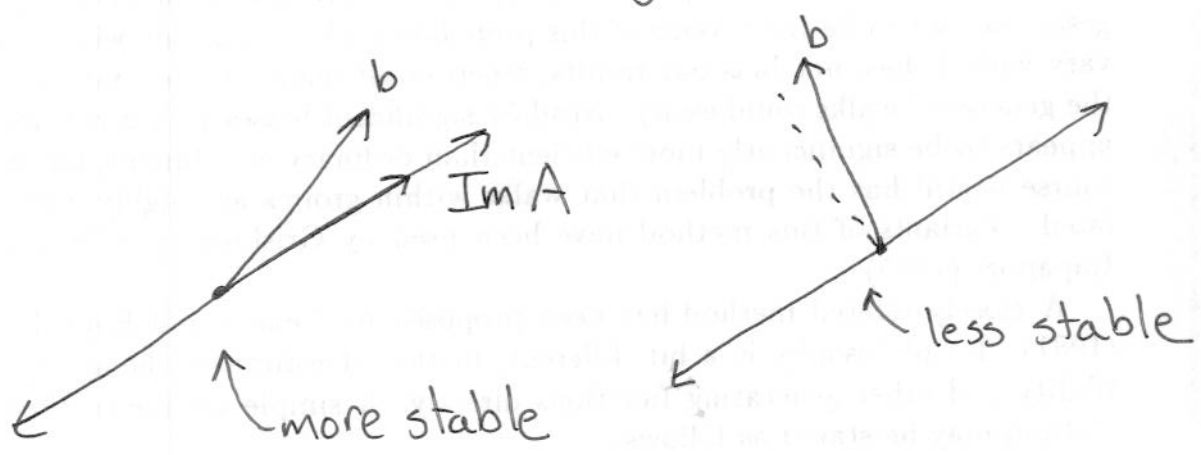


# Condition Number for Least Squares.

It is an interesting feature of least squares that the difficulty of solving a given problem may depend on the rhs  $b$  and not just on  $A$ .



We can still define a "condition number"

~~is~~ Definition. If  $A$  is  $m \times n$ ,  $m > n$ , we define

$$K_2(A) := \frac{\sigma_1(A)}{\sigma_n(A)},$$

the ratio of largest and smallest condition numbers.

Theorem. Suppose  $A$  is  $m \times n$  with  $m \geq n$  and full rank. If  $x$  minimizes  $\|Ax - b\|_2$ , then let  $r = Ax - b$  be the residual. Let  $\tilde{x}$  minimize  $\|(A + \delta A)\tilde{x} - (b + \delta b)\|_2$ . If  $\epsilon \equiv \max\left(\frac{\|\delta A\|_2}{\|A\|_2}, \frac{\|\delta b\|_2}{\|b\|_2}\right) < \frac{1}{\kappa_2(A)}$  then

$$\frac{\|\tilde{x} - x\|_2}{\|x\|_2} \leq \epsilon \cdot \left\{ \frac{2 \cdot \kappa_2(A)}{\cos \theta} + \tan \theta \kappa_2^2(A) \right\} + O(\epsilon^2)$$

where  $\sin \theta = \frac{\|r\|_2}{\|b\|_2}$  or  $\theta$  is the angle between  $b$  and  $Ax$ .

Proof. We know

$$\tilde{x} = \left( (A + \delta A)^T (A + \delta A) \right)^{-1} (A + \delta A)(b + \delta b).$$

Expand this in  $\delta A$  and  $\delta b$  and consider only linear terms.  $\square$

It turns out that there's a nicer form here for the error in the residual.

$$\frac{\|\tilde{r} - r\|_2}{\|r\|_2} \leq (1 + 2\epsilon \kappa_2(A)).$$