

# Announcements

- 1) HW due today,  
new one up tomorrow  
or later tonight

## Example 1:

$$A = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$A: \mathbb{C} \rightarrow \mathbb{C}^2$$

$$\text{rank}(A) = 1$$

$$\text{ran}(A) = \text{span} \left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right\}$$

= the line  $y=2x$ .

Use least-squares  
to "solve"

$$Ax = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

(  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  not in  $\text{ran}(A)$  )

$$\text{Let } \tilde{A} = \begin{bmatrix} 1.0001 \\ 2 \end{bmatrix}$$

$$\tilde{b} = \begin{bmatrix} .9999 \\ 0 \end{bmatrix}$$

We should expect the  
number  $\kappa(A) = \|A\|_2 \|A^+\|_2$   
to show up since that's  
what appeared when solving  
 $Ax = b$  with  $A \in \mathbb{C}^{m \times m}$ , invertible

First calculate  $K(A)$ .

We get  $K(A) = 1$ .

Let's first consider

the problem of

changing  $b$  to  $\tilde{b}$

and then solving

for  $x$  (and  $\tilde{x}$ ).

We saw that  
perturbing  $b$  to  $\tilde{b}$   
yielded relative  
errors less than  
the condition number  
of  $A$  when solving  
for either  $x$  or  $y = Ax$ .

Now we perturb  $A$  to

$\tilde{A}$  and investigate

what happens to solutions

$$x \text{ or } y = Ax.$$

We calculate

$$\frac{|x - \tilde{x}|}{|x|} \cdot \frac{\|A\|_2}{\|A - \tilde{A}\|_2}$$

We want this number

to be less than or equal to

$$\kappa(A) = 1.$$

Unfortunately,

we get

$$\frac{|x - \tilde{x}|}{|x|} \cdot \frac{\|A\|_2}{\|A - \tilde{A}\|_2}$$

$$\approx 1.341542 > 1.$$

So there is more to the conditioning of this problem than just  $K(A)$ .

For  $y$ , we get

$$\frac{\|y - \tilde{y}\|_2}{\|y\|_2} \cdot \frac{\|A\|_2}{\|A - \tilde{A}\|_2}$$

$\approx 2$ , which

is even worse!

In fact, even  $b$   
is deceptive. Our  
matrix was too  
simple to register  
any issues.

Theorem: The least-squares problem is well-conditioned, with the following condition numbers:

Solving for

Perturbing

y

x

b

$$\frac{1}{\cos \theta}$$

$$\frac{K(A)}{\eta \cos(\theta)}$$

A

$$\frac{K(A)}{\cos(\theta)}$$

$$K(A) + \frac{K(A)^2 \tan(\theta)}{\eta}$$

where

$$K(A) = \|A\|_2 \|A^+\|_2 = \frac{\sigma_1(A)}{\sigma_n(A)}$$

$$\cos \theta = \frac{\|y\|_2}{\|b\|_2}$$

$$\tilde{\kappa} = \frac{\|A\|_2 \|x\|_2}{\|y\|_2}$$

The numbers in the first row are exact and the numbers in the second row are upper bounds.

Why do these numbers show up?

Perturbing  $b$ , solving for  $x$ :

$$x = A^+ b$$

$$\tilde{x} = A^+ \tilde{b}$$

Look at

$$\frac{\|x - \tilde{x}\|_2}{\|x\|_2} = \frac{\|b\|_2}{\|b - \tilde{b}\|_2}$$

$$\frac{\|x - \tilde{x}\|_2}{\|x\|_2} = \frac{\|b\|_2}{\|b - \tilde{b}\|_2}$$

$$= \frac{\|A^+ b - A^+ \tilde{b}\|_2}{\|A^+ b\|_2} \cdot \frac{\|b\|_2}{\|b - \tilde{b}\|_2}$$

$$= \frac{\|A^+ (b - \tilde{b})\|_2}{\|A^+ b\|_2} \cdot \frac{\|b\|_2}{\|b - \tilde{b}\|_2}$$

$$\leq \frac{\|A^+\|_2 \cancel{\|b - \tilde{b}\|_2}}{\|A^+ b\|_2} \cdot \frac{\|b\|_2}{\cancel{\|b - \tilde{b}\|_2}}$$

(definition of  $\|A^+\|_2$ )

$$= \frac{\|A^+ \|_2 \|b\|_2}{\|A^+ b\|_2}$$

$$= \frac{\|A^+ \|_2 \|b\|_2}{\|A^+ b\|_2} \cdot \frac{\|y\|_2}{\|y\|_2}$$

$$= \frac{\|b\|_2}{\|y\|_2} \frac{\|A^+ \|_2 \|y\|_2}{\|A^+ b\|_2}$$

$$= \frac{\perp}{\cos \theta} \frac{\|A^+ \|_2 \|y\|_2}{\|A^+ b\|_2}$$

$\perp$   
 $\cos \theta$

$$\frac{\|A^{\dagger}\|_2 \|y\|_2}{\|A^{\dagger} b\|_2}$$

$$= \frac{1}{\cos(\theta)} \frac{\|A^{\dagger}\|_2 \|y\|_2}{\|x\|_2}$$

$$= \frac{1}{\cos(\theta)} \frac{\|A^{\dagger}\|_2 \|y\|_2}{\|x\|_2} \cdot \frac{\|A\|_2}{\|A\|_2}$$

$$= \frac{1}{\cos(\theta)} \|A^{\dagger}\|_2 \|A\|_2 \frac{\|y\|_2}{\|A\|_2 \|x\|_2}$$

$$= \frac{1}{\cos(\theta)} \kappa(A) \frac{1}{\rho} = \frac{\kappa(A)}{\rho \cos(\theta)} \quad \checkmark$$

Equality occurs  
when

$$\begin{aligned} & \|A^+ (b - \tilde{b})\|_2 \\ &= \|A^+\|_2 \|b - \tilde{b}\|_2, \end{aligned}$$

i.e.  $b - \tilde{b}$  is a maximal  
right singular vector for  
 $A^+$ .

Perturbing  $b$ , solving for  $y$ :

$$\begin{aligned} & \frac{\|y - \tilde{y}\|_2}{\|y\|_2} \cdot \frac{\|b\|_2}{\|b - \tilde{b}\|_2} \\ &= \frac{\|b\|_2}{\|y\|_2} \frac{\|y - \tilde{y}\|_2}{\|b - \tilde{b}\|_2} \\ &= \frac{1}{\cos(\theta)} \frac{\|y - \tilde{y}\|_2}{\|b - \tilde{b}\|_2} \end{aligned}$$

If  $P = AA^+$ , orthogonal  
projection onto  $\text{ran}(A)$ ,  
then  $y = Pb$ ,  $\tilde{y} = P\tilde{b}$ .

$$\begin{aligned}\text{So } \|y - \tilde{y}\|_2 &= \|Pb - P\tilde{b}\|_2 \\ &= \|P(b - \tilde{b})\|_2 \\ &\leq \|b - \tilde{b}\|_2.\end{aligned}$$

This gives

$$\frac{\|y - \tilde{y}\|_2}{\|y\|_2} \cdot \frac{\|b\|_2}{\|b - \tilde{b}\|_2}$$

$$= \frac{1}{\cos(\theta)} \cdot \frac{\|y - \tilde{y}\|_2}{\|b - \tilde{b}\|_2}$$

$$\leq \frac{1}{\cos(\theta)} \frac{\|b - \tilde{b}\|_2}{\|b - \tilde{b}\|_2}$$

$$= \frac{1}{\cos(\theta)}$$

You get equality when

$$b - \tilde{b} \in \text{ran}(A)$$

$$\Rightarrow P(b - \tilde{b}) = b - \tilde{b} .$$