

⚠ Try again once you are ready

Grade received 70% Latest Submission Grade 70% To pass 80% or higher

Try again

1. what do you think applying this filter to a grayscale image will do:

1 / 1 point

$$\begin{bmatrix} 0 & 1 & -1 & 0 \\ 1 & 3 & -3 & -1 \\ 1 & 3 & -3 & -1 \\ 0 & 1 & -1 & 0 \end{bmatrix}$$

- ☐ Detect 45 degree edges
- ☐ Detect horizontal edges
- ☐ Detect image contrast
- ☒ Detect vertical edges

↗ Expand

✓ Correct

Correct! As you can see the difference between values from the left part and values from the right of this filter is high. When convolving this filter on a grayscale image, the vertical edges will be detected.

2. Suppose your input is a 128 by 128 color (RGB) image, and you are not using a convolutional network. If the first hidden layer has 64 neurons, each one fully connected to the input, how many parameters does this hidden layer have (including the bias parameters)?

1 / 1 point

- ☐ 1048640
- ☐ 1048576
- ☒ 3145792
- ☐ 3145728

↗ Expand

✓ Correct

Correct, the number of inputs for each unit is  $128 \times 128 \times 3$  since the input image is RGB, so we need  $128 \times 128 \times 3 \times 64$  parameters for the weights and  $64$  parameters for the bias parameters, thus  $128 \times 128 \times 3 \times 64 + 64 = 3145792$ .

3. Suppose your input is a 300 by 300 color (RGB) image, and you use a convolutional layer with 100 filters that are each 5x5. How many parameters does this hidden layer have (including the bias parameters)?

1 / 1 point

- ☐ 7500
- ☒ 7600
- ☐ 2501
- ☐ 2600

↗ Expand

✓ Correct

Correct, you have  $25 \times 3 = 75$  weights and  $1$  bias per filter. Given that you have 100 filters, you get 7,600 parameters for this layer.

4. You have an input volume that is  $121 \times 121 \times 16$ , and convolve it with 32 filters of  $4 \times 4$ , using a stride of 3 and no padding. What is the output volume?

0 / 1 point

- ☒  $118 \times 118 \times 32$
- ☐  $118 \times 118 \times 16$
- ☐  $40 \times 40 \times 16$
- ☐  $40 \times 40 \times 32$

[Expand](#)

 **Incorrect**

No, check that you use  $s=3$  for the stride when computing the dimensions of the output volume.

5. You have an input volume that is  $31 \times 31 \times 32$ , and pad it using “pad=1”. What is the dimension of the resulting volume (after padding)?

1 / 1 point

- ☐  $31 \times 31 \times 34$
- ☐  $32 \times 32 \times 32$
- ☐  $33 \times 33 \times 33$
- ☒  $33 \times 33 \times 32$

[Expand](#)

 **Correct**

Yes, if the padding is 1 you add 2 to the height dimension and 2 to the width dimension.

6. You have an input volume that is  $63 \times 63 \times 16$ , and convolve it with 32 filters that are each  $7 \times 7$ , and stride of 1. You want to use a “same” convolution. What is the padding?

1 / 1 point

- ☐ 2
- ☐ 7
- ☐ 1
- ☒ 3

[Expand](#)

 **Correct**

Correct, you need to satisfy the following equation:  $n_H - f + 2 \times p + 1 = n_H$  as you want to keep the dimensions between the input volume and the output volume.

7. You have an input volume that is  $66 \times 66 \times 21$ , and apply max pooling with a stride of 3 and a filter size of 3. What is the output volume?

1 / 1 point

- ☐  $21 \times 21 \times 21$
- ☒  $22 \times 22 \times 21$
- ☐  $66 \times 66 \times 7$
- ☐  $22 \times 22 \times 7$

[Expand](#)

✓ **Correct**

Yes, using the formula  $n_H[l] = \frac{n_H[l-1]}{2} + 2 \times p - f + 1$  with  $p = 0$ ,  $f = 3$ ,  $s = 3$  and  $n_H[l-1] = 66$ .

8. Because pooling layers do not have parameters, they do not affect the backpropagation (derivatives) calculation.

0 / 1 point

☒ True

☐ False

↗ Expand

✗ **Incorrect**

Everything that influences the loss should appear in the backpropagation because we are computing derivatives. In fact, pooling layers modify the input by choosing one value out of several values in their input volume. Also, to compute derivatives for the layers that have parameters (Convolutions, Fully-Connected), we still need to backpropagate the gradient through the Pooling layers.

9. Which of the following are true about convolutional layers? (Check all that apply)

1 / 1 point

☒ It allows a feature detector to be used in multiple locations throughout the whole input volume.

✓ **Correct**

Yes, since convolution involves sliding the filter throughout the whole input volume the feature detector is computed over all the volume.

☐ It speeds up the training since we don't need to compute the gradient for convolutional layers.

☐ It allows parameters learned for one task to be shared even for a different task (transfer learning).

☒ Convolutional layers provide sparsity of connections.

✓ **Correct**

Yes, this happens since the next activation layer depends only on a small number of activations from the previous layer.

↗ Expand

✓ **Correct**

Great, you got all the right answers.

10. The sparsity of connections and weight sharing are mechanisms that allow us to use fewer parameters in a convolutional layer making it possible to train a network with smaller training sets. True/False?

0 / 1 point

☒ False

☐ True

↗ Expand

✗ **Incorrect**

No, weight sharing reduces significantly the number of parameters in a neural network, and sparsity of connections allows us to use a smaller number of inputs thus reducing even further the number of parameters.