

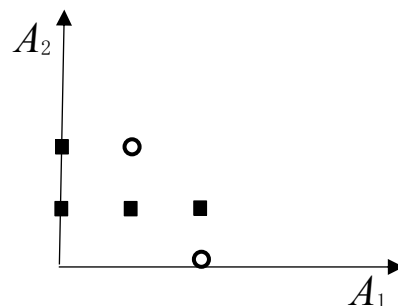
- Consider the following training set, in which each example has two tertiary attributes (0, 1, or 2) and one of two possible classes (X or Y).

Example	A ₁	A ₂	Class
1	0	1	X
2	2	1	X
3	1	1	X
4	0	2	X
5	1	2	Y
6	2	0	Y

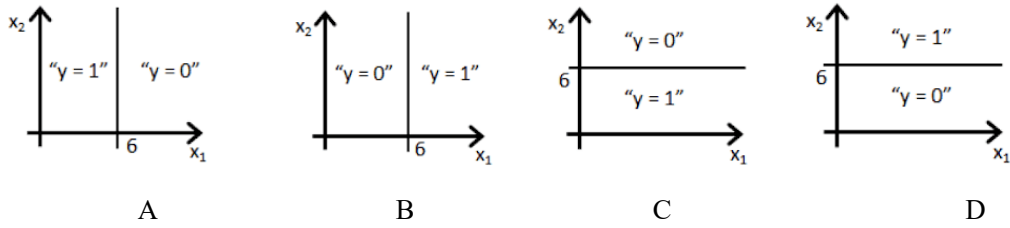
- What feature would be chosen for the split at the root of a decision tree using the information gain criterion? Show the details. (Note: we split attributes at each value of the attributes, for example, A₁=0, A₁=1, A₁=2)
- What would the Naïve Bayes algorithm predict for the class of the following new example? Show the details of the solution.

Example	A ₁	A ₂	Class
7	2	2	?

- Draw the decision boundaries for the nearest neighbor algorithm assuming that we are using standard Euclidean distance to compute the nearest neighbors.



- Which of these classifiers will be the least likely to classify the following data points correctly? Please explain the reason.
 - ID3.
 - Naïve Bayes
 - Logistic Regression
 - KNN
- You have trained a logistic classifier $y = \text{sigmoid}(w_0 + w_1x_1 + w_2x_2)$. Suppose $w_0=6$, $w_1=-1$, and $w_2=0$. Which of the following figures represents the decision boundary found by your classifier?



3. Suppose we are given a dataset $D=\{(x^{(1)},r^{(1)}),\dots,(x^{(N)},r^{(N)})\}$ and aim to learn some patterns using the following algorithms. Match the update rule for each algorithm.

Algorithms:

A: SGD for Logistic Regression $y = \text{sigmoid}(\mathbf{w}^T \mathbf{x})$
B: Least Mean Squares for Linear Regression $y = \mathbf{w}^T \mathbf{x}$
C: Perceptron $y = \text{sign}(\mathbf{w}^T \mathbf{x})$ (where $\text{sign}(a)=1$ if $a>0$ else -1)

Update Rules:

1. $\mathbf{w}_t \leftarrow \mathbf{w}_t + (\mathbf{w}_t^T \mathbf{x}^{(l)} - r^{(l)})$
2. $\mathbf{w}_t \leftarrow \mathbf{w}_t + \frac{1}{1 + \exp \eta(\mathbf{w}_t^T \mathbf{x}^{(l)} - r^{(l)})}$
3. $\mathbf{w}_t \leftarrow \mathbf{w}_t + \eta(y^{(l)} - r^{(l)})\mathbf{x}_i^{(l)}$