

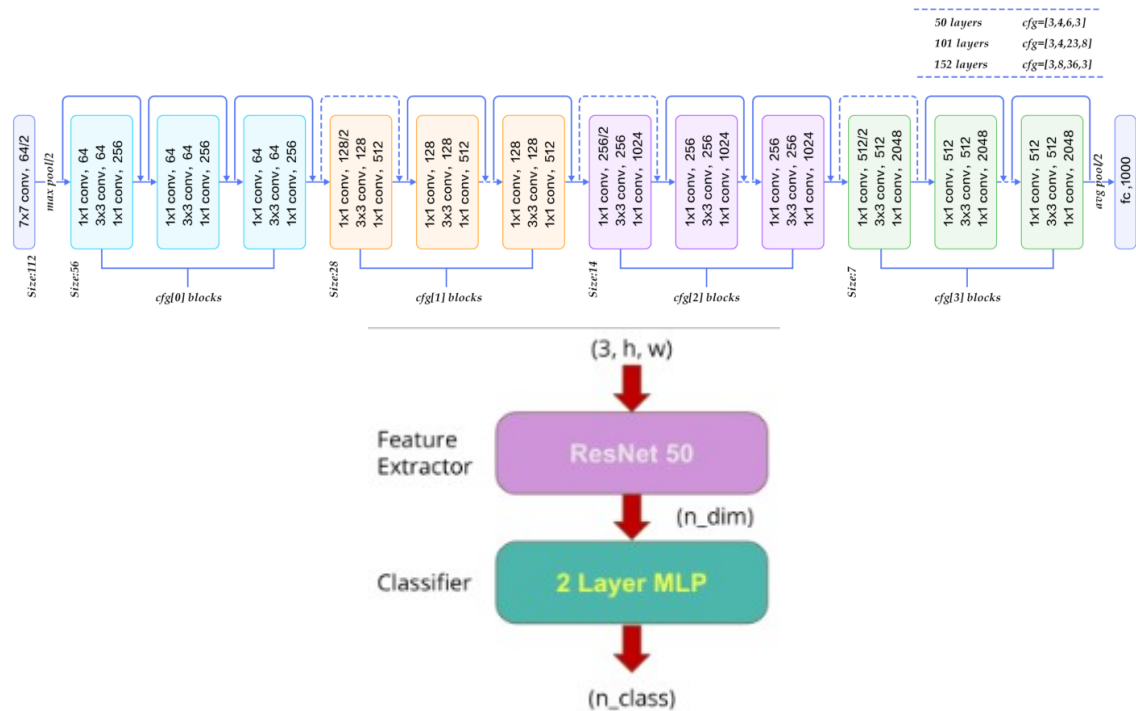
Report

Problem 1

1. Draw the network architecture of method A or B.

- The graph should be brief and clear
- It would be fine to straight copy the figure from the paper

Model B



2. Report accuracy of your models (both A, B) on the validation set.

Model	Mine CNN (from scratch)	Resnet50 (pretrain)
Accuracy	0.37	0.864

3. Report your implementation details of model A.

- Including but not limited to optimizer, loss function, cross validation method

由於是圖片的緣故，第一個想法就是套用CNN，然後因為圖片原本很小，所以有用到 ConvTranspose。

optimizer使用的是AdamW，可以快速收斂

loss function使用CrossEntropy，分類為題最常用的

cross validation用validation set來決定model的好壞而調整參數

4. Report your alternative model or method in B, and describe its difference from model A

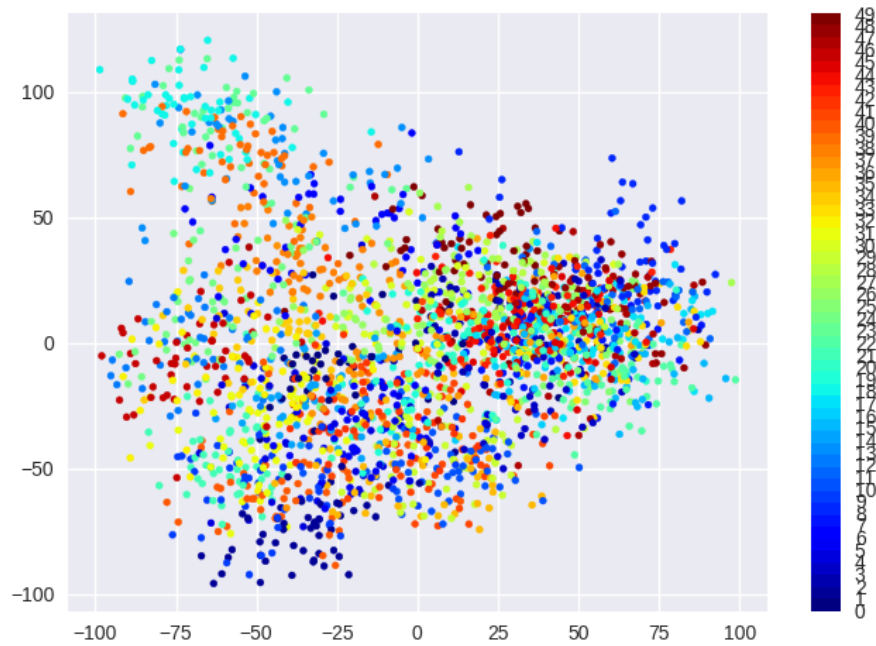
Resnet50 skip connection可以幫助model在深層時避免gradient vanishing，可以有效學習更多 features

5. Visualize the learned visual representations of model A on the validation set by

implementing PCA (Principal Component Analysis) on the output of the second last layer.

Briefly explain your result of the PCA visualization.

PCA

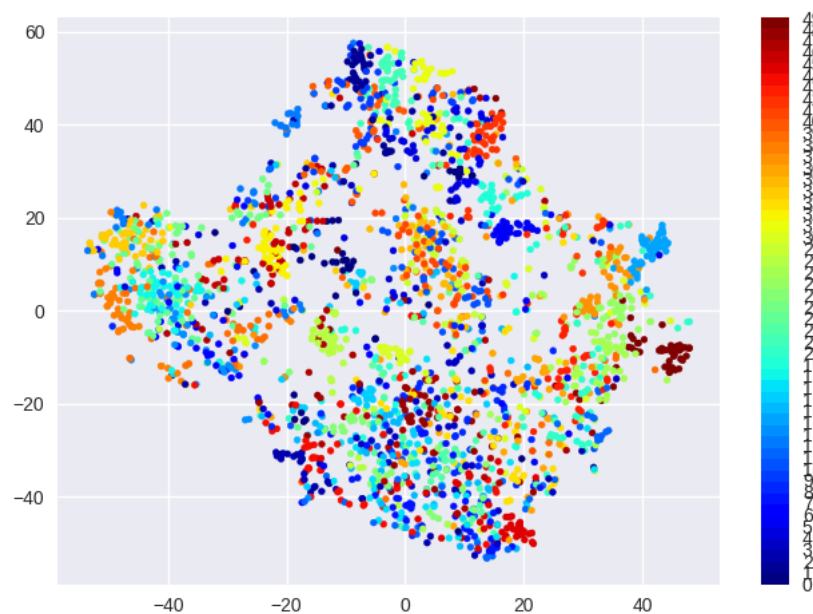


在PCA的情形下，由於NN是非線性的，但PCA是用線性的方式去降維，因此分析並不是到太符合我們要的資訊，他散落的程度還是不小。

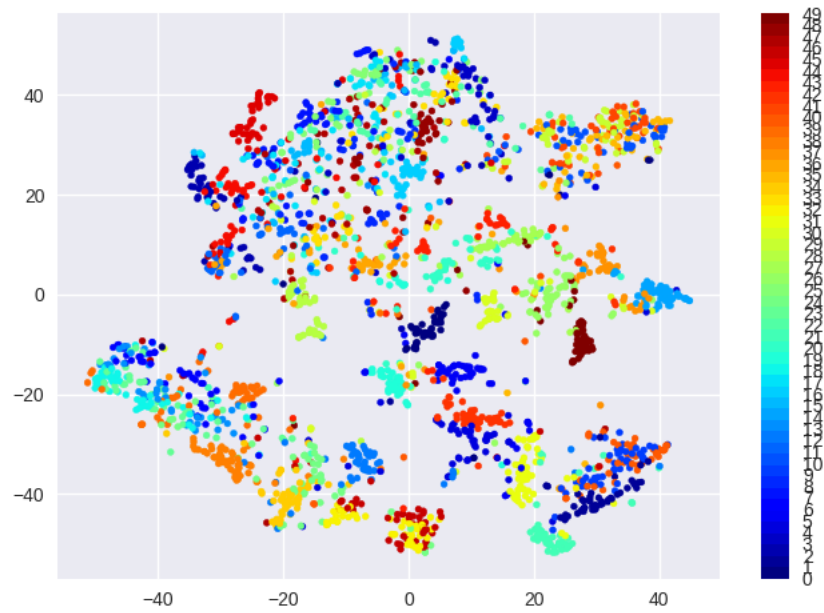
6. Visualize the learned visual representation of model A, again on the output of the second last layer, but using t-SNE (t-distributed Stochastic Neighbor Embedding) instead. Depict your visualization from three different epochs including the first one and the last one. Briefly explain the above results.

TSNE

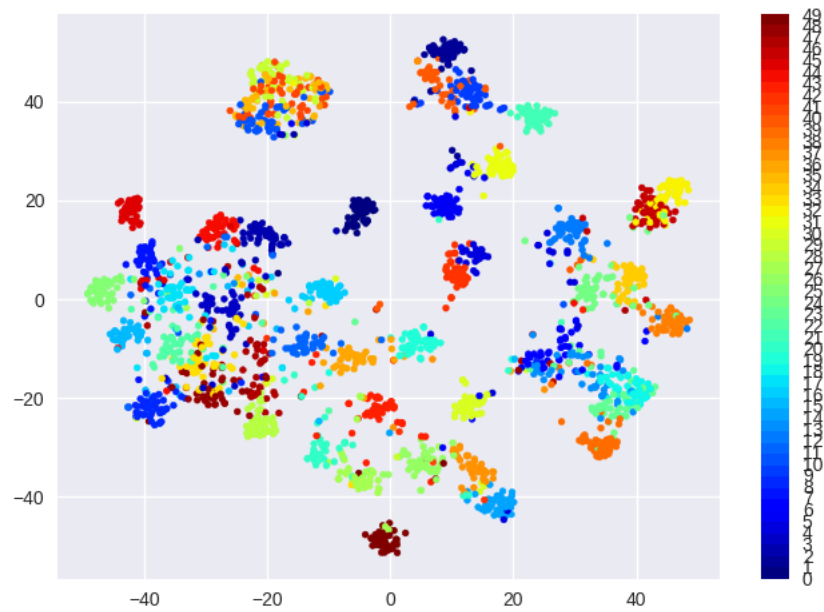
- first



○ middle



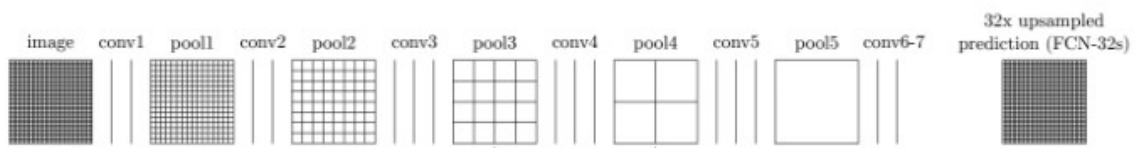
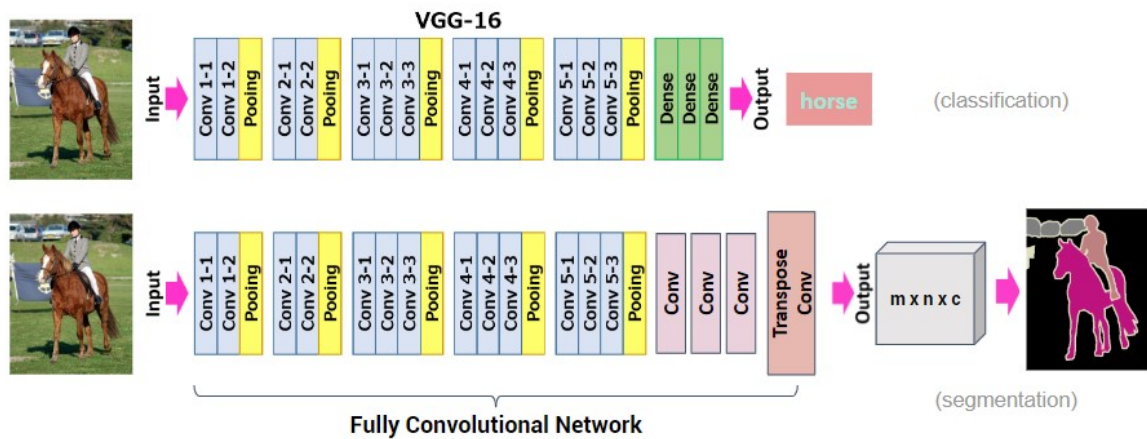
○ last



t-SNE看起來比PCA的分布還要群聚，也許是因為他本身是非線性的緣故比較貼合Model，可議看出他比較沒有同個顏色散落在各處的情形，因此認為model train的還不錯

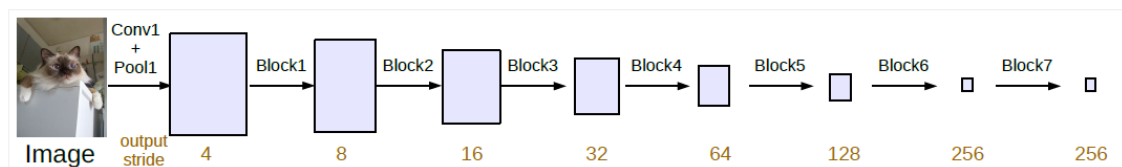
Problem 2

1. Draw the network architecture of your VGG16-FCN32s model (model A).

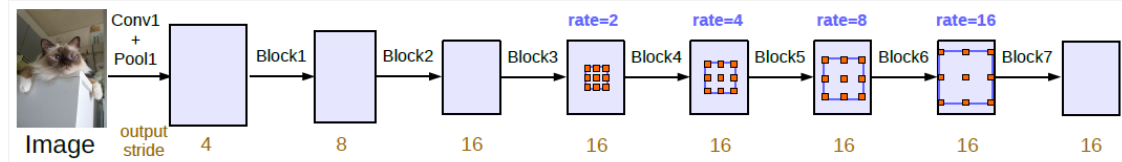


2. Draw the network architecture of the improved model (model B) and explain it differs from your VGG16-FCN32s model.

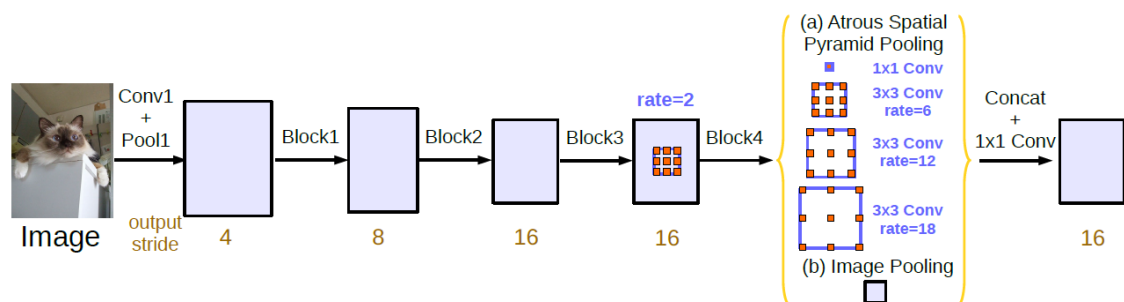
DeepLab v3



(a) Going deeper without atrous convolution.



(b) Going deeper with atrous convolution. Atrous convolution with $rate > 1$ is applied after block3 when $output_stride = 16$.








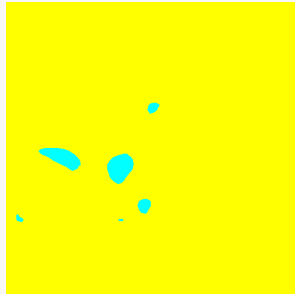


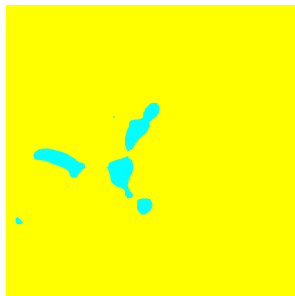





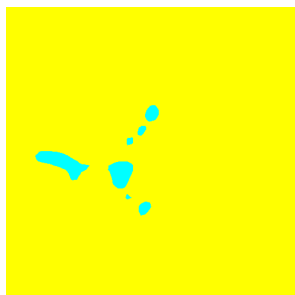
1. backbone變強了(VGG16-Resnet-50)
2. ASPP能關注不同範圍的feature
3. Report mIoUs of two models on the validation set.




Epoch / Model	VGG16_FCN32s	DEEPLAB v3
5	0.61	0.71
10	0.68	0.72
15	0.69	0.73
20	0.70	0.73

Epoch / Model	VGG16_FCN32s	DEEPLAB v3
25	0.69	0.73

4. Show the predicted segmentation mask of “validation/0013_sat.jpg”, “validation/0062_sat.jpg”, “validation/0104_sat.jpg” during the early, middle, and the final stage during the training process of the improved model.

- Tips: Given n epochs training, you could save the 1st, (n/2)-th, n-th epoch model, and draw the predicted mask by loading these saved models.

Epoch	0013_sat	0062_sat	0104_sat
5			
10			
15			
20			
25			

Epoch	0013_sat	0062_sat	0104_sat
Ground Truth			
Image	