Q1-1: K-NN algorithms can be used for:

- 1. Only classification
- 2. Only regression
- 3. Both

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- 2. Euclidean Distance
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Q1-3: Consider binary classification in 2D where the intended label of a point x = (x1, x2) is positive if x1>x2 and negative otherwise. Let the training set be all points of the form x = [4a, 3b] where a,b are integers. Each training item has the correct label that follows the rule above. With a 1NN classifier (Euclidean distance), which ones of the following points are labeled positive? Multiple answers.

- 1. [5.52, 2.41]
- 2. [8.47, 5.84]
- 3. [7 , 8.17]
- 4. [6.7 , 8.88]

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Nearest neighbors are
[4,3] => positive
[8,6] => positive
[8,9] => negative
[8,9] => negative
Individually.

Q2-1: Consider a variant of II Nim game where there are 2 piles, with 1 and 3 sticks, respectively. Each time one player takes some stick(s) from only one pile (can take 1 or 2 or 3 sticks). What's the game theoretic value of the initial state?

A. 1B. 0C. -1

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The first player could always take 3 sticks from the pile with 3 sticks, which guarantees the game value +1. No need to check the other branches.

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Q2-2: We know that the game theoretic value of the initial state is -1 in II Nim game where there are 2 piles, each with 2 sticks. Now consider a variant of II Nim game where there are 2 piles, with 2 and 3 sticks, respectively. Each time one player takes some stick(s) from only one pile (can take 1 or 2 or 3 sticks). What's the game theoretic value of the initial state?

A. 1

B. 0

C. -1



The first player could always take 1 stick from the pile with 3 sticks and make it a (2,2) state, in which case the second player will always lose.

Q2-3: Consider a variant of II Nim game with 2 piles, both with 3 sticks. Each time one player takes some stick(s) from only one pile (can take 1 or 2 or 3 sticks). What's the game theoretic value of the initial state? (Hint: use the answers of the last two questions)

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Q2-3: Consider a variant of II Nim game with 2 piles, both with 3 sticks. Each time one player takes some stick(s) from only one pile (can take 1 or 2 or 3 sticks). What's the game theoretic value of the initial state? (Hint: use the answers of the last two questions)

_{iii} Min

; Max

Min

A. 1 B. 0

C. -1

Max first moves, leading to 3 possible states.
(2,3): Min will win according to Q2-2, i.e., Min can take 1 stick from the pile with 3 sticks and make it (2,2).
(1,3): Min will win according to Q2-1, i.e., Min can take all the second pile, and make it (1,-).
(0,3): Min can take 2 sticks and will win.

i,iii <mark>Min</mark>

ii,iii ^{Min}

- A. {1}, {2,4,5,7.25}
- B. {1,2}, {4, 5, 7.25}
- C. {1,2,4}, {5, 7.25}
- D. {1,2,4,5}, {7.25}



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$$d(C_1, \{4\}) = \frac{3+2}{2} = 2.5,$$

$$d(\{4\}, \{5\}) = 1$$





- A. {1}, {2,4,5,7.25} B. {1,2}, {4, 5, 7.25}
- C. {1,2,4}, {5, 7.25}
- D. {1,2,4,5}, {7.25}



- Q3-2: Assume, you want to cluster 7 points in 2-dim into 3 clusters using K-Means clustering algorithm. After the first iteration clusters, C1, C2, C3 has the following points: C1: $\{(1,1), (3,3), (5,5)\}$ C2: $\{(0,4), (-2,4)\}$ C3: $\{(7,7), (9,9)\}$ What will be the cluster centroids if you want to proceed for the second iteration?
- A. C1: (3,3), C2: (-1,4), C3: (8,8)
- B. C1: (3,3), C2: (0,4), C3: (8,8)
- C. C1: (3,3), C2: (-1,4), C3: (7,8)
- D. C1: (2,2), C2: (-1,4), C3: (7,8)

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 C. C1: (3,3), C2: (-1,4), C3: (7,8)

D. C1: (2,2), C2: (-1,4), C3: (7,8)

Compute the average of the data points in each cluster

Q3-3: Which are true about linear regression?

- 1. When $\lambda \to +\infty$, ridge regression reduces to OLS.
- 2. The regression function must be linear in the original input features.
- 3. Gradient descent can be used to solve OLS.

A. 1,2

- B. 1,3
- C. 2,3
- D. None of the above

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- B. 1,3
- C. 2,3
- D. None of the above

No. when λ → 0, ridge regression reduces to OLS
 No. It only needs to be linear in the parameter
 Yes. Gradient descent is a general method for optimization.

Q3-4: Which are true about logistic regression?

- 1. When $\theta^T x_i = 0$, the model will predict label +1 with probability close to 1
- 2. There is ground-truth θ^* that can achieve 0 classification error on the training set
- 3. Gradient descent can be used to solve the regularized logistic regression problem
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- B. 1,3
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- D. None of the above

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- A. 1,2
- B. 1,3
- C. 2,3
- D. None of the above

- 1: No. predicts ½ for +1, ½ for -1.
- 2: No. Not 0 classification error. There is randomness in the labels.
- 3: Yes. Gradient descent is a general method for optimization.

Q3-5: Suppose we have $\theta = [0.5, 0.6, 1]$ for logistic regression. What label will the model predict for x = [1, -5, 2]?

- A. -1
- B. +1
- C. Equal probabilities for -1 and +1
- D. None of the above

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- B. +1
- C. Equal probabilities for -1 and +1
- D. None of the above

The $\theta^T x = -0.5$, so p[y=+1|x]<0.5, so we will predict -1