# Final Review

CS 4277: Deep Learning

## 8 Measuring Performance

1. \* Describe the three principle sources of errors that lead to poor generalization in machine learning and how they can be reduced. (8.2-8.3)

2. \* Describe the bias-variance tradeoff. (8.3.3)

3. \* Describe the double-descent phenomenon in deep neural networks. (8.4)

4. What is the typical approach to choosing hyperparameters? (8.5)

# 9 Regularization

5. \* What is the goal of regularization?

6. \* What is the standard approach to explicit regularization? (9.1)

7. Describe L2 regularization. (9.1.2)

8. How is implicit regularization accomplished by SGD? (9.2.2)

9. List 3 heuristic methods of implicit regularization. (9.3)

10. \* Consider a model where the prior distribution over the parameters is a normal distribution with mean zero and variance  $\sigma_{\phi}^2$  so that

$$Pr(\phi) = \prod_{i=1}^{J} \operatorname{Norm}_{\phi_j}(0, \sigma_{\phi}^2)$$

where j indexes the model parameters. When we apply a prior, we maximize  $\prod_{i=1}^{I} Pr(\mathbf{y}_i | \mathbf{x}_i, \phi) Pr(\phi)$ . The associated loss function of this model is equivalent to which regularization technique?

## 10 Convolutional Networks

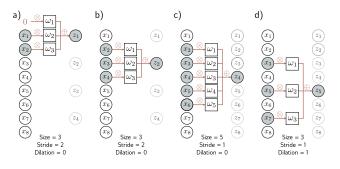
11. What is invariance? (10.1)

12. What is equivariance? (10.1)

13. \* What properties of images make convolutional neural networks well-suited to them?

14. \* What is the motivation for convolutional layers in a neural network?

15. \* Write out the equation for the 1D dilated convolution with a kernel size of three and a dilation rate of two, as pictured in Figure 10.3d (reproduced below).



- 16. \* T/F The convolution operation is equivariant to translation. in
- 17. T/F The convolution operation is invariant to translation. in
- 18. \* Consider a 1D convolutional layer computed using a kernel size of three and has four channels. How many weights and biases are needed for this convolutional layer?

- 19. Describe three methods of downsampling. (10.4.1)
- 20. Describe four methods of upsampling. (10.4.2)

#### 11 Residual Networks

21. \* Describe the shattered gradients problem in deep networks. (11.1.1)

22. \* What is a residual, a.k.a., skip, connection? (11.2)

23. What is the typical order of operations in a residual block? (11.2 - 11.2.1)

24. \* Describe the problem of exploding gradients in residual networks. (11.3)

25. \* What is batch normalization and why is it used? (11.4)

26. What is the chief drawback of batch normalization and what are its advantages? (11.4.1)

## 12 Transformers

27. \* What are the two primary design goals acheived by dot-product self-attention in a language model? (12.2)

28. \* Why is positional encoding used in language models? (12.3.1)

29. What are the typical internal sub-layers of a transformer layer? (12.4)

30. \* What is Tokenization? (12.5.1)

31. \* Embeddings (12.5.2)

32. \* Encoders and decoders. (12.6)

33. \* Pre-training (12.6.1)

34. \* Fine-tuning (12.6.2)

35. \* Auto-regressive language modeling (12.7.1)

36. Few-shot learning (12.7.4)

#### 13 Graph Neural Networks

37. Graph-level tasks (13.3.1)

38. Node-level tasks (13.3.1)

39. Edge-prediction tasks (13.3.1)

40. \* What is the defining feature of graph convolutional neural networks? (13.4)

41. \* What is meant by *relational inductive bias* in graph convolutional networks? (13.4)

42. How is parameter sharing accomplished in graph convolutional networks? (13.4.2)

# 19 Deep Reinforcement Learning

43. What is meant by *temporal credit assignment*?

44. What is the *Markov property* with respect to states  $s_1, s_2, \ldots, s_T$  where  $t \in T$  are time steps?

45. \* What is the primary advantage of deep reinforcement learning over tabular reinforcement learning?