

Linear Least Squares in a Nutshell

Let A denote a given real $n \times m$ matrix where the dimension $m < n$. Let vectors be columns and use the superscript notation T to denote vector-matrix transposition. Let b denote a given real n -vector and suppose that it is desired to solve the equation

$$A \cdot x = b$$

for an unknown m -vector x . In general this may not be possible unless b satisfies a compatibility constraint to be derived below. So if the Euclidean norm $\|v\|$ of a vector v is defined as the positive square-root of $\|v\|^2 \equiv v^T \cdot v$ then one might seek to minimize $\|A \cdot x - b\|^2$ to get an approximate solution. This is most concisely understood in terms of the following formalism.

Let A^+ denote the *pseudo-inverse* of A , a unique $m \times n$ matrix which always exists and satisfies the following four conditions:

$$A \cdot A^+ \cdot A = A, \quad A^+ \cdot A \cdot A^+ = A^+, \quad (A \cdot A^+)^T = A \cdot A^+, \quad (A^+ \cdot A)^T = A^+ \cdot A.$$

Using the pseudo-inverse, one may define *symmetric* square matrices L and R , of dimensions n and m respectively, as

$$L = I_n - A \cdot A^+ \equiv L^T, \quad R = I_m - A^+ \cdot A \equiv R^T,$$

where I_k denotes the k -dimensional identity matrix, and readily verify that

$$L \cdot A = 0, \quad A \cdot R = 0,$$

so that L and R serve as left and right *annihilators* of A . It can be proved that

$$\min \|A \cdot x - b\|^2 = \|A \cdot x^{\wedge} - b\|^2 \equiv \|L \cdot b\|^2,$$

where by definition

$$x^{\wedge} = A^+ \cdot b + R \cdot c$$

and c is an arbitrary m -vector, and (because $R \cdot A^+ \equiv 0$) it is easy to verify that

$$\|x^{\wedge}\|^2 \equiv \|A^+ \cdot b\|^2 + \|R \cdot c\|^2.$$

Accordingly the *shortest* least-squares solution is given by choosing

$$R \cdot c = 0,$$

in which case the solution is *exact* if and only if

$$L \cdot b = 0.$$

Reference: Problems 17-20, pp. 270-271, Gilbert Strang & Kai Borre, *Linear Algebra, Geodesy, and GPS*, Wellesley-Cambridge Press, 1997.