

The method of Least Squares

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In numerical analysis, the method of least squares is a classical method that used to obtain a approximation for some mathematical problems.

A classical problem is given A and b , solve x

$$Ax = b$$

But sometimes the true x cannot be found, since the system is inconsistent / contradictory, for example

$$\begin{cases} x_1 + x_2 = 3 \\ x_1 + x_2 = 2 \\ x_1 - x_2 = 1 \end{cases}$$

Equation 1 and 2 contradict to each other. This kind of inconsistent system does not have exact true solution. Then it is ok to find a *numerical approximation*

In matrix form $Ax = b$

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} \iff Ax = b$$

Perform the following

$$A^T Ax = A^T b \implies \bar{x} = (A^T A)^{-1} A^T b \iff \bar{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.75 \\ 0.75 \end{bmatrix}$$

Define the residue $r = A\bar{x} - b$

$$r = \begin{bmatrix} -0.5 \\ 0.5 \\ 0 \end{bmatrix}$$

The L^2 norm of the r , which is also called the *Eulclidean Distance* is thus

$$\|r\|_2 = \sqrt{(-0.5)^2 + (0.5)^2 + 0^2} = 0.707$$

The L^2 norm indicate the "size" of the error, and in this case, the error is "samll"

Why use L^2 norm ? Actually, we can use L^1 , L^3 , L^∞ . But L^2 norm is most simple. Further more, L^2 is larger then L^∞ norm

Then what is the *best* approximation for the problem $Ax = b$? Obviously, it is the x_{best} that minimize $\|r\|_2$

Thus , the general problem for a inconsistent system is thus

For a given system A , and meaurment vecotr b , and the unknown vector x having the relation as

$$Ax = b$$

The best approximation of x can be found by

$$\bar{x}_{best} = \min \|r\|_2 = \min \|A\bar{x}_{best} - b\|_2 = \min \|A\left((A^T A)^{-1} A^T b\right) - b\|_2$$

$$\bar{x}_{best} = \min \|(AA^\dagger - I)b\|_2$$

Where A^\dagger is the *pseudo - inverse*, or called *generalized inverse* of A : $A^\dagger = (A^T A)^{-1} A^T$

- One basic property of generalized inverse of matrix A is that such relation is unique (How to prove it ? By standard method , assume 2 matrix then prove them are equal)

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