

Q1-1: Which of the following is/are correct regarding benefits of ensemble model?

- A. *Better performance*
- B. *Generalized models*
- C. *Better interpretability*

- 1. A, C
- 2. B, C
- 3. A, B
- 4. A, B, C

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- 3. A, B
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Interpretability is lost when ensemble model is used.

Q1-2: Which of the following statement(s) is/are correct about Bagging?

- A. In bagging, we choose random subsamples of the input points with replacement*
- B. The main purpose of bagging is to decrease the bias of learning algorithms.*
- C. Bagging is ineffective with logistic regression, because all of the learners learn exactly the same decision boundary*

- 1. A
- 2. B, C
- 3. A, C
- 4. A, B, C

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1. A 

2. B, C

3. A, C

4. A, B, C

The main purpose of bagging is to decrease the **variance** of learning algorithms.


Q2-1: Select the correct statement about AdaBoost.

- A. In AdaBoost weights of the misclassified examples go up by the same multiplicative factor.*
- B. In AdaBoost, if the weighted training error ϵ_t of the t^{th} weak classifier is large, then this classifier has smaller weights in the weighted majority vote for prediction*

1. Both the statements are TRUE.
2. Statement A is TRUE, but statement B is FALSE.
3. Statement A is FALSE, but statement B is TRUE.
4. Both the statements are FALSE.

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Statement A = True, follows from the update equation.

Statement B = True. See the ensemble prediction formula.

Q2-2: You've just finished training a random forest for spam classification, and it is getting abnormally bad performance on your validation set, but good performance on your training set. What could be causing the problem?

- A. *Your decision tree is too deep*
- B. *You have too few trees in your ensemble*
- C. *Your bagging implementation is randomly sampling sample points without replacement*
- D. *You are randomly sampling too many features when you choose a split*

- 1. A, B, C
- 2. B, C, D
- 3. A, C, D
- 4. A, B, C, D

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- 1. A, B, C
- 2. B, C, D
- 3. A, C, D
- 4. A, B, C, D



We observe overfitting.

A: large models can cause overfitting

B: too few trees (extreme case: one tree) will not reduce the variance, and thus doesn't prevent overfit

C: if without replacement, the random subsampled datasets = the original training dataset, so it's not doing bagging at all

D: if too many features, then the tree models have large capacity and can overfit

Q3-1: Suppose the minimum Hamming distance between any pair of codewords is 8. For which values of x can we still get the right classification with x single-bit errors? Choose the maximum possible value.

1. 1
2. 2
3. 3
4. 4

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1. 1
2. 2
3. 3
4. 4



$d = 8$, so this code can correct upto $\text{floor}([8 - 1]/2) = \text{floor}(3.5) = 3$ errors

Q3-2: Suppose there are 3 different classifier models with 70% accuracy. You make an ensemble model by using majority voting among these classifiers. Select the correct statement about this ensemble model.

1. *Maximum accuracy you can get = 100%.*
2. *Maximum accuracy you can get would be $\geq 70\%$, but cannot be 100%.*
3. *Maximum accuracy you can get would be $< 70\%$.*
4. Can't comment on Maximum accuracy of the ensemble model.

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M1, M2, M3 are 3 classifiers (70% accuracy). E = output of the ensemble model. Y = Desired output. Consider the case shown, where the three classifiers make errors on different data points. As we can see, it is able to achieve 100% accuracy.

M1	M2	M3	E	Y
1	1	0	1	1
1	1	0	1	1
1	1	0	1	1
1	1	1	1	1
1	0	1	1	1
1	0	1	1	1
1	0	1	1	1
0	1	1	1	1
0	1	1	1	1
0	1	1	1	1