CSE 471 (Machine Learning) Class Test 1 Student No (DON'T FORGET TO FILL):

Time: 30 minutes

Circle clearly the correct choice with pen (no pencil). Any ambiguity will be considered as wrong answ	Circle clearly	the correct choice with	pen (no	pencil). An	y ambiguity will	be considered	l as wrong answe
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- 1. ABCDE
- 2. A B C D E
- 3. ABCDE
- 4. A B C D E
- 5. A B C D E

DO NOT WRITE ANYTHING AFTER THE FOLLOWING DOTTED LINE

- 1) In Random forest you can generate hundreds of trees (say $T_1, T_2, ..., T_n$) and then aggregate the results of these trees. Which of the following is true about an individual (T_k) tree in Random Forest?
- 1. Individual tree is built on a subset of the features
- 2. Individual tree is built on all the features
- 3. Individual tree is built on a subset of observations
- 4. Individual tree is built on full set of observations
- A. 1 and 3
- B. 1 and 4
- C. 2 and 3
- D. 2 and 4
- E. None of these
- 2) Which of the following is/are true about boosting trees?
- 1. Individual weak learners are independent of each other
- 2. It is the method for improving the performance by aggregating the results of weak learners
- 3. Randomly generated bootstrap samples are used to train weak learner at each stage
- A. 1 and 2
- B. 2 and 3
- C. 1 and 3
- D. 1, 2 and 3
- E. None of these

- 3) Which of the following is/are true about bagging trees?
- 1. Individual weak learners are independent of each other
- 2. It is the method for improving the performance by aggregating the results of weak learners
- 3. Randomly generated bootstrap samples are used to train weak learner at each stage
- A. 1 and 2
- B. 2 and 3
- C. 1 and 3
- D. 1, 2 and 3
- E. None of these
- 4) Which of the following cannot be true about principal component analysis (PCA)?
- 1. PCA can be used with labeled data
- 2. Explicit testing for unit length of principal component is unnecessary if we use Lagrange's multiplier during optimization
- 3. The largest eigenvalue of the co-variance matrix represents the maximum variance of projection along first principal component
- A. 1 and 2
- B. 2 and 3
- C. 1 and 3
- D. 1, 2 and 3
- E. None of these
- 5) PCA maximizes variance of projections of sample vectors along
- A. orthogonal principal components
- B. axes of low-dimensional space
- C. eigenvectors of covariance matrix
- D. All of the above
- E. None of the above