

22.2

$$\text{Sec } p_r(A) := \sum_{j=0}^p a_j A^j.$$

$$\begin{aligned} \Rightarrow \text{For eigens } v_i \text{ of } A, \text{ w.r.t to eigenval } \lambda_i, \quad p_r(A)v_i &= \sum_{j=0}^p a_j A^j v_i \\ &= \sum_{j=0}^p a_j \lambda_i^j v_i \quad (\because A^j v_i = \lambda_i A^{j-1} v_i = \dots = \lambda_i^j v_i) \\ &= p_r(\lambda_i) v_i \end{aligned}$$

$$\therefore p_r(A)w = \sum_{i=1}^n c_i p_r(A) v_i = \sum_{i=1}^n c_i p_r(\lambda_i) v_i = \sum_{i=1}^k c_i p_r(\lambda_i) v_i + \sum_{i=k+1}^n c_i p_r(\lambda_i) v_i$$

22.3.

$$(A - \lambda_{u_2}^{(m)}) V_m^{(1)} = V_m^{(1)} (H_m^{(1)} - \lambda_{u_2}^{(m)} I) + h_{m+1,m} V_{m+1} (b_{m+1}^{(1)})^T$$

$$(A - \lambda_{u_2}^{(m)}) V_m^{(1)} Q_2 = [V_m^{(1)} (H_m^{(1)} - \lambda_{u_2}^{(m)} I) + h_{m+1,m} V_{m+1} (b_{m+1}^{(1)})^T] Q_2$$

$$H_m^{(1)} = Q_2 R_2 + \lambda_{u_2}^{(m)} I$$

↓

$$(A - \lambda_{u_2}^{(m)}) V_m^{(1)} Q_2 = [V_m^{(1)} (Q_2 R_2 + \cancel{\lambda_{u_2}^{(m)} I} - \cancel{\lambda_{u_2}^{(m)} I}) + h_{m+1,m} V_{m+1} (b_{m+1}^{(1)})^T] Q_2$$

$$\therefore A V_m^{(1)} Q_2 = \lambda_{u_2}^{(m)} V_m^{(1)} Q_2 + [V_m^{(1)} (Q_2 R_2) + h_{m+1,m} V_{m+1} (b_{m+1}^{(1)})^T] Q_2$$

$$\text{Setting } V_m^{(2)} = V_m^{(1)} Q_2, \quad H_m^{(2)} = R_2 Q_2 + \lambda_{u_2}^{(m)} I, \quad b_{m+1}^{(2)} = Q_2^T \cdot b_{m+1}^{(1)}$$

$$A V_m^{(2)} = V_m^{(2)} H_m^{(2)} + h_{m+1,m} V_{m+1} (b_{m+1}^{(2)})^T \quad \square$$

22.5.

$$i) (A + \epsilon I) v = (I + \lambda v - \gamma v^T v) v = \lambda v + (I - \gamma) v = \lambda v.$$

$$E = -\gamma v^T \Rightarrow \|E\|_2 = ?$$

$$\Rightarrow E^T E = \gamma^T \gamma v v^T = \underline{(v^T v) \cdot v v^T}$$

$\Rightarrow VV^T$  is max of rank 1  $\Rightarrow I_n$  has only one non-zero eigenvalue.

(If not,  $\dim(\text{Im}(VV^T)) \geq 2$  (\*) )

$\Rightarrow$  For  $v$  is trivial eigenvector of  $VV^T \Rightarrow VV^T \cdot v = v \cdot 1$

$\therefore$  non-zero eigenvalue of  $VV^T$  is 1.

$\therefore$  non-zero eigenvalue of  $V^T \cdot V$  is  $\varepsilon^2$ .

$$\Rightarrow \|E\|_2 = \sqrt{\varepsilon^2} = \varepsilon$$

ii) If the norm of residual is small,  $\|E\|_2$  : small

$\therefore (\lambda, u)$  : exact eigenvector of the matrix  $A+E$ , close to  $A$ .

iii)

$$\begin{aligned} \text{eigs}(A+E) &= (\lambda, u) \\ \text{eigs}(A) &= (\lambda_0, v_0) \end{aligned} \quad \begin{array}{l} \text{Approximated} \\ \text{eigenvalue} \\ \text{if } \|E\|_2 \text{ are small} \end{array}$$

$\Rightarrow$  good approximate to backward error

