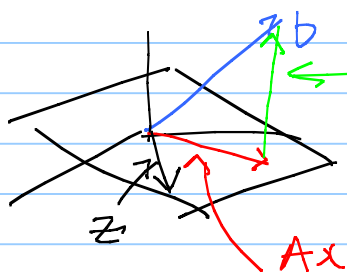


# Least Squares

Note Title

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Let  $A$  be an  $m \times n$  matrix and consider the equation  $Ax = b$ , which may or may not have a (unique?) solution. The method of least squares finds a "best" solution, which is the unique solution if it exists. Let  $A = [A_1, \dots, A_n]$  be written as an array of column  $m$ -vectors. Let  $\Pi$  be the column space of  $A$ , which is the set of linear combinations of the columns of  $A$ .  $y \in \Pi \Leftrightarrow y = x_1 A_1 + \dots + x_n A_n = Ax$ . The vector  $b$  may not be in  $\Pi$  and even if it is it may not be a unique combination of the columns. The "best" solution will be a vector  $x$  so that  $\|b - Ax\|$  is least. Choose  $x$  so that  $b - Ax$  is orthogonal to  $\Pi$ .



Let  $z$  be any vector in  $\Pi$ .

$$\|b - z\|^2 = \|b - Ax + Ax - z\|^2$$

$$= \|b - Ax\|^2 + \|Ax - z\|^2 \geq \|b - Ax\|^2$$

since  $Ax - z \in \Pi$ .

Thus  $\|b - Ax\|$  is least. Is  $x$  uniquely determined?

The condition on  $x$  is  $b - Ax \perp A_j, j = 1, \dots, n$ .

This can be written as

$$(v) \quad A^T A x = A^T b.$$

Suppose the columns of  $A$  are linearly independent.

$A^T A$  is square  $n \times n$  and if  $A^T A x = 0$ , then

$x^T A^T A x = \|Ax\|^2 = 0$ , so  $Ax = 0$  and  $x = 0$ . Thus

in this case there is a unique solution.

How do we know that there is a vector  $v \in \Pi$  so that  $b - v \perp \Pi$ , and is  $v$  unique? Yes. For simplicity suppose the first  $k$  columns of  $A$ :  $A_1, \dots, A_k$  are a basis for  $\Pi$  (linearly independent spanning set).

Then  $A^T A x = A^T b$  has a unique solution and  $v = Ax$  satisfies  $A^T (v - b) = 0$ , so  $v - b$  is  $\perp$  to  $A_1, \dots, A_k$ , hence  $\perp$  to any vector in the linear space spanned by  $\{A_1, \dots, A_k\}$ , hence  $\perp$  to  $\Pi$ .

Let  $x_0$  be such an  $x$ . Then let  $f(x) = \|Ax - b\|^2$ .

$$\begin{aligned} f(x_0 + h) &= \|Ax_0 - b\|^2 + 2 \underbrace{(Ax_0 - b)^T}_{=0} Ah + \|Ah\|^2 \\ &= f(x_0) + \|Ah\|^2, \quad (\|Ah\|^2 = h^T A^T A h) \\ &\geq f(x_0) \quad \text{for all } h, \text{ so } f(x_0) \text{ is} \\ &\quad \text{the minimum} \end{aligned}$$