

Preventing Overfitting in Decision Trees

11 试题

1
point

1.
(True/False) When learning decision trees, smaller depth USUALLY translates to lower training error.

- ☐ True
- ☒ False

1
point

2.
(True/False) If no two data points have the same input values, we can always learn a decision tree that achieves 0 training error.

- ☒ True
- ☐ False

1
point

3.
(True/False) If decision tree T1 has lower training error than decision tree T2, then T1 will always have better test error than T2.

- ☐ True

☒ False

1
point

4.

Which of the following is true for decision trees?

- ☐ Model complexity increases with size of the data.
 - ☒ Model complexity increases with depth.
 - ☐ None of the above
-

1
point

5.

Pruning and early stopping in decision trees is used to

- ☒ combat overfitting
 - ☐ improve training error
 - ☐ None of the above
-

1
point

6.

Which of the following is NOT an early stopping method?

- ☐ Stop when the tree hits a certain depth
- ☐ Stop when node has too few data points (minimum node "size")
- ☒ Stop when every possible split results in the same amount of error reduction
- ☐ Stop when best split results in too small of an error

1
point

7.

Consider decision tree T1 learned with minimum node size parameter = 1000. Now consider decision tree T2 trained on the same dataset and parameters, except that the minimum node size parameter is now 100. Which of the following is always true?

- ☒ The depth of T2 \geq the depth of T1
- ☒ The number of nodes in T2 \geq the number of nodes in T1
- ☐ The test error of T2 \leq the test error of T1
- ☒ The training error of T2 \leq the training error of T1

1
point

8.

Questions 8 to 11 refer to the following common scenario:

Imagine we are training a decision tree, and we are at a node. Each data point is (x_1, x_2, y) , where x_1, x_2 are features, and y is the label. The data at this node is:

x_1	x_2	y
0	1	+1
1	0	+1
0	1	+1
1	1	-1

What is the classification error at this node (assuming a majority class classifier)?

0.25

1
point

Refer to the scenario presented in Question 8.

9. If we split on x_1 , what is the classification error?

0.25

1
point

Refer to the scenario presented in Question 8.

10. If we split on x_2 , what is the classification error?

0.25

1
point

11.

Refer to the scenario presented in Question 8.

If our parameter for minimum gain in error reduction is 0.1, do we split or stop early?

- ☐ Split
- ☒ Stop early



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