

1. Minimum Number of Binary classifiers required to classify N class classification problem

- a. $\log(N)$
- b. $\ln(N)$
- c. N
- d. $\ln(N)/\ln(2)$

D Assign a number from 0-N-1 to a class and convert it to binary, Use one classifier to predict one bit, The setup is not optimal but one can solve the problem with this.

2. Does a perceptron always converge?

- a. Yes
- b. No
- c. No, in some cases.

C converges for linearly separable data for others it need not converge.

3. You have observed oscillation in gradient descent of a convex function, What do you do to prevent/decrease oscillation?

- a. Increase learning rate
- b. Decrease learning rate
- c. Use second order methods to decide learning rate
- d. Give up hope

B Decreasing learning rate would decrease step size and reduce oscillation.

C Using Second order methods like Newton's method would scale the learning rate to prevent oscillation.

4. The output of a softmax activation function is bounded by

- a. 0,1
- b. -1,1
- c. It is unbounded
- d. None of the above

A, B

5. Given a two class classification problem with 2 features per sample then.

- a. SVM linear == Naive Bayes Gaussian
- b. SVM linear > Naive Bayes gaussian
- c. SVM linear < Naive Bayes Gaussian
- d. Can't Say.

C

A linear svm can never learn a nonlinear decision boundary hence performs worse on non linearly separable data hence a Naive bayes with gaussian distribution performs better as it can learn

When linearly separable data is present both perform equally well.

