PSC 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

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

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

#### **Absolute Loss**

The empirical risk for the absolute loss is

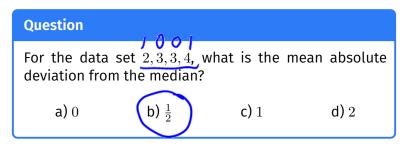
$$R(h) = \frac{1}{n} \sum_{i=1}^{n} |y_i - h|.$$

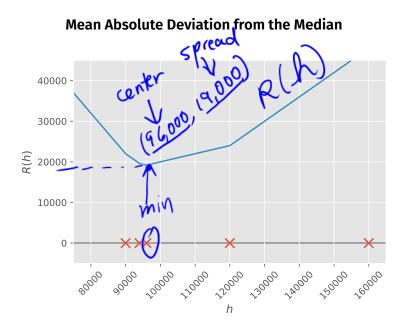
- ▶ R(h) is minimized at  $h^* = \text{median}(y_1, y_2, \dots, y_n)$ .
- Therefore, the minimum value of R(h) is

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

### Mean Absolute Deviation from the Median

- The minimium value of R(h) is the mean absolute deviation from the median. 2 $\frac{1}{n}\sum_{i=1}^{n} |y_i - \text{median}(y_1, y_2, \dots, y_n)|$
- It measures how far each data point is from the median, on average.





## **Square Loss**

► The empirical risk for the square loss is

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

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

### **Square Loss**

The empirical risk for the square loss is

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

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

• Therefore, the minimum value of  $R_{sq}(h)$  is

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

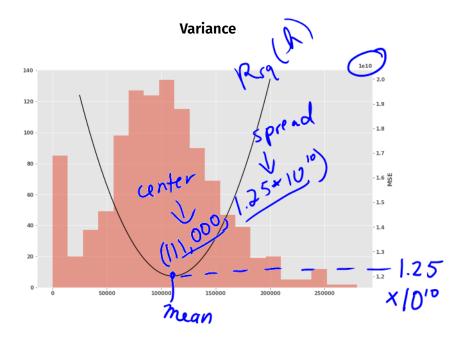
$$= \frac{1}{n} \sum_{i=1}^{n} (y_i - \text{mean}(y_1, y_2, \dots, y_n))^2$$
average squared deviations from the mean

# Variance

The minimium value of R<sub>sq</sub>(h) is the mean squared deviation from the mean, more commonly known as the variance.

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

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



### 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.
- ▶  $R_{0,1}(h)$  is minimized at  $h^* = \text{mode}(y_1, y_2, \dots, y_n)$ .
- ► 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.

### 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.

#### 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.