

**DSC 40A**

Lecture 02

Learning via Optimization, pt II

# Announcements

- ▶ Extension students: join Gradescope/Campuswire using codes found on [www.dsc40a.com](http://www.dsc40a.com)
- ▶ Need iClicker starting next week for tokens.

## **Last Time**

How do we turn the problem of learning into a math problem?

## Last Time

- ▶ What will be your future salary?

- ▶ Collect data:

90,000   94,000   96,000   120,000   160,000

- ▶ Could use the **mean** or the **median** as a prediction.
- ▶ But why?
- ▶ What is the best prediction?

## Last Time: The Mean Error of a Prediction

- Suppose we predicted a future salary of  $h_1 = 150,000$  *before* collecting data.

salary	error of $h_1$
90,000	60,000
94,000	56,000
96,000	54,000
120,000	30,000
160,000	10,000
total error: 210,000	
mean error: 42,000	

- A good prediction is one with small mean error.

## Last Time: The Best Prediction

- ▶ Any (non-negative) number is a valid prediction.
- ▶ Goal: out of all possible predictions, find the prediction  $h^*$  with the smallest **mean error**.
- ▶ This is an **optimization problem**.

# Today

We've turned learning into an **optimization problem**.  
How do we solve it?

## A Formula for the Mean Error

- ▶ We have data:

90,000   94,000   96,000   120,000   160,000

- ▶ Suppose our prediction is  $h$ .
- ▶ The **mean error** of our prediction is:

$$R(h) = \frac{1}{5} \left( |90,000 - h| + |94,000 - h| + |96,000 - h| \right. \\ \left. + |120,000 - h| + |160,000 - h| \right)$$



## A Formula for the Mean Error

- We have a function for computing the mean error of **any** possible prediction.

$$\begin{aligned} R(\mathbf{150,000}) &= \frac{1}{5} \left( |90,000 - \mathbf{150,000}| + |94,000 - \mathbf{150,000}| \right. \\ &\quad + |96,000 - \mathbf{150,000}| + |120,000 - \mathbf{150,000}| \\ &\quad \left. + |160,000 - \mathbf{150,000}| \right) \\ &= \mathbf{42,000} \end{aligned}$$

## A Formula for the Mean Error

- We have a function for computing the mean error of **any** possible prediction.

$$\begin{aligned} R(\mathbf{115,000}) &= \frac{1}{5} \left( |90,000 - \mathbf{115,000}| + |94,000 - \mathbf{115,000}| \right. \\ &\quad + |96,000 - \mathbf{115,000}| + |120,000 - \mathbf{115,000}| \\ &\quad \left. + |160,000 - \mathbf{115,000}| \right) \\ &= \mathbf{23,000} \end{aligned}$$

## A Formula for the Mean Error

- We have a function for computing the mean error of **any** possible prediction.

$$\begin{aligned} R(\pi) &= \frac{1}{5} \left( |90,000 - \pi| + |94,000 - \pi| \right. \\ &\quad + |96,000 - \pi| + |120,000 - \pi| \\ &\quad \left. + |160,000 - \pi| \right) \\ &= 111,996.8584... \end{aligned}$$

## A General Formula for the Mean Error

- ▶ Suppose we collect  $n$  salaries,  $y_1, y_2, \dots, y_n$ .
  - ▶ The mean error of the prediction  $h$  is:
- 

- ▶ Or, using **summation notation**:
-

## The Best Prediction

- ▶ We want the best prediction,  $h^*$ .
- ▶ The smaller  $R(h)$ , the better  $h$ .
- ▶ Goal: find  $h$  that minimizes  $R(h)$ .

### Discussion Question

Can we use calculus to minimize  $R$ ?

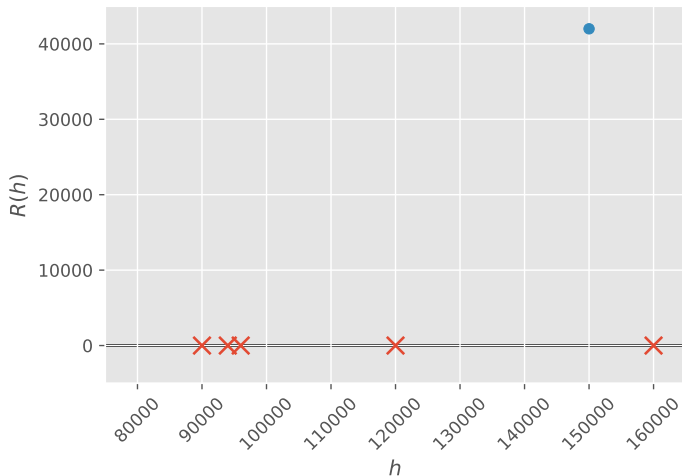
# Minimizing with Calculus

- ▶ Calculus: take derivative, set equal to zero, solve.

## Uh oh

- ▶  $R$  is **not differentiable**.
- ▶ We can't use calculus to minimize it.
- ▶ Let's try plotting  $R(h)$

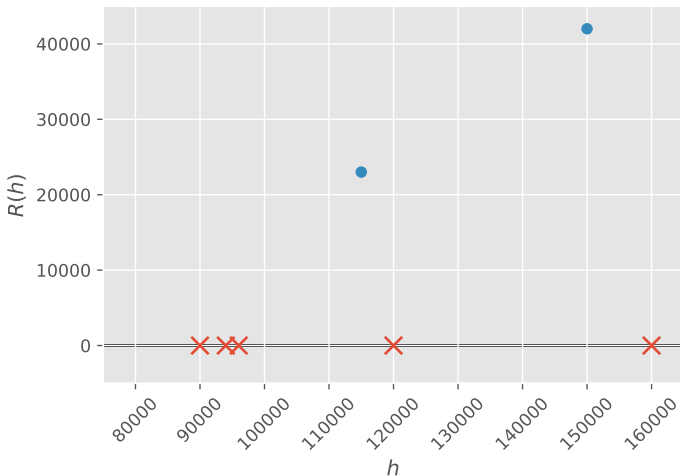
# Plotting the Mean Error



Recall:  $R(150,000) = 42,000$

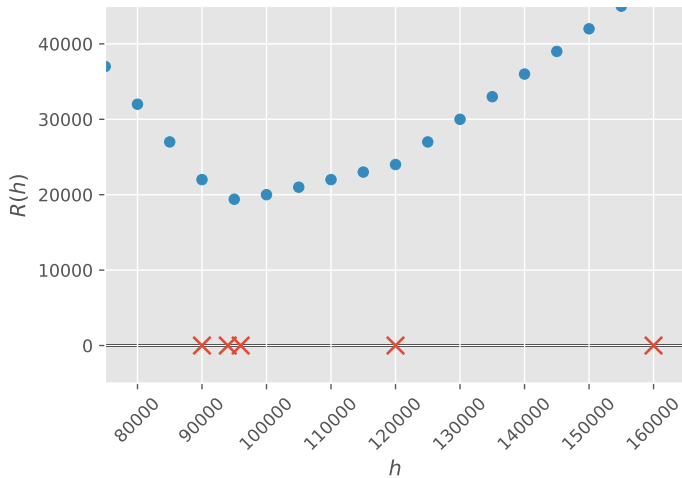


# Plotting the Mean Error

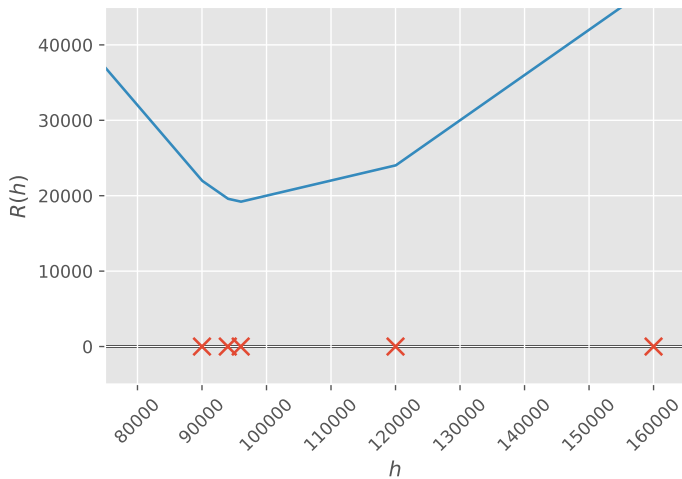


Recall:  $R(115,000) = 23,000$

# Plotting the Mean Error



# Plotting the Mean Error

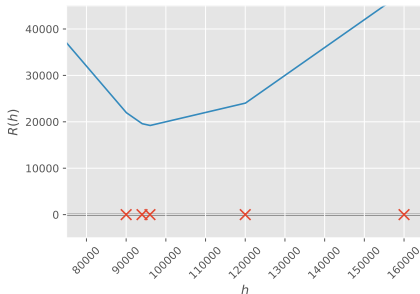


### Discussion Question

A local minimum occurs when the slope goes from \_\_\_\_\_. Select all that apply.

- A) positive to negative
- B) negative to positive
- C) positive to zero.
- D) negative to zero.
- E) zero to zero.

# Goal



- Find where slope of  $R$  goes from negative to non-negative.
- Want a formula for the slope of  $R$  at  $h$ .

## Sums of Linear Functions

- ▶ Let  $f_1(x) = 3x + 2$
- ▶ Let  $f_2(x) = 5x + 1$
- ▶ What is the slope of  $f(x) = f_1(x) + f_2(x)$ ?

## Sums of Absolute Values

- ▶ Let

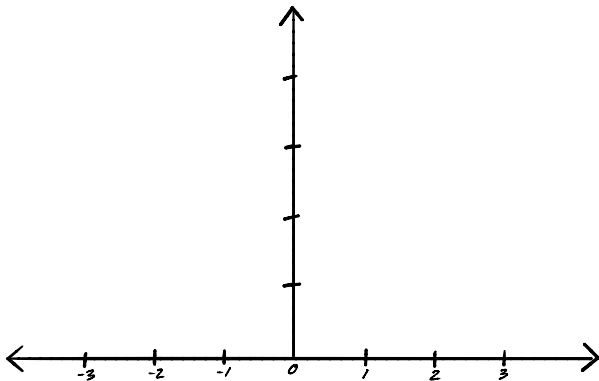
$$g_1(x) = |x - 2| \quad g_2(x) = |x + 1|$$

- ▶ Let  $g(x) = g_1(x) + g_2(x)$ .

### Discussion Question

What is the slope of  $g$  at  $x = 1$ ?

**Answer**





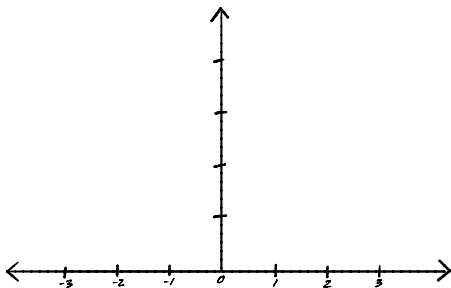
## Sums of More Absolute Values

- ▶ Let  $y_1 < y_2 < y_3$

$$h_1(x) = |x - y_1| \quad h_2(x) = |x - y_2| \quad h_3(x) = |x - y_3|$$

- ▶ Let  $h(x) = h_1(x) + h_2(x) + h_3(x)$ .
- ▶ The slope changes at  $y_1, y_2, y_3$ .

## Sums of More Absolute Values



- ▶ Slope when  $x < y_1$ :
- ▶ Slope when  $y_1 < x < y_2$ :
- ▶ Slope when  $y_2 < x < y_3$ :
- ▶ Slope  $x > y_3$ :

Slope at  $x = (\# \text{ of } y_i' \text{'s } \_\_\_\_\_\_ x) - (\# \text{ of } y_i' \text{'s } \_\_\_\_\_\_ x)$

# The Slope of Error Function

- $R$  is the sum of absolute value functions (times  $\frac{1}{n}$ ):

$$R(h) = \frac{1}{n} (|h - y_1| + |h - y_2| + \dots + |h - y_n|)$$

- So:

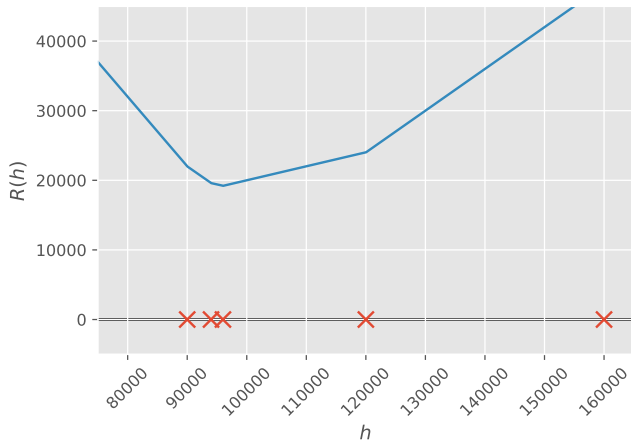
$$\text{Slope at } h = \frac{1}{n} \cdot [(\# \text{ of } y_i' \text{'s } \leq h) - (\# \text{ of } y_i' \text{'s } < h)]$$

## Discussion Question

Suppose that  $n$  is odd. At what value of  $h$  does the slope go from negative to positive?

- A)  $h$  = mean of  $y_1, \dots, y_n$   
 B)  $h$  = median of  $y_1, \dots, y_n$   
 C)  $h$  = mode of  $y_1, \dots, y_n$

# Where the Slope's Sign Changes



# The Median Minimizes the Mean Error

- ▶ Our problem was: find  $h^*$  which minimizes the mean error,  $R(h) = \frac{1}{n} \sum_{i=1}^n |y_i - h|$ .
- ▶ The answer is:  $\text{Median}(y_1, \dots, y_n)$ .
- ▶ The **best prediction**<sup>1</sup> is the **median**.

---

<sup>1</sup>in terms of mean error

## Status Update

- ▶ Last time, we turned predicting salary into a math problem: minimize the mean error.
- ▶ Today: we solved it. The **median** minimizes the mean error.

## What's Left?

- ▶ We did all this because  $R(h)$  isn't differentiable.
- ▶ What if we tried to minimize a *different* measure of error that *is* differentiable?