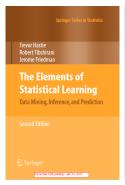
Machine Learning Lecture 3

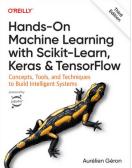


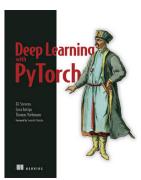




Literature







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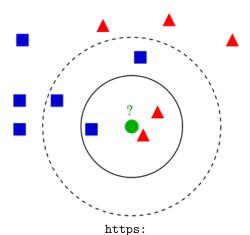
Content

- **▶** Weighted *k*NN
- ► *k*NN for Regression
- **▶** Pandas





Weighted k Nearest Neighbors (kNN)



//en.wikipedia.org/wiki/K-nearest_neighbors_algorithm





Weighted k Nearest Neighbors (kNN)

$$a(x) = \underset{c \in \mathbb{Y}}{\operatorname{argmax}} \sum_{s=1}^{k} w_{s} \left[y^{(i_{s})} = c \right].$$

$$\begin{array}{l} w_s = \frac{k+1-s}{k} \\ w_s = q^s \\ w_s = K\left(\frac{d(x,\ x^{(i_s)})}{h}\right), \ \text{where} \ K \ \text{is a non-increasing function defined} \\ \text{on} \ [0,\ 1]. \end{array}$$





k Nearest Neighbors (kNN) for Regression: training

- ► We are given $(x^{(i)}, y^{(i)}), i = 1, 2, ..., N$.
- $ightharpoonup \mathbb{Y} = \mathbb{R}$ (classification problem).
- training = memorizing of the given data.





k Nearest Neighbors (kNN) for Regression: prediction

- ▶ We have a new feature x.
- ▶ Define the distance between new feature and $x^{(i)}$: $d(x, x^{(i)})$.
- Rearrange the objects by the closeness to x:

$$d\left(x, \ x^{(i_1)}\right) \leq d\left(x, \ x^{(i_2)}\right) \leq \ldots \leq d\left(x, \ x^{(i_N)}\right).$$

► Look at the first *k* labels and assign the class with the highest number of representatives:

$$a(x) = \frac{1}{k} \sum_{s=1}^{k} y^{(i_s)}.$$





k Nearest Neighbors (kNN) Weighted Regression

$$a(x) = \frac{\sum\limits_{s=1}^{k} w_s y^{(i_s)}}{w_s}.$$

If
$$w_s = K\left(\frac{d(x, x^{(i_s)})}{h}\right)$$
, it is a Nadaraya — Watson estimator.





Pros

- ► If there are many examples, the algorithm gives a good prediction.
- Simple training.
- Few hyperparameters.
- ► There exist problems, where *k*NN works better, e.g., texts classification with many classes.





Cons

- ► Usually, other models work better.
- Extensive use of storage.
- ► Sorting for finding the neighbors.
- ► Limited tuning of the model.





Pandas

https://github.com/anton-selitskiy/RIT_ML



