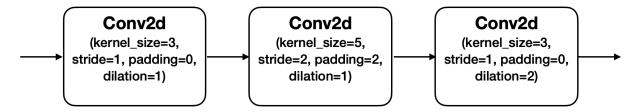
Al6126: Assignment 2

Deadline: 31 March 2023 11:59PM

Question 1: Segmentation-related questions. (7 marks)

- i) What is the motivation for using downsampling and upsampling in the Fully Convolutional Network for semantic segmentation?
- ii) Given the following network, calculate the receptive field.



- iii) What are the advantages of dilated convolution over standard convolution for the semantic segmentation task?
- iv) Given a transposed convolution kernel as follow, whose stride=1, padding=0, dilation=1,

$$\begin{vmatrix}
1 & 2 & 3 \\
2 & 3 & 4 \\
3 & 4 & 5
\end{vmatrix}$$

and input,

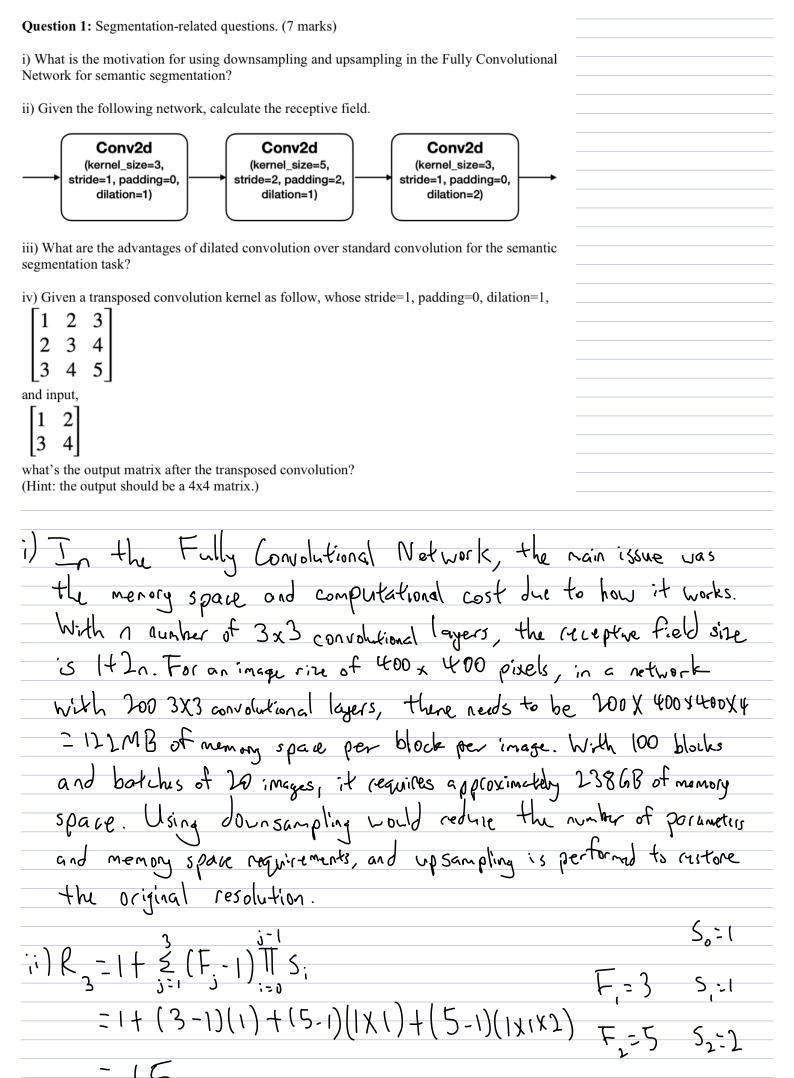
$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

what's the output matrix after the transposed convolution?

(Hint: the output should be a 4x4 matrix.)

Question 2: What is the key difference between VQVAE and VAE? (2 marks)

Question 3: What is training objective of Generative Adversarial Networks? Please answer this question conceptually, i.e., do not just posing mathematical equations. (3 marks)



Dilated convolution provides a larger field of view while maintaining the same feature map output size. By stacking several convolutions of different dilation values, it is possible to get finer segmentation feature maps as they contain information of different sizes of sub-regions of the image.

Question 2: What is the key difference between VQVAE and VAE? (2 marks)
VQ VAE utilizes vector quantization, which is the use of a vector
which is quantized from an input image in order to get the
latent representation, whereas in VAE, the latent representation
is generated from a coatinuous probability distribution, such as Gaussian.
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Question 3: What is training objective of Generative Adversarial Networks? Please answer this question conceptually, i.e., do not just posing mathematical equations. (3 marks)
GAN's consists of a generator network and a discriminator
network. The generator attempts to make realistic images based
on randon noise, and the discriminator attempts to differentiate the
realistic but fake images produced by the generator from the real sample
images. As training progresses, the gonerator develops bother images to
fool the discriminator while the discriminator improves its detection
to determine whether the input image is real or fake. This
training process becomes a minimax game where the generator
attempts to minimize the discriminator's detection capability while the
discriminator attempts to maximise its detertion to differentiate
the real sample images from the fake.