

Boosting a decision stump

5 试题

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1.

Recall that the **classification error for unweighted data** is defined as follows:

$$\text{classification error} = \frac{\# \text{ mistakes}}{\# \text{ all data points}}$$

Meanwhile, the **weight of mistakes for weighted data** is given by

$$\text{WM}(\alpha, \hat{y}) = \sum_{i=1}^n \alpha_i \times 1[y_i \neq \hat{y}_i].$$

If we set the weights $\alpha=1$ for all data points, how is the weight of mistakes **WM**(α, \hat{y}) related to the classification error?

- ☐ **WM**(α, \hat{y}) = [classification error]
- ☐ **WM**(α, \hat{y}) = [classification error] * [weight of correctly classified data points]
- ☒ **WM**(α, \hat{y}) = N * [classification error]
- ☐ **WM**(α, \hat{y}) = 1 - [classification error]

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2.

Refer to section **Example: Training a weighted decision tree**.

Will you get the same model as

small_data_decision_tree_subset_20 if you trained a decision tree with only 20 data points from the set of points in **subset_20**?

- ☒ Yes
- ☐ No
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3.

Refer to the 10-component ensemble of tree stumps trained with Adaboost.

As each component is trained sequentially, are the component weights monotonically decreasing, monotonically increasing, or neither?

- ☐ Monotonically decreasing
- ☐ Monotonically increasing
- ☒ Neither
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4.

Which of the following best describes a **general trend in accuracy** as we add more and more components? Answer based on the 30 components learned so far.

- ☐ Training error goes down monotonically, i.e. the training error reduces with each iteration but never increases.
- ☒ Training error goes down in general, with some ups and downs in the middle.
- ☐ Training error goes up in general, with some ups and downs in the middle.

- ☐ Training error goes down in the beginning, achieves the best error, and then goes up sharply.
- ☐ None of the above
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5.

From this plot (with 30 trees), is there massive overfitting as the # of iterations increases?

- ☐ Yes
- ☒ No
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