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Problem with repeated evals

$$A(\varepsilon) = \begin{pmatrix} \lambda & \varepsilon \\ 0 & \lambda \end{pmatrix}$$

$$\begin{aligned} p_\varepsilon(z) &= \det(A(\varepsilon) - zI) = (\lambda - z)^2 - \varepsilon \\ &= z^2 - 2\lambda z + \lambda^2 - \varepsilon \\ &= (z - (\lambda + \sqrt{\varepsilon})) (z - (\lambda - \sqrt{\varepsilon})) \end{aligned}$$

evals $\approx \lambda \pm \sqrt{\varepsilon}$ continuous func of ε
but not diff

$\varepsilon \rightarrow 0$ $A(0) = \begin{pmatrix} \lambda & \varepsilon \\ 0 & \lambda \end{pmatrix}$ Jordan block

$O(\varepsilon)$ perturbation $\Rightarrow \sqrt{\varepsilon}$ change cond number $= \infty$

$$s \left\{ \begin{pmatrix} \lambda & \varepsilon & \dots & \varepsilon \\ \vdots & \vdots & \ddots & \vdots \\ \varepsilon & \dots & \lambda & \varepsilon \end{pmatrix} \right. \text{evals} = z + O(\varepsilon^{1/s})$$

Better perturbation theory for unsym. matrices

$$Ax = \lambda x, \lambda \text{ simple, } \|x\|_2 = 1 \text{ (w.r.t } \|\cdot\| \text{ over } \mathbb{C})$$

$$S^{-1}AS = \text{diag}(J_1, \dots, J_k) \quad J_i = \begin{cases} \left(\begin{array}{ccc} \lambda_i & & 0 \\ & \ddots & \\ 0 & & \lambda_i \end{array} \right) & J_i e_1 = \lambda_i e_1 \end{cases}$$

$$AS = S \text{diag}(J_1, \dots, J_k) \quad e_s^T J_i = \lambda_i e_s^T \quad e_s^T e_1 = 0$$

$$S^{-1}A = \text{diag}(J_1, \dots, J_k) S^{-1}$$

Left evecs: $y^H A = \lambda y^H, \|y\|_2 = 1$ (w.r.t $\|\cdot\|$ over \mathbb{C})

$S^{-1}AS = \Lambda \Rightarrow$ right evecs given by cols of S
left evecs given by rows of S^{-1}

$$(A + \delta A)(x + \delta x) = (\lambda + \delta \lambda)(x + \delta x)$$

$$Ax = \lambda x$$

$$y^H (A \delta x + \delta A x) = y^H (\lambda \delta x + \delta \lambda x) + \text{higher-order terms}$$

$$y^H A \delta x + y^H \delta A x = \lambda y^H \delta x + \delta \lambda y^H x$$

$$y^H \delta A x = \delta \lambda y^H x$$

$$\delta \lambda = \frac{y^H \delta A x}{y^H x} + \text{higher-order}$$

$$|\delta \lambda| \leq \frac{\|\delta A\|_2}{\|y^H x\|} + O(\|\delta A\|^2)$$

① Symm $y^H = x^T$

② \rightarrow Jordan block (nontrivial)
 $\rightarrow \infty$

Last time: Bauer-Fike $X^{-1}AX = \Lambda \quad \mu \in \lambda(A+E)$

$$\min_j |\mu - \lambda_j| \leq \|X\|_p \|X^{-1}\|_p \|E\|_p$$

Evals of $A+E \in \bigcup_j D_{\lambda_j}(\|X\| \|X^{-1}\| \|E\|) \quad \|X^{-1}EX\|_\infty$

Better: $\in \bigcup_j D_{\lambda_j}(\|E\|_2 / |y_j^H x_j|)$

Gershgorin's: evals of $M \in \bigcup_j D_{m_{jj}}(\sum_{i \neq j} |M_{ij}|) \quad \text{C-S}$

$$X^{-1}(A+E)X = \Lambda + \underbrace{X^{-1}EX}_F \quad |\mu - \lambda_j| \leq \sum_{i \neq j} |F_{ji}| \leq \sqrt{n} \|F(j, :)\|_2$$

$$\|F(j, :)\|_2 = \|e_j^T X^{-1} E X\|_2 \leq \|e_j^T X^{-1}\|_2 \|E\|_2 \|X\|_2 \sqrt{n}$$

choose X so that cols have unit norm $\Rightarrow \|X\|_2 \leq \sqrt{n}$

rows of X^{-1} are left evecs $\Rightarrow X^{-1} = \begin{bmatrix} e_1 y_1^H \\ \vdots \\ e_n y_n^H \end{bmatrix} \quad \|y_j\|_2 = 1$

$$(X^{-1}X) = I$$

$$I = (X^{-1}X)_{jj} = c_j y_j^H x_j \Rightarrow c_j = \frac{1}{y_j^H x_j} \Rightarrow \|e_j^T X^{-1}\|_2 = \frac{1}{|y_j^H x_j|}$$

$$A = [Q_1 \ Q_2] \begin{bmatrix} \Lambda_1 & 0 \\ 0 & \Lambda_2 \end{bmatrix} \begin{bmatrix} Q_1^T \\ Q_2^T \end{bmatrix}$$

E symmetric

$$A+E = [\tilde{Q}_1 \ \tilde{Q}_2] \begin{bmatrix} \tilde{\Lambda}_1 & 0 \\ 0 & \tilde{\Lambda}_2 \end{bmatrix} \begin{bmatrix} \tilde{Q}_1^T \\ \tilde{Q}_2^T \end{bmatrix}$$

$$\Lambda_1 \subseteq [a, b]$$

$$\delta = \min_k \{ |x - (\tilde{\Lambda}_2)_{kk}| : x \in [a, b] \}$$

$$\| \tilde{Q}_2^T Q_1 \| \leq \| \tilde{Q}_2^T E Q_1 \| / \delta$$

$$= d(\text{range}(Q_1), \text{range}(\tilde{Q}_1)) \leq \|E\| / \delta$$

$$= \|\sin \theta\| \quad \tilde{Q}_2^T \perp \text{range}(Q_1)$$

Proof:

$$A Q_1 = Q_1 \Lambda_1$$

$$\begin{aligned} \underline{E} Q_1 &= (A+E-A) Q_1 \\ &= (A+E) Q_1 - Q_1 \Lambda_1 \end{aligned}$$

$$\tilde{Q}_2^T (A+E) = \tilde{\Lambda}_2 \tilde{Q}_2^T$$

$$\tilde{Q}_2^T E Q_1 = \tilde{Q}_2^T [(A+E) Q_1 - \Lambda_1 Q_1]$$

$$= \tilde{\Lambda}_2 \tilde{Q}_2^T Q_1 - \tilde{Q}_2^T Q_1 \Lambda_1$$

$$= \tilde{\Lambda}_2 M - M \Lambda_1$$

$$\| \tilde{\Lambda}_2 M - M \Lambda_1 \|$$

$$= \| (\tilde{\Lambda}_2 - cI) M - M (\Lambda_1 - cI) \|$$

$$\geq \| (\tilde{\Lambda}_2 - cI) M \| - \| M (\Lambda_1 - cI) \|$$

$$c = (a+b)/2$$



$$\| (\tilde{\Lambda}_2 - cI) M \| \geq \left(\min_k |(\tilde{\Lambda}_2)_{kk} - c| \right) \|M\|$$

$$\geq (r+s) \|M\|$$

$$\leq \|M\| \| \Lambda_1 - cI \| \leq \|M\| r$$

$$\| \tilde{Q}_2^T E Q_1 \| \geq (r+s) \|M\| - r \|M\|$$

$$= \delta \| \tilde{Q}_2^T Q_1 \|$$