



courserd

Convolutional Neural Networks

수 4주 차

Special Applications: Face Recognition & Neural Style Transfer

〈 이전 다음 〉

<u>:=</u>

- Face Recognition
- Neural Style Transfer
- Lecture Notes (Optional)
- Quiz
- 0

테스트: Special Applications: Face Recognition & Neural Style Transfer

- Programming Assignments
- References & Acknowledgments

테스트테스트 • 30 min30 minutes

♥ 과제 제출
기한년 9월 13일 오후 3:59 KST년 9월 13일 오후 3:59 KST 시도하기8 hours당 3회
다시 시도해주십시오
성적 받기 8
성적 100%
피드백 보기
최고 점수가 유지됩니다. 2
탐색 확인
이 페이지에서 나가시겠습니까?
이 페이지에 머물기 이 페이지에서 나가기
\leftarrow
Special Applications: Face Recognition & Neural Style Transfer
성적 평가 퀴즈 • 30 min
만료 년 9월 13일 오후 3:59 KST
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축하합니다! 통과하셨습니다! 통과 점수: 80% 이상
학습 계속하기
성적 100%
Special Applications: Face Recognition & Neural Style Transfer
최신 제출물 성적 15%
1. 질문 1
Face verification requires comparing a new picture against one person's face, whereas face recognition requires comparing a new picture against K person's faces.
1/12
True
$O_{\mathbb{O}}$
False
\checkmark
맞습니다
Correct.
2. 질문 2
Why do we learn a function $d(img1, img2)$ d(img1, img2) for face verification? (Select all that apply.)
0.5 / 1월 □
□□ We need to solve a one-shot learning problem.
Given how few images we have per person, we need to apply transfer learning.

This allows us to learn to predict a person's identity using a softmax output unit, where the number of classes equals the number of persons

in the database plus 1 (for the final "not in database" class).
×
이것은 선택하면 안 됩니다.
This is not the methodology used in face verification.
This allows us to learn to recognize a new person given just a single image of that person.
✓ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □
No
3. 질문 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.
$lackbox{0/12}{O}_{lackbox{O}}$
True
O_{\bigcirc}
False
×
틀립니다 답안을 선택하지 않았습니다.
4. 질문 4
Which of the following is a correct definition of the triplet loss? Consider that $\alpha > 0\alpha > 0$. (We encourage you to figure out the answer from first principles, rather than just refer to the lecture.)
0/1점 ○○
$max(f(A) - f(N) ^2 - f(A) - f(P) ^2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(P) 2 - \alpha, 0) \\ max(f(A) - f(N) 2 - f(A) - f(A) - f(N) 2 - f(A) - f(A) - f(N) 2 - f(A) - f(A) - f(N) 2 - f(A) - f(A) - f(N) 2 - f(A) - $
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O_{O}
$max(f(A) - f(P) ^2 - f(A) - f(N) ^2 + \alpha, 0) \max(f(A) - f(P) ^2 - f(A) - f(N) ^2 + \alpha, 0)$
X 틀립니다 답안을 선택하지 않았습니다. 5. 질문 5
Consider the following Siamese network architecture:
The upper and lower neural networks have different input images, but have exactly the same parameters.
${f O}_{f O}$
False
$O_{\mathbb{O}}$
True
X Eaulo
틀립니다 답안을 선택하지 않았습니다.
6. 질문 6

You train a ConvNet on a dataset with 100 different classes. You wonder if you can find a hidden unit which responds strongly to pictures of cats. (I.e., a neuron so that, of all the input/training images that strongly activate that neuron, the majority are cat pictures.) You are more likely to find this unit in layer 4 of the network than in layer 1.
True
\circ_{\circ}
False
×
틀립니다 답안을 선택하지 않았습니다. 7.
질문 7
Neural style transfer is trained as a supervised learning task in which the goal is to input two images (xx) , and train a network to output a new, synthesized image (yy) .
True
$O_{\mathcal{O}}$
False
★ 틀립니다 답안을 선택하지 않았습니다. 8. 질문 8
In the deeper layers of a ConvNet, each channel corresponds to a different feature detector. The style matrix $G^{[I]}G[1]$ measures the degree to which the activations of different feature detectors in layer I vary (or correlate) together with each other.
True
\circ_{\circ}
False
×
틀립니다 답안을 선택하지 않았습니다. 9.
질문 9
In neural style transfer, what is updated in each iteration of the optimization algorithm?
The pixel values of the generated image GG
O_O
The regularization parameters
O_{\bigcirc}
The pixel values of the content image CC
\circ_{\circ}
The neural network parameters
×
틀립니다 답안을 선택하지 않았습니다. 10. 질문 10
You are working with 3D data. You are building a network layer whose input volume has size 32x32x32x16 (this volume has 16 channels),



Undefined: This convolution step is impossible and cannot be performed because the dimensions specified don't match up.



30x30x30x16



30x30x30x32



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