

LINEAR ALGEBRA AND ITS APPLICATIONS UE19MA251

Projections And Least Squares



The failure of Gaussian Elimination is almost certain when we have several equations in one unknown.

$$a_1 x = b_1$$

$$a_2 x = b_2$$

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$$a_m x = b_m$$

This system is solvable if $b = (b_1, ..., b_m)$ is a multiple of $a = (a_1,, a_m)$.

Projections And Least Squares



If the system is inconsistent, then we choose that value of a that minimizes an average error E in the m equations. The most convenient average comes from the

sum of squares:

$$E^{2} = \sum_{i=1}^{m} (a_{i} x - b_{i})^{2}$$

If there is an exact solution the minimum error is E = 0. If not, the minimum error occurs when $\frac{dE^2}{dx}$ = 0

Solving for x, the least squares solution is $\hat{x} = \frac{a^T b}{a^T a}$

Least Squares Problem With Several Variables



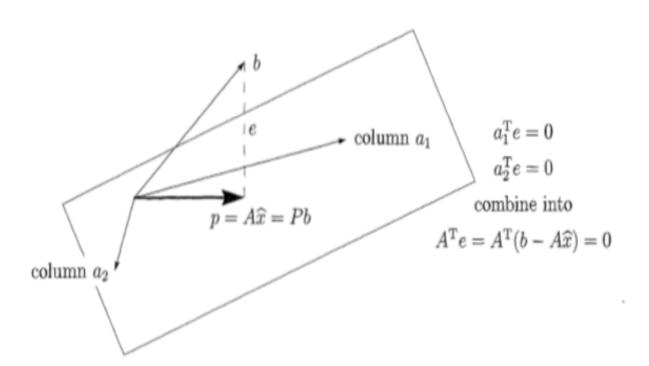
Consider a system of equations Ax = b that is inconsistent.

The vector b lies outside C(A) and we need to project it onto C(A) to get the point p in C(A) that is closest to b. The problem here is the same as to minimize the error E = ||Ax - b|| and this is exactly the distance from b to the point Ax in C(A).

Searching for the least squares solution \hat{x} is the same as locating the point p that is closest to b.

Least Squares Problem With Several Variables





Least Squares Problem With Several Variables



The error vector $e=b-A\hat{x}$ must be perpendicular to C(A) and hence can be found in the left null space of A.

Thus,
$$A^{T}(b-A\hat{x}) = 0 \text{ or } A^{T}A\hat{x} = A^{T}b$$

These are called the *Normal Equations*.

Solving them, we get the optimal solution \hat{x}

Note :

If b is orthogonal to C(A) then its projection is the zero vector.



THANK YOU