Lecture 2 – Minimizing Mean Absolute Error



DSC 40A, Winter 2024

Announcements

- Look at the readings linked on the course website!
- Homework 1 is out today and due Wed., January 17 midnight.
- See Calendar on course website for office hours.
 - Plan to come to office hours at least once a week for help on homework.

Agenda

- 1. Recap from Lecture 1 learning from data.
- 2. Minimizing mean absolute error.
- 3. Identifying another choice of error.

Recap from Lecture 1 – learning from data

What is good/bad, intuitively?

The data:

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

Consider these hypotheses:

$$h_1 = 150,000$$
 $h_2 = 115,000$



Quantifying our intuition

- Intuitively, a good prediction is close to the data.
- Suppose we predicted a future salary of h₁ = 150,000 before collecting data.

salary	absolute error of h_1
90,000	60,000
94,000	56,000
96,000	54,000
120,000	30,000
160,000	10,000
	sum of absolute errors: 210,000

mean absolute error: 42,000

Quantifying our intuition

Now suppose we had predicted h_2 = 115,000.

salary	absolute error of h_2
90,000	25,000
94,000	21,000
96,000	19,000
120,000	5,000
160,000	45,000
	sum of absolute errors: 115,000

mean absolute error: 23,000

Mean absolute error (MAE)

Mean absolute error on data:

 $h_1: 42,000$ $h_2: 23,000$

- Conclusion: h_2 is the better prediction.
- In general: pick prediction with the smaller mean absolute error.

We are making an assumption...

- We're assuming that future salaries will look like present salaries.
- That a prediction that was good in the past will be good in the future.

Discussion Question

Is this a good assumption?

Which is better: the mean or median?

Recall:

mean = 112,000 median = 96,000

We can calculate the mean absolute error of each:

mean : 22,400 median : 19,200

The median is the best prediction so far!

But is there an even better prediction?

Finding the best prediction

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

We have data:

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

- Suppose our prediction is *h*.
- ► The mean absolute error of our prediction is: $R(h) = \frac{1}{5} (|90,000 - h| + |94,000 - h| + |96,000 - h|) + |120,000 - h| + |160,000 - h|)$

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

$$R(150,000) = \frac{1}{5} (|90,000 - 150,000| + |94,000 - 150,000| + |96,000 - 150,000| + |120,000 - 150,000| + |160,000 - 150,000|)$$

= 42,000

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

$$R(115,000) = \frac{1}{5} (|90,000 - 115,000| + |94,000 - 115,000| + |96,000 - 115,000| + |120,000 - 115,000| + |160,000 - 115,000|)$$

= 23,000

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

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

= 111,996.8584...

Discussion Question

Without doing any calculations, which is correct? A. *R*(50) < *R*(100) B. *R*(50) = *R*(100) C. *R*(50) > *R*(100)

Suppose we collect *n* salaries, $y_1, y_2, ..., y_n$.

► The mean absolute 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*).

Summary

We started with the learning problem:

Given salary data, predict your future salary.

We turned it into this problem:

Find a prediction h^{*} which has smallest mean absolute error on the data.

- We have turned the problem of learning from data into a specific type of math problem: an optimization problem.
- **Now**: we solve this math problem.

Minimizing mean absolute error

Many possible predictions

h

h

So far, we have considered four possible hypotheses for future salary, and computed the mean absolute error of each.

$$h_1 = 150,000 \implies R(150,000) = 42,000$$

$$h_2 = 115,000 \implies R(115,000) = 23,000$$

▶
$$h_3$$
 = mean = 112,000 \implies $R(112,000) = 22,400$

$$h_{L} = \text{median} = 96,000 \implies R(96,000) = 19,200$$

Of these four options, the median has the lowest MAE. But is it the **best possible prediction overall**?

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

$$R(h) = \frac{1}{n} \left(|h - y_1| + |h - y_2| + \dots + |h - y_n| \right)$$
$$= \frac{1}{n} \sum_{i=1}^n |h - y_i|$$

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 with respect to h, set equal to zero, solve.

Minimizing with calculus

Calculus: take derivative with respect to h, 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) instead.

Plotting the mean absolute 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.

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

$$f_1(x) = 3x + 7$$
 $f_2(x) = 5x - 4$ $f_3(x) = -2x - 8$

• What is the slope of $f(x) = f_1(x) + f_2(x) + f_3(x)$?

Absolute value functions

Recall, f(x) = |x - a| is an absolute value function centered at x = a.



Sums of absolute values

Let

$$f_1(x) = |x-2|$$
 $f_2(x) = |x+1|$ $f_3(x) = |x-3|$

• What is the slope of $f(x) = f_1(x) + f_2(x) + f_3(x)$?





The slope of the mean absolute error

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

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

The slope of the mean absolute error

The slope of R at h is:

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



Where the slope's sign changes

The slope of R at h is:

$$\frac{1}{n} \cdot \left[(\# \text{ of } y_i \text{'s} < h) - (\# \text{ of } y_i \text{'s} > h) \right]$$

Discussion Question

Suppose that n is odd. At what value of h does the slope of R go from negative to non-negative?

A)
$$h = \text{mean of } y_1, \dots, y_n$$

B) $h = \text{median of } y_1, \dots, y_n$
C) $h = \text{mode of } y_1, \dots, y_n$

The median minimizes mean absolute error, when *n* is odd

- Our problem was: find h^* which minimizes the mean absolute error, $R(h) = \frac{1}{n} \sum_{i=1}^{n} |y_i - h|$.
- We just determined that when n is odd, the answer is Median(y₁,..., y_n). This is because the median has an equal number of points to the left of it and to the right of it.
- But wait what if n is even?

Discussion Question

Consider again our example dataset of 5 salaries.

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

Suppose we collect a 6th salary, so that our data is now

90,000 94,000 96,000 108,000 120,000 160,000

Which of the following correctly describes the h^* that minimizes mean absolute error for our new dataset?

- A) 96,000 only
- B) 108,000 only
- C) 102,000 only
- D) Any value in the interval [96,000, 108,000]

Plotting the mean absolute error, with an even number of data points



What do you notice?

The median minimizes mean absolute error

- Our problem was: find h^* which minimizes the mean absolute error, $R(h) = \frac{1}{n} \sum_{i=1}^{n} |y_i - h|$.
- Regardless of if n is odd or even, the answer is h* = Median(y₁,..., y_n). The best prediction, in terms of mean absolute error, is the median.
 - ▶ When *n* is odd, this answer is unique.
 - When n is even, any number between the middle two data points also minimizes mean absolute error.
 - We define the median of an even number of data points to be the mean of the middle two data points.

Identifying another type of error

Two things we don't like

- 1. Minimizing the mean absolute error wasn't so easy.
- 2. Actually **computing** the median isn't so easy, either.
- Question: Is there another way to measure the quality of a prediction that avoids these problems?

The mean absolute error is not differentiable

We can't compute
$$\frac{d}{dh}|y_i - h|$$
.

Remember: $|y_i - h|$ measures how far h is from y_i .

- ▶ Is there something besides $|y_i h|$ which:
 - 1. Measures how far *h* is from *y*_i, and
 - 2. is differentiable?

The mean absolute error is not differentiable

► We can't compute
$$\frac{d}{dh}|y_i - h|$$
.

Remember: $|y_i - h|$ measures how far h is from y_i .

- Is there something besides |y_i h| which:
 - 1. Measures how far h is from y_i, and
 - 2. is differentiable?

Discussion Question

Which of these would work?

- b) $|y_i h|^2$
- a) $e^{|y_i h|}$ c) $|y_i h|^3$ d) $\cos(y_i - h)$

Why?

Summary

Summary

- > Our first problem was: find h* which minimizes the mean absolute error, R(h) = 1/n ∑_{i=1}ⁿ |y_i h|.
 > The answer is: Median(y₁,..., y_n).
 - The best prediction, in terms of mean absolute error, is the median.
- We then started to consider another type of error that is differentiable and hence is easier to minimize.
- Next time: We will find the value of h* that minimizes this other error, and see how it compares to the median.