

Special Applications: Face Recognition & Neural Style Transfer

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1. Face verification and face recognition are the two most common names given to the task of comparing a new picture against one person's face. True/False?

1 / 1 point

☒ False

☐ True

↗ Expand

✓ Correct

Correct. This is the description of face verification, but not of face recognition.

2. Why is the face verification problem considered a one-shot learning problem? Choose the best answer.

1 / 1 point

- ☐ Because of the sensitive nature of the problem, we won't have a chance to correct it if the network makes a mistake.
- ☒ Because we might have only one example of the person we want to verify.
- ☐ Because we are trying to compare to one specific person only.
- ☐ Because we have only have to forward pass the image one time through our neural network for verification.



Correct

Correct. One-shot learning refers to the amount of data we have to solve a task.

3. In order to train the parameters of a face recognition system, it would be reasonable to use a training set comprising 100,000 pictures of 100,000 different persons.

1 / 1 point



False



True



Expand



Correct

Correct, to train a network using the triplet loss you need several pictures of the same person.

4. In the triplet loss:

1 / 1 point

$$\max \left(\|f(A) - f(P)\|^2 - \|f(A) - f(N)\|^2 + \alpha, 0 \right)$$

Which of the following are true about the triplet loss? Choose all that apply.



We want that $\|f(A) - f(P)\|^2 < \|f(A) - f(N)\|^2$ so the negative images are further away from the anchor than the positive images.



Correct

Correct. Being a positive image the encoding of P should be close to the encoding of A .



$f(A)$ represents the encoding of the Anchor.



Correct

Correct. f represents the network that is in charge of creating the encoding of the images, and A represents the anchor image.



α is a trainable parameter of the Siamese network.



A the anchor image is a hyperparameter of the Siamese network.



Expand

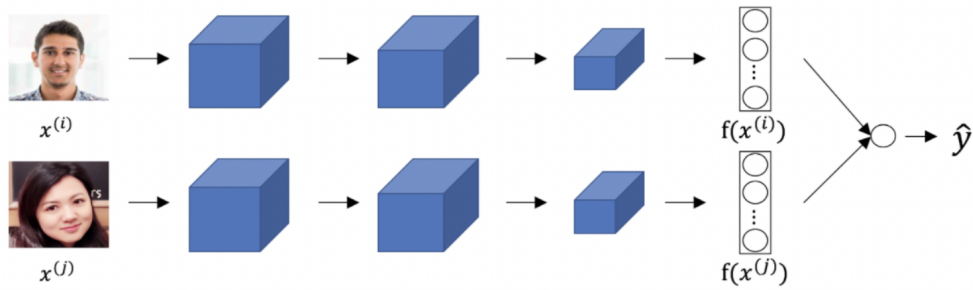


Correct

Great, you got all the right answers.

5. Consider the following Siamese network architecture:

1 / 1 point



Which of the following do you agree with the most?

- ☐ The two neural networks depicted in the image have the same architecture, but they might have different parameters.
- ☒ The upper and lower neural networks depicted have exactly the same parameters, but the outputs are computed independently for each image.
- ☐ Although we depict two neural networks and two images, the two images are combined in a single volume and pass through a single neural network.
- ☐ This depicts two "different" neural networks with different architectures, although we use the same drawing.

[Expand](#)

✓ **Correct**

Correct. Both neural networks share the same weights, and each image passes through the neural network in an independent manner.

6. Our intuition about the layers of a neural network tells us that units that respond more to complex features are more likely to be in deeper layers. True/False?

1 / 1 point

☐ False

☒ True

[Expand](#)

✓ **Correct**

Correct. Neurons that understand more complex shapes are more likely to be in deeper layers of a neural network.

7. Neural style transfer uses images Content C , Style S . The loss function used to generate image G is composed of which of the following: (Choose all that apply.)

1 / 1 point

☐ J_{corr} that compares C and S .

☒ J_{style} that compares S and G .

✓ **Correct**

Correct, in neural style transfer we are interested in the similarity between S and G , and the similarity between G and C .

☒ $J_{content}$ that compares C and G .

✓ **Correct**

Correct, in neural style transfer we are interested in the similarity between S and G , and the similarity between G and C .

☐ T that calculates the triplet loss between S , G , and C .

 Expand

 **Correct**

Great, you got all the right answers.

8. In the deeper layers of a ConvNet, each channel corresponds to a different feature detector. The style matrix $G^{[l]}$ measures the degree to which the activations of different feature detectors in layer l vary (or correlate) together with each other.

1 / 1 point

☒ True

☐ False

 Expand

 **Correct**

Yes, the style matrix $G^{[l]}$ can be seen as a matrix of cross-correlations between the different feature detectors.

9. In neural style transfer, which of the following better express the gradients used?

1 / 1 point

- ☐ $\frac{\partial J}{\partial W^{[l]}}$
- ☐ $\frac{\partial J}{\partial S}$
- ☒ $\frac{\partial J}{\partial G}$
- ☐ Neural style transfer doesn't use gradient descent since there are no trainable parameters.

[Expand](#)

✓ **Correct**

Correct, we use the gradient of the cost function over the value of the pixels of the generated image.

10. You are working with 3D data. The input "image" has size $64 \times 64 \times 64 \times 3$, if you apply a convolutional layer with 16 filters of size $4 \times 4 \times 4$, zero padding and stride 2. What is the size of the output volume?

1 / 1 point

- ☒ $31 \times 31 \times 31 \times 16$.
- ☐ $31 \times 31 \times 31 \times 3$.
- ☐ $64 \times 64 \times 64 \times 3$.
- ☐ $61 \times 61 \times 61 \times 14$.

[Expand](#)

✓ **Correct**

Correct, we can use the formula $\lfloor \frac{n^{[l-1]} - f + 2 \times p}{s} \rfloor + 1 = n^{[l]}$ to the three first dimensions.