

# Machine Learning

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Quiz & Notes on final exam

# Quiz, 1

- (T or F) A classifier trained on less training data is less likely to overfit.
- Answer: **F**

# Quiz, 3

- (T or F) If today I want to predict the **probability** that a student sleep more than 8 hours on average (SA) given the Course loading (C), I will choose to use linear regression over logistic regression.
- Answer: **F**

# Quiz, 8

## Underfitting

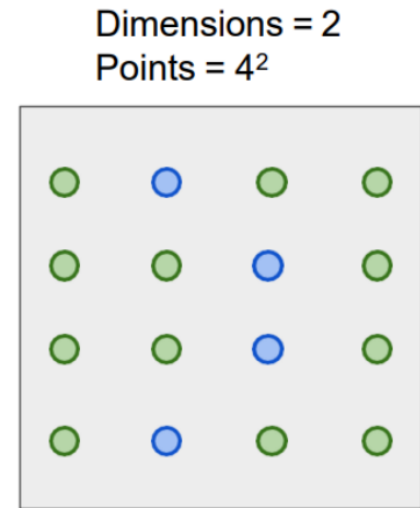
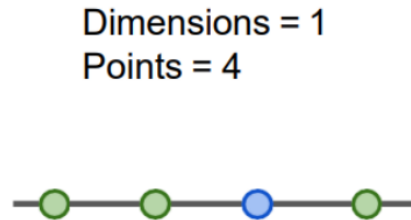
- You train a linear classifier on 10,000 training points and discover that the training accuracy is only 67%. Which of the following, done in isolation, has a good chance of improving your training accuracy?
  - A. Add novel features
  - B. Train on more data
  - C. Use linear regression
  - D. Train on less data
- Answer: **AD**

# Quiz, 9

## Overfitting

- You train a classifier on 10,000 training points and obtain a training accuracy of 99%. However, when you submit to Kaggle (A testing platform), your accuracy is only 67%. Which of the following, done in isolation, has a good chance of improving your performance on Kaggle?
  - A. Set your regularization value ( $\lambda$ ) to 0
  - B. Train on more data
  - C. Use validation to tune your hyperparameters
  - D. Train on less data
- Answer: **BC**

# Quiz, 10



- You are performing classification in  $d$  dimensions (where  $d$  is very large) using a  $k$ -nearest neighbor classifier. You have  $N$  training samples and get an acceptable performance. Your boss hands you a similar classification task but in  $2d$  dimensions. How many training samples will you need to get a comparable performance? Make an educated guess.

A.  $2N$

B.  $e^N$

C.  $N^2$

D.  $d^{0.5}$

E.  $e^d$

F.  $N^{0.5}$

G.  $\log(N)$

H.  $d^2$

I.  $2d$

J.  $\log(d)$

- Answer: **C** (KNN P37 Curse of dimensionality)

# Quiz, 11

- Consider the following joint distribution on  $X$  and  $Y$ , where  $X \in \{-1, 0, 1\}$  and  $Y \in \{0, 1\}$ :  $p(X = -1, Y = 0) = 0.05$ ,  $p(X = -1, Y = 1) = 0.05$ ,  $p(X = 0, Y = 0) = 0.1$ ,  $p(X = 0, Y = 1) = 0.1$ ,  $p(X = 1, Y = 0) = 0.3$ ,  $p(X = 1, Y = 1) = 0.4$ . You learn that  $X \geq 0$ . What is the largest probability of being correct you can achieve when predicting  $Y$  in this case?

- A. 5/9                      B. 1                      C. 2/3                      D. 1/4                      E. 1/2  
F. 1/7                      G. 1/3                      H. 6/7                      I. 4/7                      J. 3/7

- Answer: **5/9**

$X \backslash Y$	0	1
-1	0.05	0.05
0	0.1	0.1
1	0.3	0.4

$$P(Y = 1 | X \geq 0) = \frac{P(X \geq 0, Y = 1)}{P(X \geq 0)}$$

# Quiz, 13

- Consider a linear regression problem with  $N$  samples  $\{(x_n, y_n)\} (n = 1 \dots N)$ , where each input  $x_n$  is a  $D$ -dimensional vector  $\{-1, +1\}^D$ , and all output values are  $y_i \in \mathbb{R}$ . Which of the following statements is correct?
  - A. Linear regression always “works” very well for  $N \ll D$
  - B. A linear regressor works very well if the data is linearly separable.
  - C. Linear regression always “works” very well for  $D \ll N$
  - D. None of the above.
- Answer: **D**



# Quiz, 14A 16D

- (14) Which of the following are true about decision trees?  
A. They can be used only for classification.
- (16) Which of the following are true for the k-nearest neighbor (k-NN) algorithm?  
D. k-nearest-neighbors cannot be used for regression.
- Both choices are **wrong**.

# Quiz, 21

- Consider the problem of binary classification using the Naive Bayes classifier. You are given two dimensional features ( $X_1, X_2$ ) and the categorical class conditional distributions in the tables below. The entries in the tables correspond to  $P(X_1 = x_1|C_i)$  and  $P(X_2 = x_2|C_i)$  respectively.

The two classes are equally likely.

Class \ $X_1 =$	$C_1$	$C_2$
-1	0.2	0.3
0	0.4	0.6
1	0.4	0.1

Class \ $X_2 =$	$C_1$	$C_2$
-1	0.4	0.1
0	0.5	0.3
1	0.1	0.6

- Given a data point  $(-1, 1)$ , calculate the following posterior probabilities:
  - $P(C_1|X_1=-1, X_2=1)$ ,  $P(C_2|X_1=-1, X_2=1)$

# Quiz, 21

$X_1 =$ \ Class	$C_1$	$C_2$
-1	0.2	0.3
0	0.4	0.6
1	0.4	0.1

$X_2 =$ \ Class	$C_1$	$C_2$
-1	0.4	0.1
0	0.5	0.3
1	0.1	0.6

- Given a data point  $(-1, 1)$ , calculate the following posterior probabilities:
  - $P(C_1|X_1=-1, X_2=1)$ ,  $P(C_2|X_1=-1, X_2=1)$

$$\begin{aligned}
 & \frac{P(X_1 = -1, X_2 = 1|C_1)P(C_1)}{P(X_1 = -1, X_2 = 1)} \\
 &= \frac{P(X_1 = -1|C_1)P(X_2 = 1|C_1)P(C_1)}{P(X_1 = -1|C_1)P(X_2 = 1|C_1)P(C_1) + P(X_1 = -1|C_2)P(X_2 = 1|C_2)P(C_2)} \\
 &= 0.1
 \end{aligned}$$

# Notes on final exam

- 不定项选择题：不建议选择没有把握的选项
- 注意题目中的信息： true, false, wrong, correct, incorrect
- 不能使用计算器，请注意计算准确度