

1. Problem

we consider

$$y = Ax \quad (6)$$

where $A \in R^{m \times n}$ is fat ($m < n$), i. e.,

- there are more variables than equations
- x is underspecified, i. e., many choices of x lead to the same y

we'll assume that A is full rank(m), so for each $y \in R^m$, there is a solutions et of all solutions has form

$$\{x | Ax = y\} = \{x_p + z | Az = 0\} \quad (1)$$

where x_p is any ('particular') solution, i. e., $Ax_p = y$

How to prove that $x^* = A^T(AA^T)^{-1}y$ is the solution of $y = Ax$ that minimizes $\|x\|_2^2$?

2. Proof

$x^* = A^T(AA^T)^{-1}y$ is a solution because

$$Ax^* = AA^T(AA^T)^{-1}y = Iy = y \quad (2)$$

Then we consider an optimization problem

$$\begin{aligned} \min_x \quad & \|x\|_2^2 \\ \text{s. t.} \quad & Ax = b \end{aligned} \quad (3)$$

We have $A(x - x^*) = 0$ and $\forall x \in \{x | Ax = y\}$, so that

$$\begin{aligned} (x - x^*)^T x^* &= (x - x^*)^T A^T(AA^T)^{-1}y \\ &= (A(x - x^*))^T(AA^T)^{-1}y \\ &= 0 \end{aligned} \quad (4)$$

So $(x - x^*) \perp x^*$.

According to Triangular Inequality of 2 vertical vectors, we have

$$\|x\| = \|x^* + x - x^*\|_2^2 = \|x - x^*\|_2^2 + \|x^*\|_2^2 \geq \|x^*\|_2^2 \quad (5)$$

So $x^* = A^T(AA^T)^{-1}y$ is the solution of $y = Ax$ that minimizes $\|x\|_2^2$