

# CSP 571 Data Preparation and Analysis

## Quiz - 4

### Question 1

In a **support vector machine**, in order to allow for *non-linear decision boundaries*, the expansion of feature space with non-linear functions of the predictors is accomplished via a

- ☐ a. margin
- ☐ b. hyperplane
- ☐ c. vector
- ☒ d. kernel

### Question 2

The process of building a tree via recursive binary splitting of the predictor space in a top-down approach involves making a

- ☐ a. sub-optimal choice.
- ☒ b. greedy choice.
- ☐ c. gradient optimal choice.
- ☐ d. globally optimal choice.

### Question 3

In a **support vector classifier**, a *training* observation  $x_i$  that is on the correct side of the separating hyperplane but on the incorrect side of the margin will have a slack variable value of

- ☐ a.  $1 < \varepsilon_i$
- ☐ b.  $0 \leq \varepsilon_i < 1$
- ☒ c.  $0 < \varepsilon_i \leq 1$
- ☐ d.  $\varepsilon_i \leq 0$

### Question 4

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Given a **hyperplane** which divides a  $p$  dimensional space into two halves, from which we can perform classification of an arbitrary point  $x$ , the hyperplane itself is an *affine subspace* of the below number of dimensions

- ☒ a.  $p - 1$
- ☐ b.  $p + 1$
- ☐ c.  $p$
- ☐ d. 0

### Question 5

The representation of a **support vector machine** as  $f(x) = \beta_0 + \sum_{i=1}^n \alpha_i K(x, x_i)$  will result in what value for  $\alpha_i$  for *training* observations which are not support vectors

- ☒ a. 0
- ☐ b.  $\infty$
- ☐ c. -1
- ☐ d. 1

### Question 6

Random forests improve upon bagging results by de-correlating accross bagged trees, via reduction of the number of

- ☒ a. predictors considered at each split.
- ☐ b. branches considered at each split.
- ☐ c. observations considered at each split.
- ☐ d. nodes considered at each split.

### Question 7

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Improving regression tree results via bagging involves  $B$  regression trees being induced from bootstrap samples, and final predictions being the

- ☐ a. minimum of each tree's predicted value.
- ☐ b. sum of each tree's predicted value.
- ☒ c. average of each tree's predicted value.
- ☐ d. maximum of each tree's predicted value.

### Question 8

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Decision tree methods split the predictor space into a series of non-overlapping regions  $R_j$ , each of which maps to a

- ☐ a. internal node.
- ☐ b. chlorophyll node.
- ☒ c. leaf/terminal node.
- ☐ d. root node.

### Question 9

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When inducing a classification tree, a node that is more pure will have both Gini and Entropy values that are

- ☐ a. infinite compared to other nodes.
- ☒ b. lower than compared to other nodes.
- ☐ c. negative compared to other nodes.
- ☐ d. higher than compared to other nodes.

### Question 10

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In a **maximal margin classifier**, *training* data points which are directly related to the optimal separating hyperplane and would result in a different/changed hyperplane if they are altered are referred to as

- ☐ a. optimal vectors
- ☐ b. margin vectors
- ☐ c. kernel vectors
- ☒ d. support vectors