

✓ Congratulations! You passed!

Go to next item

Grade received 100% Latest Submission Grade 100% To pass 80% or higher

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 image contrast
- ☐ Detect 45 degree edges
- ☐ Detect horizontal edges
- ☒ 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 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
- ☐ 12582912
- ☐ 4194304
- ☒ 4194560

↗ Expand

✓ Correct

Correct, the number of inputs for each unit is 128×128 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 color (RGB) image, and you use a convolutional layer with 128 filters that are each 7×7 . How many parameters does this hidden layer have (including the bias parameters)?

1 / 1 point

- ☒ 18944
- ☐ 6400
- ☐ 1233125504
- ☐ 18816

↗ Expand

✓ Correct

Yes, you have $7 \times 7 \times 3 + 1$ weights per filter with the bias. Given that you have 128 filters, you get $(7 \times 7 \times 3 + 1) \times 128 = 18944$.

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

1 / 1 point

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

↗ Expand

✓ Correct

Correct, using the formula $n_H^{[l]} = \frac{n_H^{[l-1]} + 2 \times p - f}{s} + 1$ with $n_H^{[l-1]} = 121, p = 0, f = 4$, and $s = 3$ we get 40

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

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

↗ 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 9×9 , and stride 1. You want to use a "same" convolution. What is the padding?

1 / 1 point

- ☒ 4
- ☐ 6
- ☐ 0
- ☐ 8

↗ 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

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

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

- ☐ Average weights.
- ☐ Number of filters.
- ☒ 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

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

- ☒ 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.

- ☐ 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).

Expand

Correct

Great, you got all the right answers.

10. In lecture we talked about "sparsity of connections" as a benefit of using convolutional layers. What does this mean?

1 / 1 point

- ☐ Each layer in a convolutional network is connected only to two other layers
- ☐ Regularization causes gradient descent to set many of the parameters to zero.
- ☒ Each activation in the next layer depends on only a small number of activations from the previous layer.
- ☐ Each filter is connected to every channel in the previous layer.

Expand

✓ **Correct**

Yes, each activation of the output volume is computed by multiplying the parameters from `W` with a volumic slice of the input volume and then summing all these together.