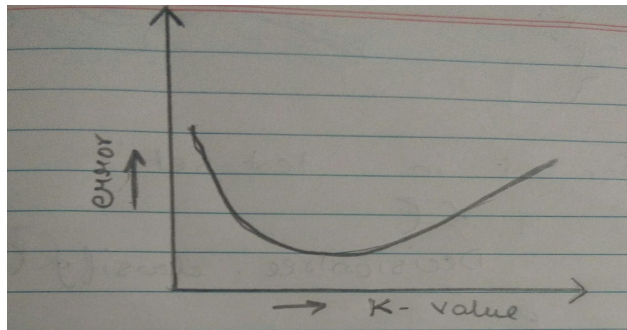


### Question 1-

- a) When  $k$  decrease from  $n$  to 1 training error will decrease
- When  $k=1$ , training error is zero as the distance of a point from itself is always 0
  - When  $k$  is larger, then that larger  $k$  value tries to bias the point toward other class also, as the points of both the classes are overlapped also
- b) Generalization error does not follow any fixed trend
- At a particular  $k$  value generalization error will be minimum and that depends on dataset
  - At first it decreases and then started increasing
  - The sketch between generalization error and  $k$  is



- c) When input dimensions are high
- Using too many features results in overfitting, the classifier starts learning exception so, when test data comes it cannot generalize well
  - As the input dimensions are high, memory requirements are high and time for computing distance between the points is high
- d) No
- Because in voronoi diagram for 1-NN the decision boundary is very complex whereas in a decision tree, the boundary the boundaries would always be parallel to the coordinate axes based on the decision node.
  - In decision tree it may possible that it may take multiple decisions but in 1-knn the decision is based on euclidean distance only.

