

Your grade: 100%

Your latest: **100%** • Your highest: **100%**

To pass you need at least 80%. We keep your highest score.

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item



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.
- ☐ Detecting image contrast.
- ☐ Detect vertical edges.

✓ **Correct**

Correct. There is a high difference between the values in the top part from those in the bottom part of the matrix. When convolving this filter on a grayscale image, the horizontal 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
- ☐ 3145728

☐ 1048576

☒ 3145792

✓ **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

☐ 2501

☒ 7600

☐ 7500

☐ 2600

✓ **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 $127 \times 127 \times 16$, and convolve it with 32 filters of 5×5 , using a stride of 2 and no padding. What is the output volume?

1 / 1 point

☐ $123 \times 123 \times 16$

☒ $62 \times 62 \times 32$

☐ $123 \times 123 \times 32$

☐ $62 \times 62 \times 16$

☒ **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 $15 \times 15 \times 8$, and pad it using “pad=2”. What is the dimension of the resulting volume (after padding)?

1 / 1 point

☐ $17 \times 17 \times 8$

☐ $17 \times 17 \times 10$

☒ $19 \times 19 \times 8$

☐ $19 \times 19 \times 12$

☒ **Correct**

Correct, padding is applied over the height and the width of the input image. If the padding is two, you add 4 to the height dimension and 4 to the width dimension.

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

1 / 1 point

☐ 8

☒ 4

☐ 6

☐ 0

☒ **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 128x128x12, and apply max pooling with a stride of 4 and a filter size of 4. What is the output volume?

1 / 1 point

☐ $64 \times 64 \times 12$
☐ $128 \times 128 \times 3$
☒ $32 \times 32 \times 12$
☐ $32 \times 32 \times 3$
☒ **Correct**

Yes, using the formula $n_H^{[l]} = \frac{n_H^{[l-1]} + 2 \times p - f}{s} + 1$ with $p = 0, f = 4, s = 4$ and $n_H^{[l-1]} = 32$.

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.

☒ Filter size.

☒ **Correct**

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

☐ Average weights.

☐ Number of filters.

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

1 / 1 point

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

☒ Convolutional layers provide sparsity of connections.

✔ **Correct**

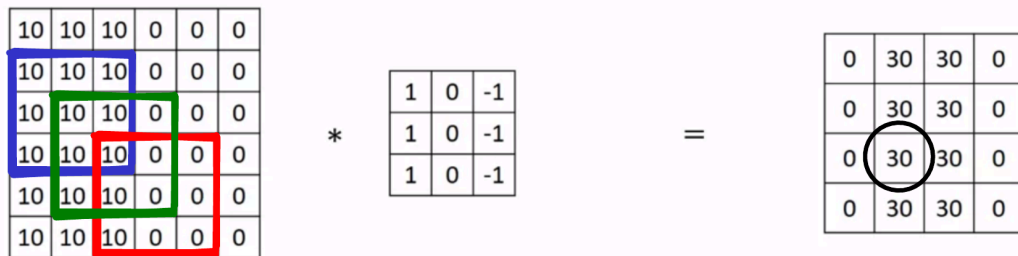
Yes, this happens since the next activation layer depends only on a small number of activations from the previous layer.

☒ 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.

10. The following image depicts the result of a convolution at the right when using a stride of 1 and the filter is shown right next.



- ☒ It depends on the pixels enclosed by the green square.
- ☐ It depends on the pixels enclosed by the blue square.
- On which pixels does the circled pixel of the activation at the right depend?
- ☐ It depends on the pixels enclosed by the red square.
- ☐ It depends on all the pixels of the image on the left.

✓ **Correct**

Yes, this is the position of the filter when we move it two pixels down and one to the right.