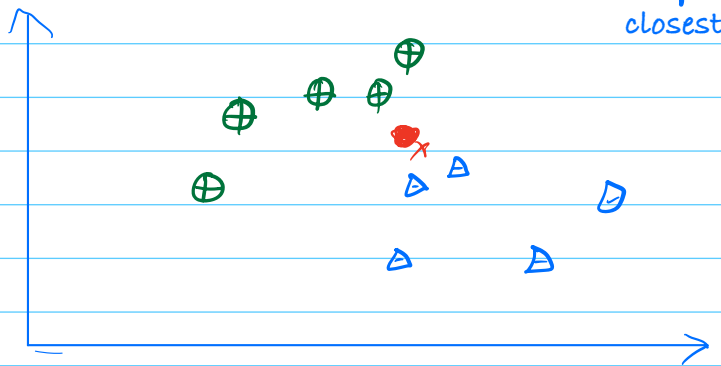




K-Nearest Neighbors Classifier

Assumption: Similar points are likely to share same label

Classification Rule: For a test point x , assign it label that is the most common label amongst the k most similar training instances.



Formally: Given $D = \{(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)\}$ and test point x

Find $S_x \subseteq D$ s.t. $|S_x| = k$ and $\forall (x', y') \in D \setminus S_x$

$$\text{dist}(x', x) \geq \max_{(x'', y'') \in S_x} \text{dist}(x'', x)$$

$$h(x) = \text{MODE}(y' : (x', y') \in S_x)$$

Pro-tip: In case of a tie, reduce k by 1 and repeat

Common distances: Minkowski distance or L_p distances

$$\text{dist}(x, x') = \left(\sum_{i=1}^d |x[i] - x'[i]|^p \right)^{1/p}$$

$p = 2$ is Euclidean distance

QUIZ: what distances do $p=0$, $p=1$ and $p \rightarrow \infty$ correspond to

Bayes Optimal Classifier

If we knew P how well could we do? $(x, y) \sim P$

Bayes Error: $\min_{\text{All possible } h} E[\ell(h(x), y)] (= P(h(x) \neq y))$

If you knew $P(Y|X=x)$, at given point x , optimal classifier:

$$h_{\text{opt}}(x) = \underset{y \in Y}{\text{argmax}} P(Y=y|X=x)$$

$$\text{Bayes Error}(x) = 1 - \max_{y \in Y} P(Y=y|X=x)$$

This is the Best we can do!

QUIZ: say a coin has probability p of falling heads when tossed

1. If the coin is tossed twice, what is the probability of two different outcomes?
2. Show that this probability is lesser than $2(1-p)$

1 - NN classifier: simplest case $k=1$

$$\text{Risk of 1-NN} \leq 2 \text{ Bayes ERROR}$$

Formal proof is involved, see Cover & Hart '67

Intuition:

1. Say P was a discrete distribution on a finite set of points. Then, as $n \rightarrow \infty$ every test point has already occurred in D_{TRAIN} . (say we pick any one of previous occurrence as the nearest neighbor)
2. Risk of 1-NN Classifier is now given by the quiz question. Why?

We are asking the question, what is the probability that, label y of a new test instance x matches that of a randomly chosen training point $x_i \in D_{\text{train}}$ such that, $x_i = x$. Its label y_i is drawn independently from $P(Y|X=x)$

$$\text{Hence, } P(h_{\text{1-NN}}(x) \neq y | X=x) \leq 2 \left(1 - \max_{y \in \{0,1\}} P(Y=y|X=x) \right)$$

Claim: Given x , let $\hat{x} \in D_{\text{TRAIN}}$ be the 1-NN of x in D_{TRAIN} as $n \rightarrow \infty$
 $\text{dist}(x, \hat{x}) \rightarrow 0, x \rightarrow \hat{x}$

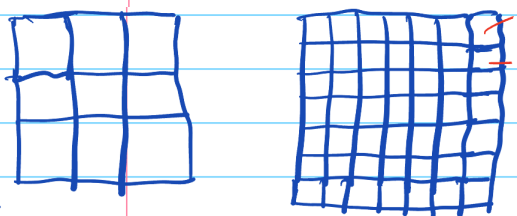
Risk of 1-NN ≤ 2 Bayes ERROR

k-NN for general, $k > 1$

a. Use larger k as n becomes larger.

Intuition: For point x , we predict as majority of k draws from $P(Y|X=x)$

b. But, if k grows too fast, then, more, we rely on farther away points to predict label for x



1. Each sub-cube has same area
2. if we randomly pick a few sub-cubes, we will pick more cubes near surface!

1. Assume points are drawn uniformly at random from a unit-hypercube

2 Hypercube of volume k/n within unit hypercube is expected to contain k out of n points

3 length l of such cube is given by $l^d = k/n$

What does this mean?

$k=10$

$n=1000$



d	2	10	100	1000
l	0.1	0.63	0.955	0.9954

$$l = \left(\frac{k}{n}\right)^{1/d}$$

