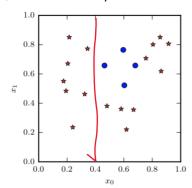
Instructions: Write all necessary steps that are needed. Submit scanned copy by 1pm.

Q1. Assume a binary classification setup with stars and bubbles as two classes shown in figure below.



A cut is made at  $x_0 = 0.4$  as shown by the red line. Compute Gini index for the left region and the right region.

[3]

Q2. Compute out of bag error for following. Suppose you are fitting a function to data distributed according to sin(x). Let the training samples be D = {(0, sin(0)), (45, sin(45)), (90, sin(90)), (225, sin(225)), (270, sin(270))}. Suppose you try to apply bagging to learn decision tree.

Let there be two sampled datasets from D. D is in form (x,y) where y is the label for x.

 $D1 = \{(0, \sin(0)), (45, \sin(45)), (90, \sin(90)), (225, \sin(225)), (225, \sin(225))\}$ 

 $D2 = \{(0, \sin(0)), (0, \sin(0)), (90, \sin(90)), (225, \sin(225)), (225, \sin(225))\}$ 

For both D1 and D2, let the learnt decision stump make a cut at x = 91. Compute OOB error for both D1 and D2.

[3]

Q3. Consider following 2-D samples  $\{(-1,-1), (-2,-2), (0,0), (2,2)\}$ . Assuming that k-means with k=2 is to be performed and initialized with cluster centers as  $\{(-1,-1),(0,0)\}$ , show:

a. Which points are assigned to which cluster? You must show all the steps

- [2]
- b. After assignment of points to cluster, compute cluster mean, within cluster covariance and between cluster covariance.

[3]

Q4. Consider the following training data

 $x = \{-1, 0, 2\}$  and  $f(x) = \{-4, 1, 4\}$ . Apply Gaussian process regression to find mean and variance at a test sample z=1. Assume a Gaussian kernel of the form

$$k(x,z) = \sigma^2 \exp\left[-\frac{(x-z)^2}{2l^2}\right]$$
, where  $\sigma = l = 1$ . [3]

Q5. The PCA objective formulation is given by  $T = T \cdot T$ 

$$\max_{w} w^T S w \text{ s.t } w^T w = 1$$

Show that the solution to this formulation is the eigenvector of S with maximum eigenvalue. Assume standard definition of symbols. [2]

Q6. Assuming W to be weights and X to be input, let the output of a neuron be  $\sigma(W^TX)$ , where  $\sigma$  denotes sigmoid. Give an expression for X in terms of W such that the  $\sigma(W^TX)$  attains the maximum possible value. Also assume that  $X^TX = 1$ .

[2]

Q7. Arrange in the increasing order of performance

- a. Single decision tree
- b. Bagged decision tree
- c. Random forest
- d. Give reasons. [2]