

 1 Show that  $\kappa(I) = 1$  ;

**Solution:** This is obvious because for any matrix norm  $\|I\| = \|I^{-1}\| = 1$ .  $\square$

 2 Show that  $\kappa(A) \geq 1$  ;

**Solution:** We have  $\|AA^{-1}\| = \|I\| = 1$  therefore  $1 = \|AA^{-1}\| \leq \|A\| \|A^{-1}\| = \kappa(A)$   $\square$

 5 Show that if  $\|E\|/\|A\| \leq \delta$  and  $\|e_b\|/\|b\| \leq \delta$  then

$$\frac{\|x - y\|}{\|x\|} \leq \frac{2\delta\kappa(A)}{1 - \delta\kappa(A)}$$


**Solution:** From the main theorem (theorem 1) we have

$$\frac{\|x - y\|}{\|x\|} \leq \frac{\|A^{-1}\| \|A\|}{1 - \|A^{-1}\| \|E\|} \left( \frac{\|E\|}{\|A\|} + \frac{\|e_b\|}{\|b\|} \right)$$

If  $\|E\| \leq \delta$  and  $\|e_b\|/\|b\| \leq \delta$  then:

$$\begin{aligned} \frac{\|x - y\|}{\|x\|} &\leq \frac{\kappa(A) \times 2\delta}{1 - \|A^{-1}\| \|E\|} \\ &\leq \frac{2\delta\kappa(A)}{1 - \|A^{-1}\| \|A\| \times (\|E\|/\|A\|)} \\ &\leq \frac{2\delta\kappa(A)}{1 - \delta\kappa(A)}. \end{aligned}$$



 9 Show that  $\frac{\|x - \tilde{x}\|}{\|x\|} \geq \frac{1}{\kappa(A)} \frac{\|r\|}{\|b\|}$ .

**Solution:** We start with noting that  $A(x - \tilde{x}) = b - A\tilde{x} = r$ . So:

$$\|r\| \leq \|A\| \|x - \tilde{x}\| \rightarrow \frac{\|r\|}{\|b\|} \leq \|A\| \frac{\|x - \tilde{x}\|}{\|b\|}$$

Next from  $\|x\| = \|A^{-1}b\| \leq \|A^{-1}\| \|b\|$  we get  $\|b\| \geq \|x\|/\|A^{-1}\|$

and so

$$\frac{\|r\|}{\|b\|} \leq \|A\| \frac{\|x - \tilde{x}\|}{\|x\|/\|A^{-1}\|} = \kappa(A) \frac{\|x - \tilde{x}\|}{\|x\|}$$

which yields the result after dividing the 2 sides by  $\kappa(A)$ . 

## More on normwise backward error

- We solve  $Ax = b$  and find an approximate solution  $y$

**Question:** Find smallest perturbation to apply to  $A, b$  so that \*exact\* solution of perturbed system is  $y$

- We now consider the situation where only  $b$  or  $A$  are perturbed (not both)

Suppose we model entire perturbation in RHS  $b$ .

- Let  $r = b - Ay$  be the residual. Then  $y$  satisfies  $Ay = b + \Delta b$  with  $\Delta b = -r$  exactly.
- The relative perturbation to the RHS is  $\frac{\|r\|}{\|b\|}$ .

Suppose we model entire perturbation in matrix  $A$ .

- Then  $y$  satisfies  $\left(A + \frac{ry^T}{y^Ty}\right)y = b$
- The relative perturbation to the matrix is

$$\left\| \frac{ry^T}{y^Ty} \right\|_2 / \|A\|_2 = \frac{\|r\|_2}{\|A\| \|y\|_2}$$

## Proof of Theorem 3

Let  $D \equiv \|E\|\|y\| + \|e_b\|$  and  $\eta \equiv \eta_{E,e_b}(y)$ . The theorem states that  $\eta = \|r\|/D$  (recall that  $r = b - Ay$ ). Proof in 2 steps.

**First:** Any  $\Delta A, \Delta b$  pair satisfying (1) is such that  $\epsilon \geq \|r\|/D$ . Indeed from (1) we have:

$$\begin{aligned} Ay + \Delta Ay &= b + \Delta b \rightarrow r = \Delta Ay - \Delta b \rightarrow \\ \|r\| &\leq \|\Delta A\|\|y\| + \|\Delta b\| \\ &\leq \epsilon(\|E\|\|y\| + \|e_b\|) \rightarrow \\ \epsilon &\geq \frac{\|r\|}{D} \end{aligned}$$

**Second:** We need to show an instance where the minimum value of  $\|r\|/D$  is reached. Take the pair  $\Delta A, \Delta b$ :

$$\Delta A = \alpha r z^T; \Delta b = \beta r \text{ with } \alpha = \frac{\|E\|\|y\|}{D}; \beta = -\frac{\|e_b\|}{D}$$

The vector  $z$  depends on the norm used - for the 2-norm:  $z = y/\|y\|^2$ . **Here: Proof only for 2-norm**

Next, we need to verify that first part of (1) is satisfied:

$$\begin{aligned}
 (A + \Delta A)y &= Ay + \alpha r \frac{y^T}{\|y\|^2} y = b - r + \alpha r \\
 &= b - (1 - \alpha)r \\
 &= b - \left(1 - \frac{\|E\|\|y\|}{\|E\|\|y\| + \|e_b\|}\right) r \\
 &= b - \frac{\|e_b\|}{D} r = b + \beta r \quad \rightarrow \\
 (A + \Delta A)y &= b + \Delta b \quad \leftarrow \text{The desired result}
 \end{aligned}$$

**Finally:** Must now verify that  $\|\Delta A\| = \eta\|E\|$  and  $\|\Delta b\| = \eta\|e_b\|$ . **Exercise:** Show that  $\|uv^T\|_2 = \|u\|_2\|v\|_2$

$$\|\Delta A\| = \frac{|\alpha|}{\|y\|^2} \|ry^T\| = \frac{\|E\|\|y\| \|r\|\|y\|}{D \|y\|^2} = \eta\|E\|$$

$$\|\Delta b\| = |\beta|\|r\| = \frac{\|e_b\|}{D} \|r\| = \eta\|e_b\| \quad \text{QED}$$