

## Report DLCV Homework 2

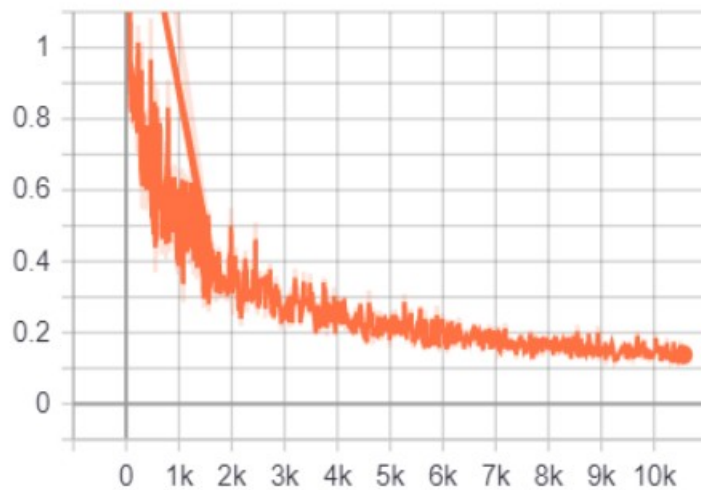
### Problem 1

#### 1. Baseline model

1.1. I only normalized the data with  $\text{MEAN} = [0.485, 0.456, 0.406]$  and  $\text{STD} = [0.229, 0.224, 0.225]$ .

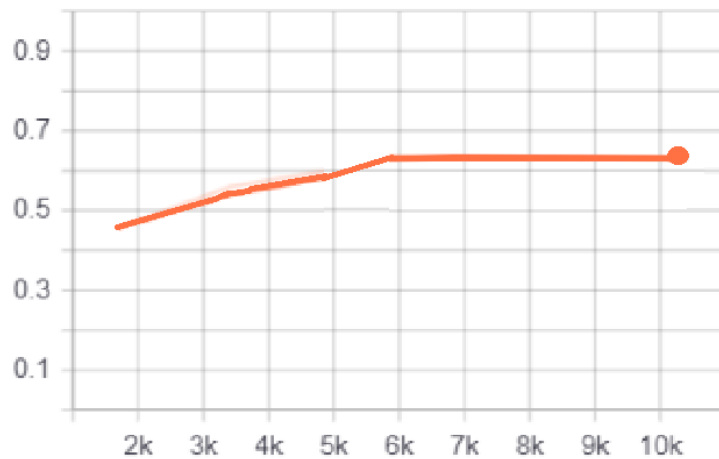
1.2. Training loss versus number of training iterations

loss

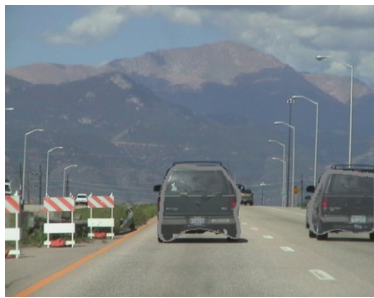


IoU score on validation set versus number of training iterations

val\_acc



### 1.3. Visualization of one semantic segmentation result for each class



Class 0: background



Class 1: person



Class 2: aeroplane



Class 3: bus



Class 4: Monitor



Class 5: horse



Class 6: dog



Class 7: cat



Class 8: car

### 1.4.

Validation mIoU score of the baseline model: 0.6270

Validation per-class IoU score of the baseline model:

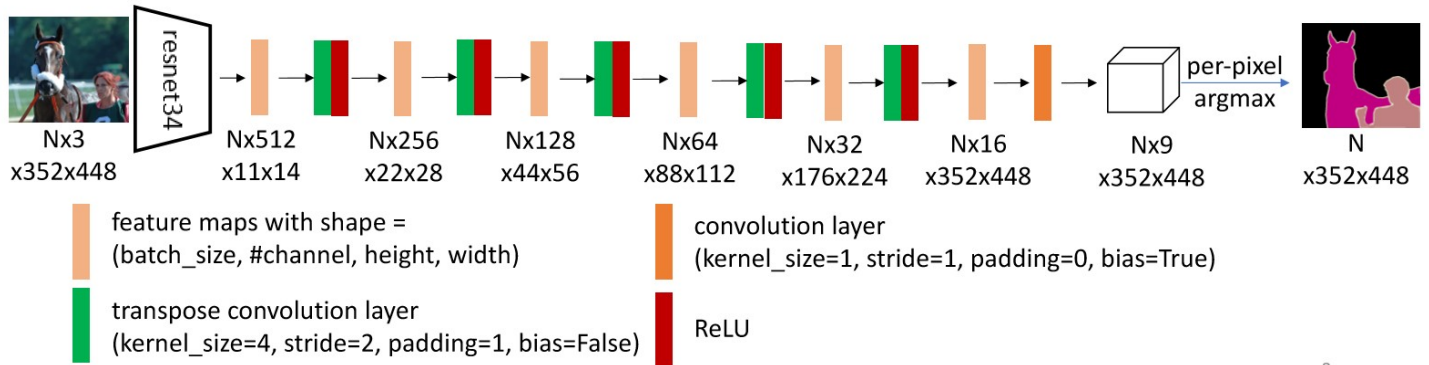
class #0 : 0.89724 class #1 : 0.74072 class #2 : 0.64078 class #3 : 0.72161  
class #4 : 0.36778 class #5 : 0.42074 class #6 : 0.56239 class #7 : 0.69957  
class #8 : 0.59262

The TV-monitor class has the lowest IoU score. It may be because it is the hardest object to differentiate from its surrounding since there often is another image pictured in the monitor.

The background class has the highest IoU score, maybe because it is the largest « object » in each image thus the model can learn more about it than for other classes. Moreover the model has a lot of data to learn from since every picture has a background.

## 2. Improved model

2.1 The model architecture of my improved model is as follows:



8

I simply changed resnet18 for resnet34.

### 2.2

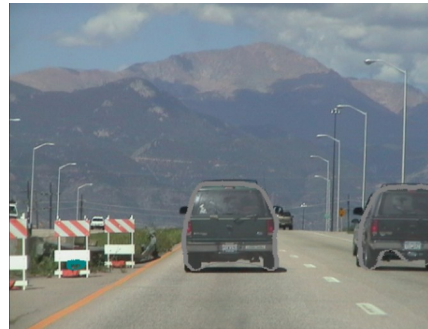
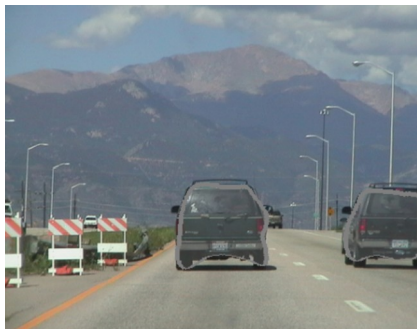
Resnet34 has 34 layers in the network. It is therefore a larger network than resnet18 and a larger network can model more complex problems.

### 2.3

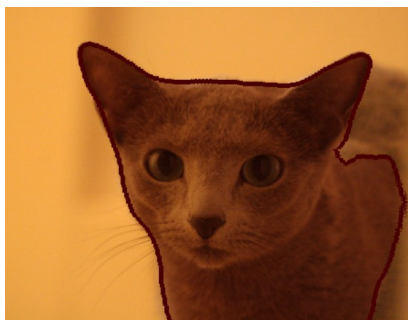
Validation mIoU score of the improved model: 0,6596

Segmentation of the baseline model:

Segmentation of the improved model:







## Problem 2

1.

① A 2D Gaussian filter is written as :

$$G(x, y) = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{x^2 + y^2}{2\sigma^2}\right)$$
$$= \frac{1}{\sqrt{2\pi}\sigma} \times \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x^2}{2\sigma^2}\right) \exp\left(-\frac{y^2}{2\sigma^2}\right)$$
$$= \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x^2}{2\sigma^2}\right) \times \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{y^2}{2\sigma^2}\right)$$
$$= G(x) \times G(y) \quad \text{two 1D Gaussian filters}$$

2.

The Gaussian filter is blurring the image : it reduces the image noise and reduces details which smoothes the image.

Input image



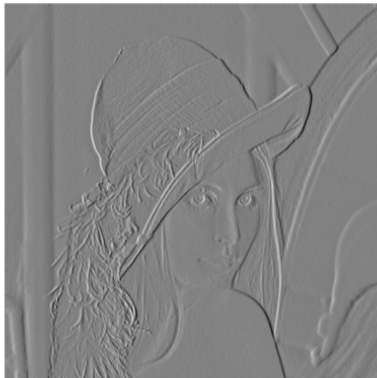
Output image



3.

$k_x = [-1/2, 0, 1/2]$  and  $k_y = [-1/2, 0, 1/2]^T$

$I_x$  output



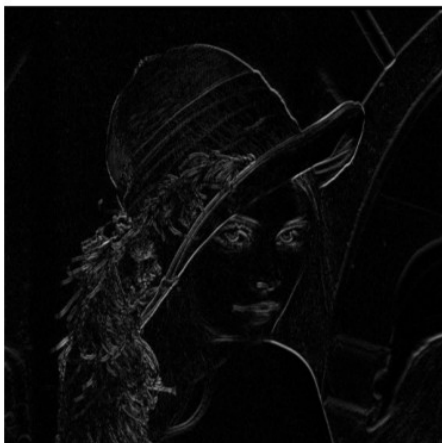
$I_y$  output



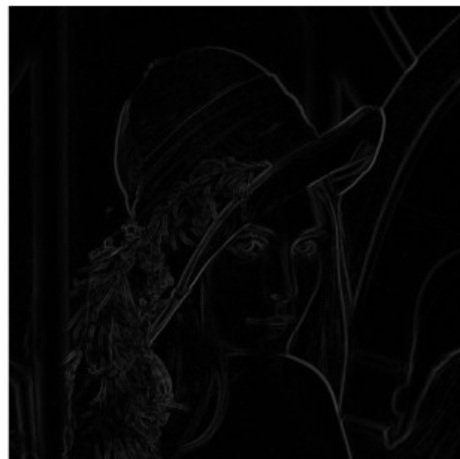
4.

The Gaussian filter removed noise from the image thus it has less detail and contrast than the original image when visualising the gradient magnitude.

Original image



Gaussian filtered image



Collaborators: Eloi Dussy-Lachaud  
Thibaut Fenain