

Least Squares: Part 1

Lecture 3-1 - CMSE 382

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Topics covered

- Ordinary Least squares (OLS)
- Linear fitting
- Polynomial fitting

Section 1

Least squares

Overdetermined systems

Recall Matrix Rank

The rank of matrix $A \in \mathbb{R}^{m \times n}$ is a non-negative integer $\text{rank}(A) \leq \max(m, n)$ given by any of the following:

- the maximum number of its linearly independent rows or columns.
- the number of non-zero rows in its row-echelon form
- the number of non-zero singular values in its SVD

A linear system of the form $\mathbf{Ax} = \mathbf{b}$, where $\mathbf{A} \in \mathbb{R}^{m \times n}$ is of full rank and $\mathbf{b} \in \mathbb{R}^m$ is

- **overdetermined** if $m > n$.
 - ▶ Almost always inconsistent (has no solution)
- **determined** if $m = n$, and the equations are independent
 - ▶ Has a unique solution
- **underdetermined** if $m < n$
 - ▶ Either inconsistent (no solution), or infinitely many solutions

Goal: Obtain an approximate solution for overdetermined systems

Ordinary Least Squares Approximation

Definition

The **residual sum of squares error** (or total sum of square errors) is

$$\text{RSS} = \|\mathbf{A}\mathbf{x}_{\text{LS}} - \mathbf{b}\|^2$$

The minimizer of this equation is the **least squares estimate**

$$\mathbf{x}_{\text{LS}} = (\mathbf{A}^\top \mathbf{A})^{-1} \mathbf{A}^\top \mathbf{b}$$

- RSS measures how well \mathbf{x}_{LS} approximates the data \mathbf{b}
- The Least squares estimate minimizes RSS

$$\min_{\mathbf{x} \in \mathbb{R}^n} \|\mathbf{A}\mathbf{x} - \mathbf{b}\|^2$$

- You did this in CMSE381, but that textbook used the notation β_i for the coefficients. In this notation, the entries of \mathbf{x} are the coefficients of the linear function we're trying to learn.

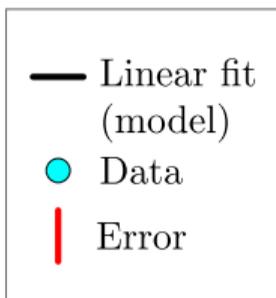
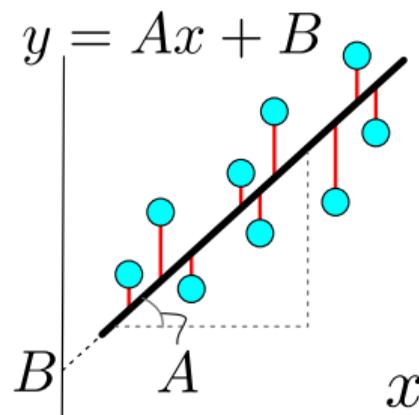
Linear Data Fitting

- Given a set of data points (\mathbf{s}_i, t_i) , $i = 1, 2, \dots, m$, where $\mathbf{s}_i \in \mathbb{R}^n$ and $t_i \in \mathbb{R}$
- Assume that the target t_i can be approximated by the linear transformation of the data \mathbf{s}_i

$$t_i \approx \mathbf{s}_i^\top \mathbf{x}$$

- Then the least squares method is applied to find the parameter \mathbf{x} that solves the following optimization problem:

$$\min_{\mathbf{x} \in \mathbb{R}^n} \sum_{i=1}^m (\mathbf{s}_i^\top \mathbf{x} - t_i)^2 = \min_{\mathbf{x} \in \mathbb{R}^n} \|\mathbf{S}\mathbf{x} - \mathbf{t}\|^2$$



Polynomial fitting

If we know that the points are approximately related to a polynomial of degree at most d , i.e., there exists a_0, \dots, a_d such that

$$\sum_{j=0}^d a_j u_i^j \approx y_i, \quad i = 1, \dots, m$$

we can use OLS $\min_{\mathbf{x} \in \mathbb{R}^n} \|\mathbf{S}\mathbf{x} - \mathbf{t}\|^2$ with

$$\begin{bmatrix} 1 & u_1 & u_1^2 & \dots & u_1^d \\ 1 & u_2 & u_2^2 & \dots & u_2^d \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 1 & u_m & u_m^2 & \dots & u_m^d \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_d \end{bmatrix} = \begin{bmatrix} y_0 \\ y_1 \\ \vdots \\ y_m \end{bmatrix}$$

$$\mathbf{S} = \begin{bmatrix} 1 & u_1 & u_1^2 & \dots & u_1^d \\ 1 & u_2 & u_2^2 & \dots & u_2^d \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 1 & u_m & u_m^2 & \dots & u_m^d \end{bmatrix}$$

$$\mathbf{x} = \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_d \end{bmatrix} \quad \mathbf{t} = \begin{bmatrix} y_0 \\ y_1 \\ \vdots \\ y_m \end{bmatrix}$$

Groups - Round 2

Group 1

Abigail, Shanze, Jack,
Quang Minh,

Group 2

Igor, Atticus, K M
Tausif, Long,

Group 3

Yousif, Zheng, Jake,
Purvi,

Group 4

Maye, Alice, Arjun,
Kyle,

Group 5

Monirul Amin, Jay,
Brandon, Luis,

Group 6

Scott, Ha, Lora,
Tianjian,

Group 7

Braedon, Sai, Joseph,
Noah,

Group 8

Michal, Aidan, Jonid,
Dev,

Group 9

Vinod, Saitej, Anthony,
Breena,

Group 10

Karen, Dori, Lowell,
Aaron,

Group 11

Jamie, Sanskaar,
Dominic, Lauryn,

Group 12

Andrew, Arya, Daniel,
Morgan,

Next time

- Check course webpage for videos and reading for next class
- Office hours posted on the course webpage
- Homework posted, due Friday Jan 30 at 11:59pm
- Quiz 2 on Weds 2/4