

## Quiz: Chapter 3 (Section 3.4)

Printed Name:

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### Quiz rules:

1. You MAY use:
  - (a) any notes (handwritten, printed, or electronic),
  - (b) any computer programs (including websites like WolframAlpha or ChatGPT), and
  - (c) any additional scratch paper.
2. You MAY NOT communicate with another student.
3. If you finish the quiz early, stay seated. I will collect all the quizzes at the same time.

**Problem 1.** For each statement below, circle **True** if the statement is known to be true, **False** if the statement is known to be false, and **Open** if the statement is not known to be either true or false. You will receive +2 point for each correct answer, **-1 points for each incorrect answer**, and 0 points for each blank answer.

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|---------|-------|------|--|
| 1. True | False | Open | Let $\mathcal{H}_{\text{perceptron}}$ be the perceptron hypothesis class and let $\mathcal{H}_{\text{stump}}$ be the decision stump hypothesis class. If $k$ is a breakpoint for $\mathcal{H}_{\text{stump}}$ , then $k$ is guaranteed to be a breakpoint for $\mathcal{H}_{\text{perceptron}}$ as well.   |
| 2. True | False | Open | Let $\mathcal{H}_{\Phi_2}$ be the perceptron hypothesis class with the 2nd degree polynomial feature embedding, and $\mathcal{H}_{\Phi_4}$ be perceptron hypothesis class with the 4th degree polynomial feature embedding. Let $g_{\Phi_2} \in \mathcal{H}_{\Phi_2}$ and $g_{\Phi_4} \in \mathcal{H}_{\Phi_4}$ be the empirical risk minimizers. It is guaranteed that $E_{\text{in}}(g_{\Phi_2}) \leq E_{\text{in}}(g_{\Phi_4})$ .   |
| 3. True | False | Open | Let $\mathcal{H}_{\Phi_2}$ be the perceptron hypothesis class with the PCA feature embedding with output dimension 2, and $\mathcal{H}_{\Phi_4}$ be perceptron hypothesis class with the PCA feature embedding with output dimension 4. Let $g_{\Phi_2} \in \mathcal{H}_{\Phi_2}$ and $g_{\Phi_4} \in \mathcal{H}_{\Phi_4}$ be the empirical risk minimizers. VC theory predicts that with high probability, the generalization error of $g_{\Phi_2}$ will be less than the generalization error of $g_{\Phi_4}$ . |
| 4. True | False | Open | VC theory predicts that when using the $\mathcal{H}_{\text{axis2}}$ hypothesis class, centering your data points results in a better generalization error with high probability.   |