

Final exam 2022

Deep Learning (Trường Đại học Bách khoa Hà Nội)



Scannen om te openen op Studocu

SCHOOL OF INFORMATION AND COMMUNICATION TECHNOLOGY – HUST –

Exam Key Code: 112

(The exam sheet consists of 8 pages)

INTRODUCTION TO DEEP LEARNING 2022-2023 Final Exam

Time: 120'

| | Clas | | tudent ID: |
|---------------------------------------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|----------------------------------------------------------------------|-------------------------------|
| SINGLE ANSW | VER MULTIPLE CHOICE QUESTION | S | |
| A Deep Lear B None of the C Data augr | Which of the following is FALSE abourning algorithms are more interpretable he above mentation can be done easily in Deep Lerning algorithms efficiently solve comparing algorithms. | e as compared to Machine L earning as compared to Mac | earning algorithms |
| • | 7 1 | 1 | |
| All of the | Which of the following is a type of neu above Neural Networks | Iral network? B Autoencoders D CNN (Convolutional) | Neural Network) |
| A During ba B We can us | Which of the following is FALSE about ackward propagation, we update the we see different activation functions in different gradient descent algorithms the above | eights using gradient descer ent layers | nt algorithm |
| Question 04. A It is linear C None of the | | B It either outputs 0 or 1 | l nold activation function |
| A Activation B Commonl C Activation | Which of the following is FALSE about functions help in achieving non-linear by used activation functions are step, sign functions help in reducing overfitting also called squashing functions as these | ity in deep neural networks moid, tanh, ReLU and softr problem | max |
| Question 06. (A) 0 to 1 | Output of step (threshold) activation f (B) -1 to 1 | unction ranges from: © Either 0 or 1 | D Either -1 or 1 |
| A None of the B Both are r | Which of the following is FALSE about he above non-linear activation functions both sigmoid and tanh is smooth, cont sigmoid ranges from -1 to 1 while outp | inuous and differentiable | |
| B Dropout i C None of the | Which of the following is FALSE about can be used in input, hidden and output is a hyper-parameter he above is implemented per layer in a network | - | |
| A Dropout i | Which of the following is FALSE abous a learnable parameter in the network ntroduces sparsity in the network | t Dropout? | |

© Dropout increases the accuracy and performance of the model

Dropout makes training process noisy

Question 10. Which of the following is TRUE about Dropout?

- A Dropout can be compared to boosting technique in machine learning
- B Dropout should be implemented only during training phase, not in testing phase
- C Dropout is computationally complex as compared to L1 and L2 regularization methods
- Dropout should be implemented during training phase as well as during testing phase

Question 11. Which of the following is TRUE about local and global minima?

- (A) Hyper-parameter tuning plays a vital role in avoiding global minima
- (B) All of the above
- C Ideally, SGD should reach till local minima and should not stuck in global minima
- D Sometimes local minimas are as good as global minimas

Question 12. Which of the following is a way to avoid local minima?

(A) All of the above

B Use momentum and adaptive learning

C Increase the learning rate

D Add some noise while updating weights

Question 13. Which of the following SGD variants is **NOT** based on adaptive learning?

- (A) Adagrad
- **B** Nesterov
- C AdaDelta
- (D) RMSprop

Question 14. Which of the following is TRUE about Weight Initialization?

- (A) If weights are too high, it may lead to vanishing gradient
- B If weights are too low, it may lead to exploding gradient
- (C) All of the above
- D Model may never converge due to wrong weight initialization

Question 15. Which of the following is **TRUE** about Momentum?

- All of the above
- **B** It helps in faster convergence
- C It helps in accelerating SGD in a relevant direction
- D It helps SGD in avoiding local minima

Question 16. Which of the following is FALSE about Pooling Layer in CNN?

- (A) It does down-sampling of an image which reduces dimensions by retaining vital information
- (B) It does feature extraction and detects components of the image like edges, corners etc.
- C Output of convolutional layer acts as an input to the pooling layer
- D Pooling layer must be added after each convolutional layer

Question 17. Which of the following is a valid reason for not using fully connected networks for image recognition?

- (A) It creates a lot more parameters for computation as compared to CNN
- B CNN is far efficient in terms of performance and accuracy for image recognition
- (C) All of the above
- D It may overfit easily as compared to CNN

Question 18. Which of the following is **FALSE** about Padding in CNN?

- (A) There are two types of padding: Zero Padding and Valid Padding (no padding)
- B In zero padding, we pad the image with zeros so that we do not lose any edge information
- \bigcirc Padding is used to prevent the loss of information about edges and corners during convolution
- D There is no reduction in dimension when we use valid padding

Question 19. Which of the following is **FALSE** about Kernels in CNN?

- (A) Kernels extract simple features in initial layers and complex features in deeper layers
- B Kernels can be used in convolutional as well as in pooling layers
- C Kernels keep sliding over an image to extract different components or patterns of an image
- (D) None of the above

| \simeq | of convolution and size of ke in a convoluti | ernels in a convolutional lay onal layer (zero or valid pa | er | |
|-----------------------------------------------------------------------------------------------------|----------------------------------------------------|-------------------------------------------------------------------------------------------------------------------------------|-------------------------------------------------------------------------|---------------------------------------------|
| A None of t B LSTM is a C LSTM ena | he above an extension : ables RNN to | te following is FALSE about for RNA which extends its to learn long-term dependen- ding gradients issue in RN | memory cies | |
| Question 22. A Algorithm C Image con | nic trading | e following is NOT an appl | ication of RNN? B Image captioning D Understanding DNA | sequence |
| A Anomaly | | e following is NOT an appl n | ication of RNN? (B) Weather prediction (D) Time series prediction | n |
| Question 24. A 3 | How many | parts can the GAN be divid | led into? | D 4 |
| | nerated using | e following words is used to g probabilistic models. B Discriminator | o familiarize yourself with a | generative models and to explai D Networks |
| Question 26. A Naive Bay C GAN mo | yes | e following is not an examp | ole of a generative model? B Discriminator models D PixelRNN/PixelCNN | |
| Question 27. A None of t | | standard form of YOLO? B You Once Look Only | © YOu Look Once | D You Only Look Once |
| Question 28. (a) Model da (b) Hypothes (c) Feature d (d) Hypothes (a) and (c) (b) (c) and (d) | tabase, sizer, etector, sis verifier | (B) (a) and (b) (E) All (a), (b), (c) | © No | ystem one of the above |
| Question 29. (a) Biometric | dentificatio | | | |
| (b) Human a (A) (a) | na computer | B None of the above | © Both (a) and (b) | (b) |
| Question 30. A Count veo B All of the C N-gram v | ctorization co above vectorization | ne following is TRUE about onsiders count and weightage considers context of the woonsiders both count and co | types of Vectorization in Nge of each word in a text | LP? ue of N |

| Question 31. Which of the following in A Google Assistant B Chatbook | * * | f NLP? Google Translate | D All of the above |
|---------------------------------------------------------------------------------------------------------------------------------------------------------------------------|----------------------|----------------------------|------------------------------------------------------------------------|
| Question 32. Which of the following is A NLP can be used for spam filtering B All of the above C We must take care of Syntax, Seman D Preprocessing tasks include Tokeni | , sentiment analys | sis and machine translat | |
| Question 33. Which of the following t | techniques can be | used to reduce model o | verfitting? |
| (a) Data augmentation | | | |
| (b) Dropout | | | |
| (c) Batch Normalization | | | |
| (d) Using Adam instead of SGD | | | |
| A All the choices (a),(b),(c) and (d) are (a), (b) and (c) (E) (b) | e correct (I | (a) and (b) (a) | |
| Question 34. Which of the following i | is true about drop | out? | |
| (a) Dropout leads to sparsity in the tra | ined weights | | |
| (b) At test time, dropout is applied wit | th inverted keep p | robability | |
| (c) The larger the keep probability of a | ı layer, the stronge | er the regularization of t | he weights in that layer |
| (c) | (I | All the choices (a),(b) | and (c) are correct |
| (C) (b) (E) None of the above | Œ | (a) | |
| | | | |
| Question 35. You are training a Generator might be showing mod lem? | | O | mages of reptiles. But, you think could be indicators of this prob- |
| (a) The generator is only producing im | nages of komodo d | lragons | |
| (b) The generator loss is oscillating. | | | |
| (c) The generator loss remains low wh | ereas the discrimi | nator loss is high | |
| (d) The discriminator has high accurac | y on real images l | out low accuracy on fak | e ones |
| (a) and (b) | | 3 (c) | |
| C All the choices (a),(b),(c) and (d) are (d) | e correct (| (c) and (d) | |
| Question 36. Which one of the follow | | | |
| A converting to lowercase | | Stemming and Lemm | natization |
| © removal of stop words E removing punctuations | Œ | Sentiment analysis | |
| _ | ~ | | ı think your Generator might be roblem? |
| (a) The generator is only producing im | nages of komodo d | lragons. | |
| (b) The generator loss is not stable. | | | |

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(c) The generator loss remains low whereas the discriminator loss is high.

(d) The discriminator has high accuracy on real images but low accuracy on fake ones.

| (a) and (b | (a),(b) and (c) | (a) | (b) | E All of the above |
|----------------|---------------------------------|------------------|------------------------|---------------------------|
| Question 38. | Which of the following prop | ositions are T | RUE about a CONV layer | ? (Check all that apply) |
| (a) The num | ber of weights depends on th | e depth of the | input volume. | |
| (b) The num | ber of biases is equal to the n | umber of filter | S. | |
| (c) The total | number of parameters deper | nds on the stric | de. | |
| (d) The total | number of parameters deper | nds on the pad | ding. | |
| (a) (b) and (d | B (0 | c) and (d) | (C) (a) a | nd (b) |

(E) (a) and (d)

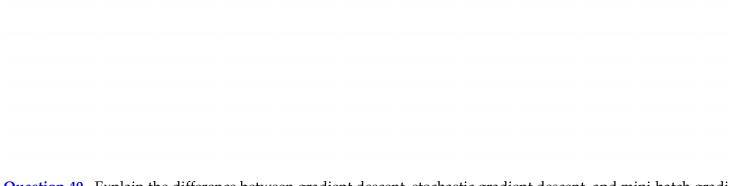
II. SHORT ANSWER QUESTIONS

D All the choices are correct

Question 39. You have been tasked to build a classifier that takes in an image of a movie poster and classifies it into one of four genres: comedy, horror, action, and romance. You have been provided with a large dataset of movie posters where each movie poster corresponds to a move with exactly one of these genres.

Your model now has 100% accuracy on the training set, and 96% accuracy on the validation set! You now decide to expand the model to posters of movies belonging to multiple genres. Now, each poster can have multiple genres associated with it; for example, the poster of a movie like "Lật mặt: Nhà có khách" falls under both "comedy" and "action". Propose a way to label new posters, where each example can simultaneously belong to multiple classes?

To avoid extra work, you decide to retrain a new model with the same architecture (softmax output activation with cross-entropy loss). Explain why this is problematic?



Question 40. Explain the difference between gradient descent, stochastic gradient descent, and mini-batch gradient descent

Question 41. Consider the convolutional neural network defined by the layers in the left column below.

Fill in the shape of the output volume and the number of parameters at each layer. You can write the activation shapes in the format (H, W, C), where H, W, C are the height, width and channel dimensions, respectively. Unless specified, assume padding 1, stride 1 where appropriate.

Notation:



- **CONV**x-N denotes a convolutional layer with N filters with height and width equal to x.
- **POOL-**n denotes a $n \times n$ max-pooling layer with stride of n and 0 padding.
- FLATTEN flattens its inputs, identical to torch.nn.flatten / tf.layers.flatten
- FC-N denotes a fully-connected layer with N neurons

| Layer | Activation Volume Dimension | Number of Parameters |
|------------|-----------------------------|----------------------|
| Input | $32 \times 32 \times 3$ | 0 |
| CONV3-8 | | |
| Leaky ReLU | | |
| POOL-2 | | |
| BATCHNORM | | |
| CONV3-16 | | |
| Leaky ReLU | | |
| POOL-2 | | |
| FLATTEN | | |
| FC-10 | | |

| Question 42. | . Give a method to fight vanishing gradient in fully-connected neural networks | s. Assume we are using |
|--------------|--------------------------------------------------------------------------------|------------------------|
| a network wi | ith Sigmoid activations trained using SGD. | |

Question 43. How do we train the deep network?

| Question 44. Explain the difference between the sigmoid and tanh activation function. | |
|---------------------------------------------------------------------------------------|--|
| | |
| | |
| Question 45. What is the Jacobian Matrix? | |
| | |
| Question 46. Explain Generative Adversarial Network. | |
| | |
| | |
| Question 47. How LSTM differ from the RNN? | |
| | |
| | |

| Question 48. What is the difference between the same padding and valid padding? |
|-----------------------------------------------------------------------------------------------------|
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| Question 49. What is IoU? |
| |
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| |
| |
| Overting 50. In NLD becomes described directed by investigating the distance between 2 televis 2 |
| Question 50. In NLP, how word embedding techniques help to establish the distance between 2 tokens? |
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