

✓ Congratulations! You passed!

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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 horizontal edges
- ☒ Detect vertical edges
- ☐ Detect image contrast
- ☐ Detect 45 degree 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 grayscale image, and you are not using a convolutional network. If the first hidden layer has 256 neurons, each one fully connected to the input, how many parameters does this hidden layer have (including the bias parameters)?

1 / 1 point

- ☐ 12583168
- ☐ 4194304
- ☒ 4194560
- ☐ 12582912

↗ Expand

✓ Correct

Correct, the number of inputs for each unit is  $128 \times 128$  since the input image is grayscale, so we need  $128 \times 128 \times 256$  parameters for the weights and 256 parameters for the bias thus  $128 \times 128 \times 256 + 256 = 4194560$ .

3. Suppose your input is a 256 by 256 grayscale image, and you use a convolutional layer with 128 filters that are each  $3 \times 3$ . How many parameters does this hidden layer have (including the bias parameters)?

1 / 1 point

- ☐ 3584
- ☐ 1152
- ☒ 1280
- ☐ 75497600

↗ Expand

✓ Correct

Yes, since the input volume has only one channel each filter has  $3 \times 3 + 1$  weights including the bias, thus the total is  $(3 \times 3 + 1) \times 128$ .

4. You have an input volume that is  $127 \times 127 \times 16$ , and convolve it with 32 filters of  $5 \times 5$ , using a stride of 2 and no padding. What is the output volume?

1 / 1 point

- ☐  $62 \times 62 \times 16$
- ☐  $123 \times 123 \times 32$
-

- ☐  $123 \times 123 \times 16$
- ☒  $62 \times 62 \times 32$

Expand

Correct

Correct, using the formula  $n_H^{[l]} = \frac{n_H^{[l-1]} + 2 \times p - f}{s} + 1$  with  $n_H^{[l-1]} = 127$ ,  $p = 0$ ,  $f = 5$ , and  $s = 2$  we get 62.

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

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

Expand

Correct

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

6. You have a volume that is  $64 \times 64 \times 32$ , and convolve it with 40 filters of  $9 \times 9$ , and stride 1. You want to use a "same" convolution. What is the padding?

1 / 1 point

- ☐ 6
- ☒ 4
- ☐ 8
- ☐ 0

Expand

Correct

Yes, when using a padding of 4 the output volume has  $n_H = \frac{64 - 9 + 2 \times 4}{1} + 1$ .

7. You have an input volume that is  $32 \times 32 \times 16$ , and apply max pooling with a stride of 2 and a filter size of 2. What is the output volume?

1 / 1 point

- ☐  $15 \times 15 \times 16$
- ☒  $16 \times 16 \times 16$
- ☐  $32 \times 32 \times 8$
- ☐  $16 \times 16 \times 8$

Expand

Correct

Correct, using the following formula:  $n_H^{[l]} = \frac{n_H^{[l-1]} + 2 \times p - f}{s} + 1$

8. Which of the following are hyperparameters of the pooling layers? (Choose all that apply)

1 / 1 point

☒ Whether it is max or average.

Correct

Yes, these are the two types of pooling discussed in the lectures, and choosing which to use is considered a hyperparameter.

☒ Stride

Correct

Yes, although usually, we set  $s_f = s$  this is one of the hyperparameters of a pooling layer.

☐  $W^{\{i\}}$  weights.

☐  $b^{\{i\}}$  bias.

 Expand

 **Correct**

Great, you got all the right answers.

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

1 / 1 point

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

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

☒ 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.

☒ 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?

1 / 1 point

☐ False

☒ True

 Expand

 **Correct**

Yes, 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.