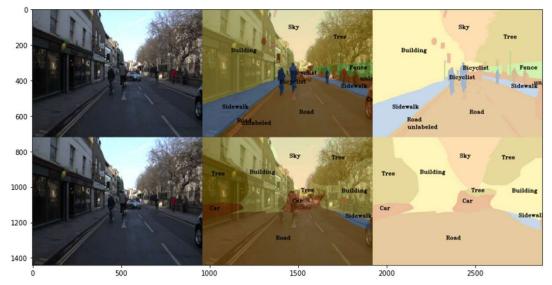
[Paste a figure of the generated semantic segmentation from Colab. It should be a 2x3 grid, with ground truth on the top row, and your predictions on the bottom

row.]



Part 5: PSPNet qualitative results

[Paste a figure of the generated semantic segmentation from Colab. It should be a 2x3 grid, with ground truth on the top row, and your predictions on the bottom row.]



Part 6: Transfer Learning

Report your model's IoU for the Kitti Dataset.

	mloU	mAcc/	allAcc
Train result	0.8898	0.9344	0.9652
Val result	0.8867	0.9334	0.9641

Class Index	Class name	iou	accuracy
0	Road	0.8161	0.8856
1	Not_Road	0.9573	0.9813

Part 6: Transfer Learning

Compare the training loss generated when training on Kitti dataset and Camvid dataset. Which decreases at a faster rate? If Camvid or Kitti training loss decreases at a faster rate than the other, why do you think this happened? Or, if the loss decreases at a similar rate, why do you think that is so?

The Kitti loss decreases at a much faster rate. This is most likely because it is a binary classifier with just "road" or "not road". Since Camvid has much more features to analyze, it's loss is inevitibly going to be harder to decrease. Thus, it is expected that Kitti will decrease loss faster.