point

- Which of the following is NOT an ensemble method?
 - Gradient boosted trees
 - AdaBoost
 - Single decision trees

Random forests

2. Each binary classifier in an ensemble makes predictions on an input x as listed in the table below. Based on the ensemble coefficients also listed in the table, what is the final ensemble point model's prediction for x?

	Classifier coefficient w_t	Prediction for x
Classifier 1	0.61	+1
Classifier 2	0.53	-1
Classifier 3	0.88	-1
Classifier 4	0.34	+1

×

(True/False) Boosted trees tend to be more robust to overfitting than decision trees. point

- True
 - False

number of iterations.

True

point

point

point

point

point

point

5. Which of the following conditions must be true in order for $w_t > 0$?

4. (True/False) For AdaBoost, test error is an appropriate criterion for choosing the optimal

- False
- $weighted_error(f_t) < .25$
 - weighted_error(f_t) < .5 $weighted_error(f_t) > .75$
 - $weighted_error(f_t) > .5$
- 6. If you were using AdaBoost and in an iteration of the algorithm were faced with the following
- weighted error = 0.1 weighted error = 0.3
 - weighted error = 0.5

classifiers, which one would you be more inclined to include in the ensemble? A classifier with:

weighted error = 0.99

Weight

weighted error = 0.7

weights of the data. The data at this node is:

- Imagine we are training a decision stump in an iteration of AdaBoost, and we are at a node. Each data point is (x1, x2, y), where x1,x2 are features, and y is the label. Also included are the point
- 0.3 0 +1 1 0 -1 0.35

x1

ppose we s	plit on featu	re x2. Calculate the	weighted error of this s	split. Round your ans

 $8. \quad \text{After each iteration of AdaBoost, the weights on the data points are typically normalized to} \\$

x2

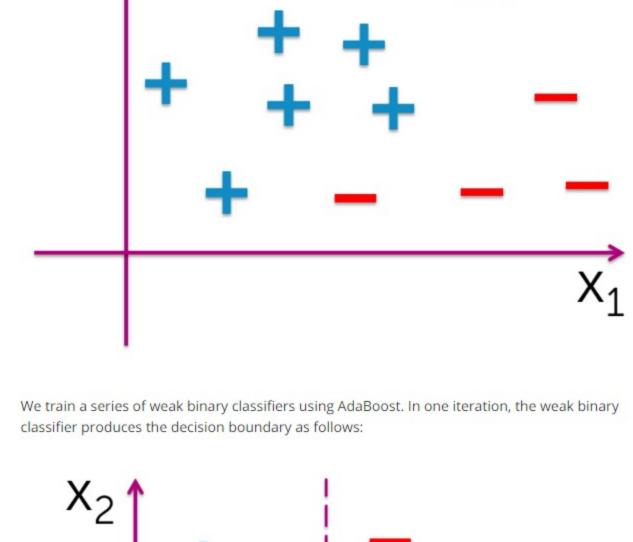
У

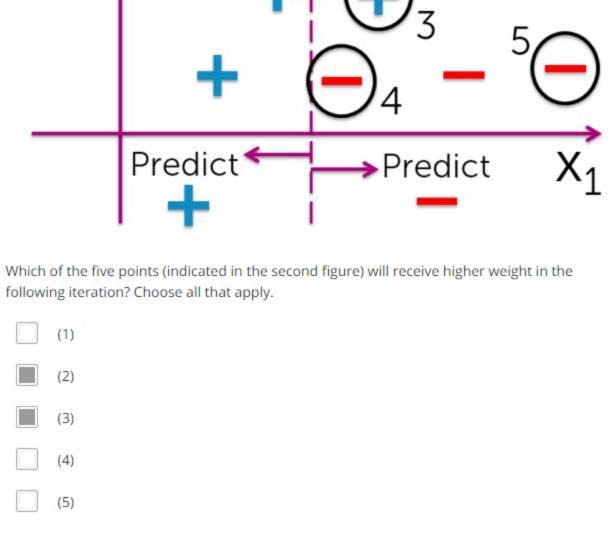
of issues with numerical instability (underflow/overflow) the weak learners can only learn with normalized weights

- Consider the following 2D dataset with binary labels.

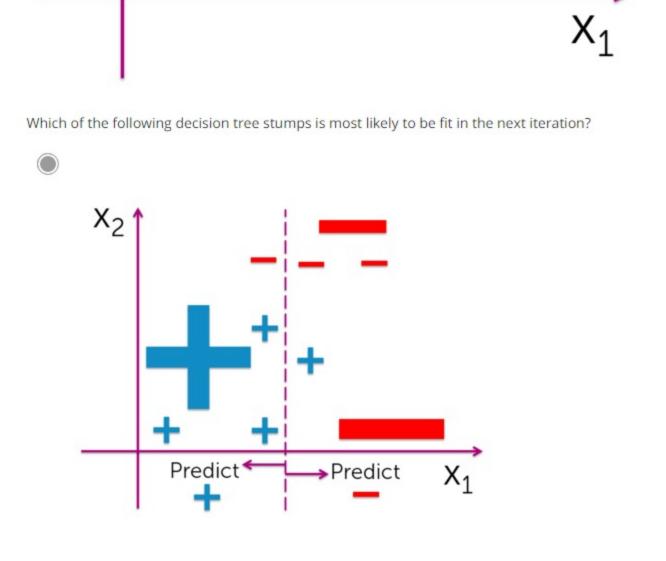
sum to 1. This is used because

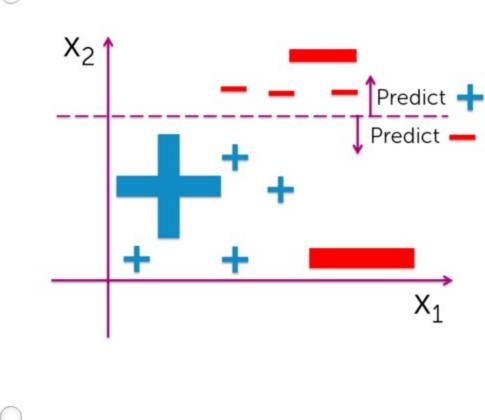
none of the above

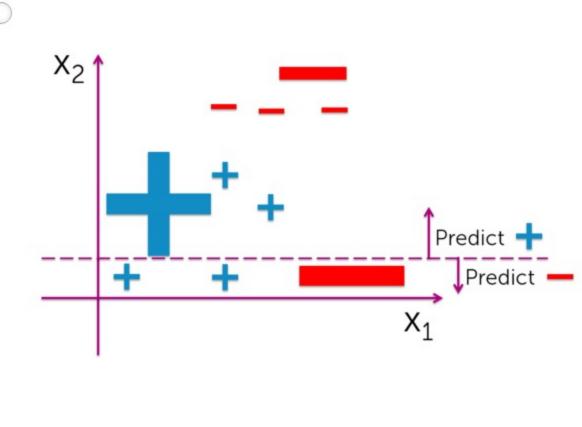




10. Suppose we are running AdaBoost using decision tree stumps. At a particular iteration, the data points have weights according the figure. (Large points indicate heavy weights.)







- 11. (True/False) AdaBoost achieves zero training error after a sufficient number of iterations, as long as we can find weak learners that perform better than random chance at each iteration of AdaBoost (i.e., on weighted data).
 - False

True

point