

Non-Parametric Learners: KNN, Decision Trees

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Objectives

At the end of today's lecture you should:

- 1 Be able to describe the KNN algorithm.
- 2 Describe the curse of dimensionality.
- 3 Recognize the conditions under which the curse may be problematic.
- 4 Enumerate strengths and weaknesses of KNN.

KNN

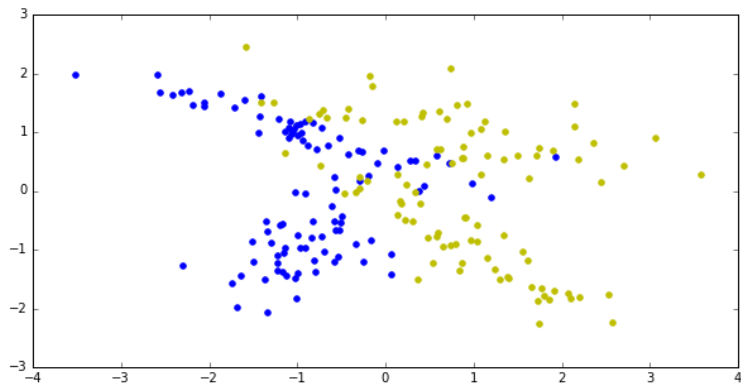


Figure 1: A classification problem

New data point

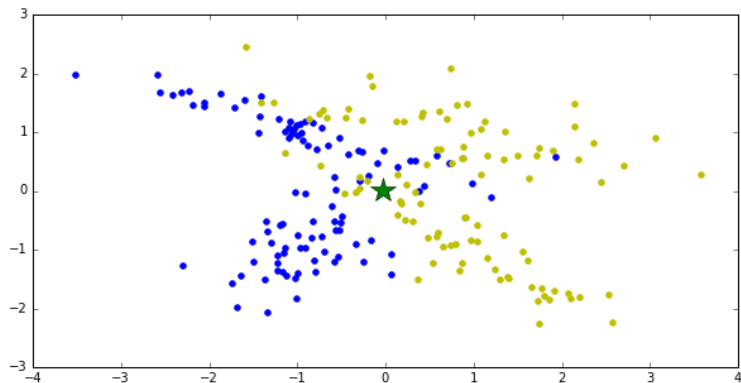


Figure 2: An unknown point

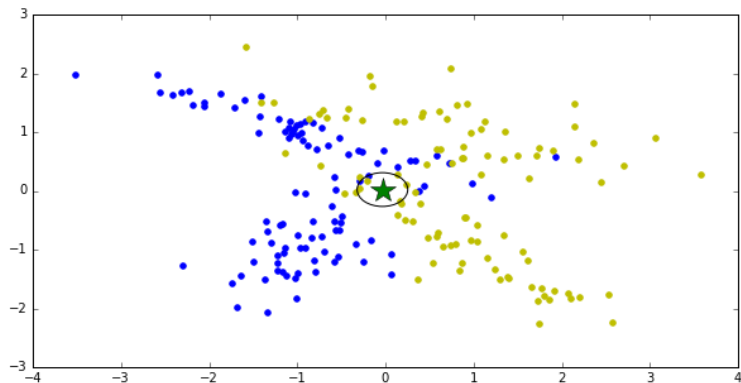


Figure 3: The approach

The KNN algorithm

Training:

- 1 Store all data.

Prediction:

- 1 Calculate the distance from new point to all points in dataset.
- 2 Keep the k nearest points to new point.
- 3 Predict the majority label.

What's k?

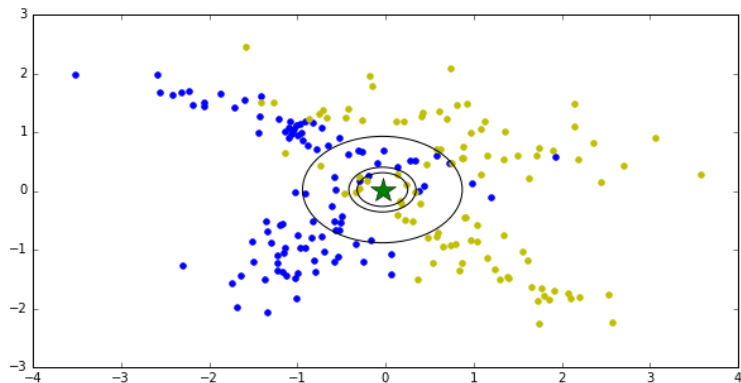


Figure 4: $k = 5, 10, 40$

1-neighbor

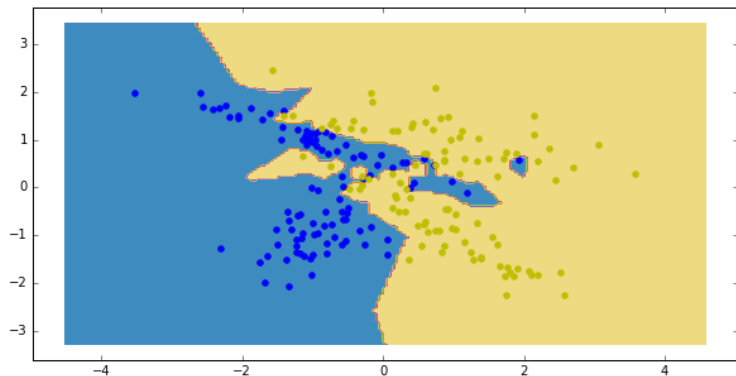


Figure 5: $k = 1$

3-neighbor

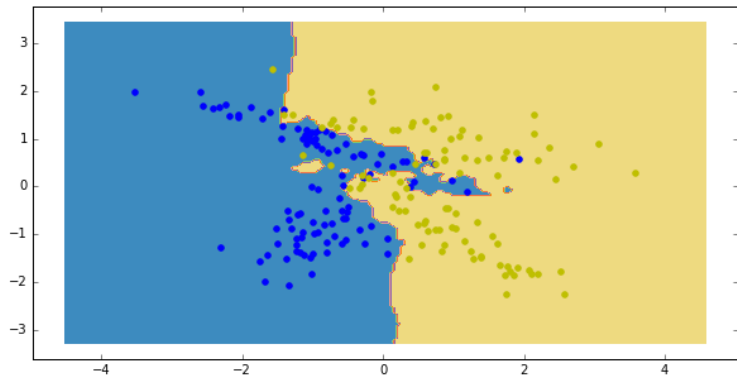


Figure 6: $k = 3$

10-neighbor

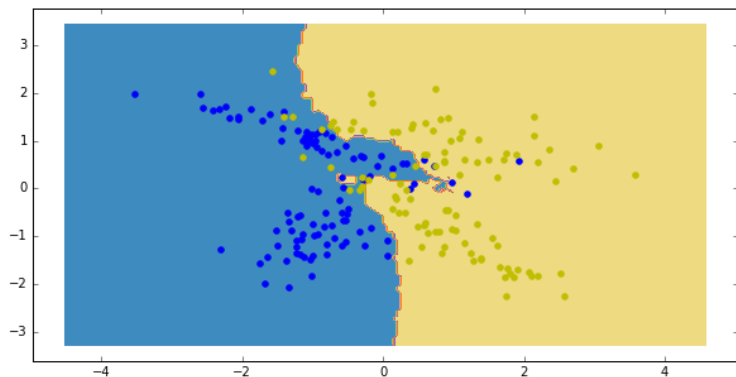


Figure 7: $k = 10$

100-neighbor

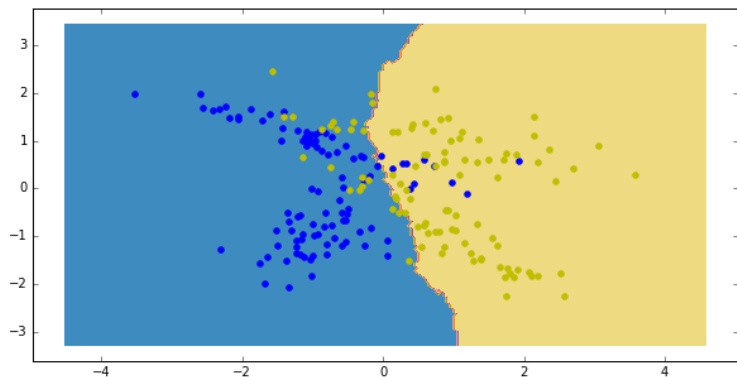


Figure 8: $k = 100$

Distance Metrics

So far we haven't been explicit about what distance metrics we're using.
Some choices:

- Euclidean:

$$\sqrt{\sum_i (a_i - b_i)^2}$$

- Manhattan:

$$\sum_i |a_i - b_i|$$

- Cosine:

$$1 - \frac{a \cdot b}{||a|| ||b||}$$

Variants

- One variant is to weight the votes by $\frac{1}{d_i}$ so closer points get more weight.
- Use for regression, take (optionally, weighted) mean of continuous target rather than vote.
- Approximate nearest neighbors, overcomes performance issues.

Curse of Dimensionality

Curse of dimensionality

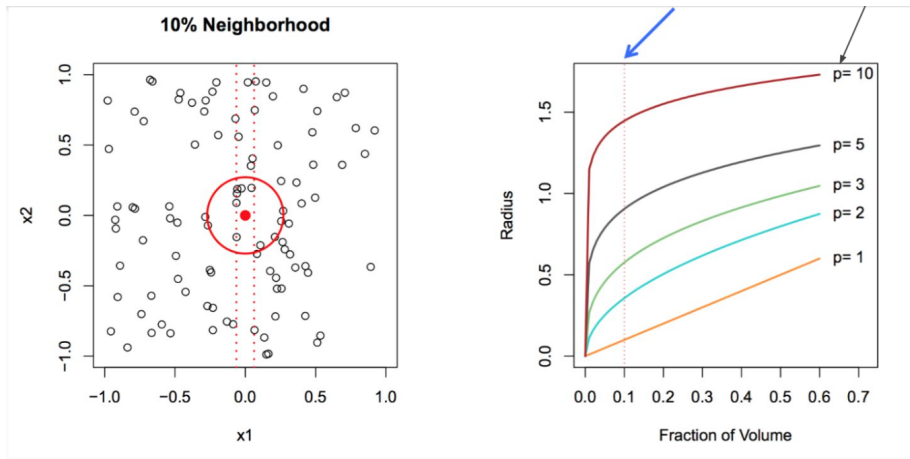


Figure 9: Curse of Dimensionality

Another view

Say we have a unit (hyper)cube.

We want to create another (hyper)cube inside the outer cube so that we fill X% of the outer cube.

How long must the edges of the inner cube be?

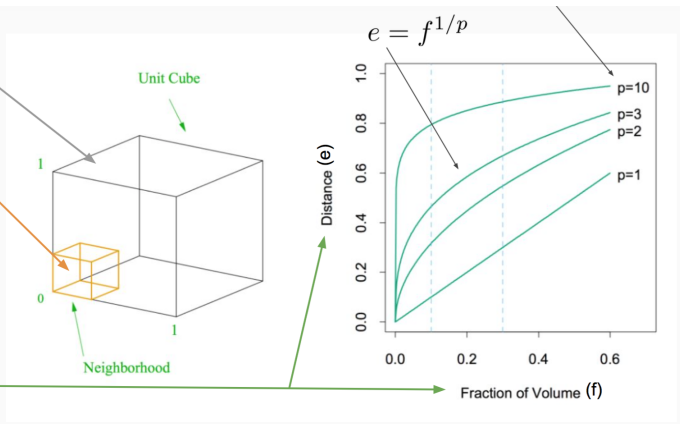


Figure 10: Hypercubes

Still another view

You have a dataset with 100 samples and one predictor.

You decide 1 predictor isn't enough so you decide to measure 10 predictors instead.

How many samples do you need to achieve the same sample density you originally had?

$$100^{10} = 1,000,000,000,000,000,000$$

Last one

$$\lim_{d \rightarrow \infty} \frac{V_{\text{sphere}}(R, d)}{V_{\text{cube}}(R, d)} = \lim_{d \rightarrow \infty} \frac{\frac{\pi^{d/2} R^d}{\Gamma(d/2+1)}}{(2R)^d} = \lim_{d \rightarrow \infty} \frac{\pi^{d/2}}{2^d \Gamma(d/2+1)} = 0$$