DSC 40A

Theoretical Foundations of Data Science I

In This Video

The optimal prediction h^* that minimizes R(h) is a measure of center. What is the meaning of the value of $R(h^*)$?

Recommended Reading

Course Notes: Supplement 1

General Approach

- ▶ We start with a loss function L(h, y).
- ► Then we minimize the empirical risk:

$$R(h) = \frac{1}{n} \sum_{i=1}^{n} L(h, y_i).$$

- The input h^* that minimizes R(h) is some measure of the center of the data set.
- The minimum output $R(h^*)$ represents some measure of the **spread**, or variation, in the data set.

Absolute Loss

► The empirical risk for the absolute loss is

$$R(h) = \frac{1}{n} \sum_{i=1}^{n} |y_i - h|.$$

ightharpoonup R(h) is minimized at $h^* = \text{median}(y_1, y_2, \dots, y_n)$.

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- ightharpoonup Therefore, the minimum value of R(h) is

$$R(h^*) = R(\text{median}(y_1, y_2, \dots, y_n))$$

$$= \frac{1}{n} \sum_{i=1}^{n} |y_i - \text{median}(y_1, y_2, \dots, y_n)|.$$

Mean Absolute Deviation from the Median

The minimium value of R(h) is the mean absolute deviation from the median.

$$\frac{1}{n}\sum_{i=1}^{n}|y_i-\mathsf{median}(y_1,y_2,\ldots,y_n)|$$

It measures how far each data point is from the median, on average.

Question

For the data set 2,3,3,4, what is the mean absolute deviation from the median?

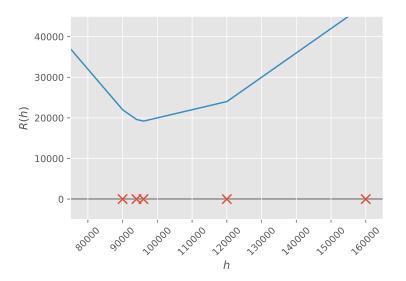
a) 0

 $\frac{1}{2}$

c)

d) 2

Mean Absolute Deviation from the Median



Square Loss

► The empirical risk for the square loss is

$$R_{sq}(h) = \frac{1}{n} \sum_{i=1}^{n} (y_i - h)^2.$$

 $ightharpoonup R_{sq}(h)$ is minimized at $h^* = mean(y_1, y_2, \dots, y_n)$.

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- $ightharpoonup R_{sq}(h)$ is minimized at $h^* = mean(y_1, y_2, \dots, y_n)$.
- ► Therefore, the minimum value of $R_{sq}(h)$ is

$$R_{sq}(h^*) = R_{sq}(mean(y_1, y_2, ..., y_n))$$

= $\frac{1}{n} \sum_{i=1}^{n} (y_i - mean(y_1, y_2, ..., y_n))^2$.

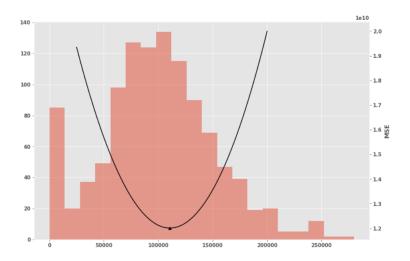
Variance

The minimium value of $R_{sq}(h)$ is the mean squared deviation from the mean, more commonly known as the variance.

$$\frac{1}{n}\sum_{i=1}^{n}(y_i - \text{mean}(y_1, y_2, \dots, y_n))^2$$

- It measures the squared distance of each data point from the mean, on average.
- Its square root is called the standard deviation.

Variance



0-1 Loss

► The empirical risk for the 0-1 loss is

$$R_{0,1}(h) = \frac{1}{n} \sum_{i=1}^{n} \begin{cases} 0, & \text{if } h = y_i \\ 1, & \text{if } h \neq y_i \end{cases}$$

- This is simply a count of the number of data points not equal to h.
- $ightharpoonup R_{0,1}(h)$ is minimized at $h^* = \mathsf{mode}(y_1, y_2, \dots, y_n)$.
- Therefore, $R_{0,1}(h^*)$ is a count of the number of data points not equal to the mode.

A Poor Way to Measure Spread

- The minimium value of $R_{0,1}(h)$ is the number of data points not equal to the mode.
- Higher value means less of the data is clustered at the mode.
- Just as the mode is a very simplistic way to measure the center of the data, this is a very crude way to measure spread.

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Question

For two different data sets, does it make sense say the data set with more data points not equal to the mode is more spread out?

Summary

- ▶ Different loss functions lead to empirical risk functions that are minimized at various measures of center.
- ► The minimum values of these risk runctions are various measures of spread.
- There are many different ways to measure both center and spread. These are sometimes called descriptive statistics.