

Laczos, Rayleigh quotient $\Rightarrow \lambda_1, \lambda_n$ of A symmetric

Theorem

Let $A \in \mathbb{R}^{n \times n}$ symmetric and $U \in \mathbb{R}^{n \times n-1}$ such that $U^T U = I_{n-1}$

Let $B = U^T A U \in \mathbb{R}^{n-1 \times n-1}$.

Then the eigenvalues $\lambda_1, \dots, \lambda_n$ of A and

the eigenvalues μ_1, \dots, μ_{n-1} of B

$$\lambda_n \leq \mu_{n-1} \leq \lambda_{n-1} \leq \dots \leq \mu_2 \leq \lambda_2 \leq \mu_1 \leq \lambda_1$$

$$U_1 \in \mathbb{R}^{n \times 1}$$

$$B_1 = U_1^T A U_1$$

$$\mu_1^{(1)}$$

$$\lambda_n \leq \mu_1^{(1)} \leq \lambda_1$$

$$U_2 \in \mathbb{R}^{n \times 2}$$

$$B_2 = U_2^T A U_2$$

$$\lambda_n < \mu_2^{(2)} < \mu_1^{(2)} < \mu_1^{(2)} < \lambda_1$$

$$U_3 \in \mathbb{R}^{n \times 3}$$

$$B_3 = U_3^T A U_3$$

$$\mu_3^{(3)} < \mu_2^{(2)} <$$

$$\mu_1^{(2)} < \mu_1^{(3)}$$

$$U_4 \in \mathbb{R}^{n \times 4}$$

$$B_4 = U_4^T A U_4$$

$$\mu_4^{(4)} < \mu_3^{(3)} <$$

$$< \mu_1^{(3)} < \mu_1^{(4)}$$

$$U_n \in \mathbb{R}^{n \times n-1}$$

$$U_n \in \mathbb{R}^{n \times n}$$

$$B_n = U_n^T A U_n$$

$$\lambda_n < \mu_{n-1}^{(n-1)} <$$

$$\mu_1^{(n-2)} < \lambda_1$$

B_n similar to A

$$\mu_n^{(n)}$$

$$\mu_1^{(n)}$$

Corollary

Let $A \in \mathbb{R}^{n \times n}$

symmetric

$$\lambda_1 = \max_{\substack{x \neq 0 \\ x \in \mathbb{R}^n}} R_2(x) = \max_{\substack{x \neq 0 \\ x \in \mathbb{R}^n}} \frac{x^T A x}{x^T x}$$

$$\lambda_n = \min_{\substack{x \neq 0 \\ x \in \mathbb{R}^n}} R_2(x) = \min_{\substack{x \neq 0 \\ x \in \mathbb{R}^n}} \frac{x^T A x}{x^T x}$$

Moreover, the functional $R_2(x) : \mathbb{R}^n \rightarrow \mathbb{R}$ is stationary in v if and only if $Av = \lambda v$ $\lambda = R_2(v)$

$$R_2 = \frac{x^T A x}{x^T x}$$

$$\frac{\partial x}{\partial x_i} = e_i$$

$$\begin{aligned} (\nabla R_2)_i(x) &= \frac{\partial R_2(x)}{\partial x_i} = \left(\frac{\partial (x^T A x)}{\partial x_i} \cdot \frac{x^T x}{(x^T x)^2} - \frac{\partial (x^T x)}{\partial x_i} \frac{x^T A x}{(x^T x)^2} \right) \\ &= \left((e_i^T A x + x^T A e_i) \frac{x^T x}{(x^T x)^2} - (e_i^T x + x^T e_i) \frac{x^T A x}{(x^T x)^2} \right) \\ &= 2 e_i^T A x \frac{x^T x}{(x^T x)^2} - 2 e_i^T x \frac{x^T A x}{(x^T x)^2} \\ &= \frac{2 e_i^T}{x^T x} \left(A x - \frac{x^T A x}{x^T x} x \right) = \frac{2 e_i^T}{x^T x} (A x - R_2(x) x) \\ &= \frac{2}{x^T x} e_i^T [A x - R_2(x) x] \end{aligned}$$

$$\nabla R_2(x) = \frac{2}{x^T x} [A x - R_2(x) x]$$

$$\boxed{\nabla R_2(x) \in \text{span}(x, Ax)}$$

$$\begin{aligned} x_1 \quad R_2(x_1) &\nearrow x_2^\uparrow = x_1 + \delta_1^\uparrow \nabla R_2(x_1) \quad R_2(x_2^\uparrow) > R_2(x_1) \quad \nabla R_2(x_1) \in \text{span}(x_1, Ax_1) \\ &\searrow x_2^\downarrow = x_1 - \delta_1^\downarrow \nabla R_2(x_1) \quad R_2(x_2^\downarrow) < R_2(x_1) \\ &x_2^\uparrow, x_2^\downarrow \in \text{span}(x_1, \nabla R_2(x_1)) \end{aligned}$$

$$\begin{aligned} x_2^\uparrow \quad R_2(x_2^\uparrow) &\Rightarrow x_3^\uparrow = x_2^\uparrow + \delta_2^\uparrow \nabla R_2(x_2^\uparrow) \quad \nabla R_2(x_2^\uparrow) \in \text{span}(x_2^\uparrow, Ax_2^\uparrow) \\ &\nabla R_2(x_2^\uparrow) \in \text{span}(x_1, Ax_1, A^2 x_1) \\ &x_3^\uparrow \in \text{span}(x_1, Ax_1, A^2 x_1) \end{aligned}$$

$$\begin{aligned} x_2^\downarrow \quad R_2(x_2^\downarrow) &\Rightarrow x_3^\downarrow = x_2^\downarrow - \delta_2^\downarrow \nabla R_2(x_2^\downarrow) \\ &\nabla R_2(x_2^\downarrow) \in \text{span}(x_1, Ax_1, A^2 x_1) \\ &x_3^\downarrow \in \text{span}(x_1, Ax_1, A^2 x_1) \end{aligned}$$

$$U_1 \rightarrow V_1$$

$$U_2 \rightarrow V_2$$

$$V_k^T A V_k = T_k = \begin{bmatrix} \alpha_1 & \beta_2 & & \\ \beta_2 & \alpha_2 & \beta_3 & \\ & \beta_3 & \ddots & \\ & & & \alpha_n \end{bmatrix}$$

$$A V_k = V_k T_k$$

$$A V_j = \beta_j V_{j-1} + \alpha_j V_j + \beta_{j+1} V_{j+1}$$

$$V_j^T A V_j = \beta_j \underbrace{V_j^T V_{j-1}}_0 + \alpha_j \underbrace{V_j^T V_j}_1 + \beta_{j+1} \underbrace{V_j^T V_{j+1}}_0$$

$$\beta_{j+1} v_{j+1} = -\beta_j v_{j-1} + (A - \alpha_j I) v_j$$

$$w = -\beta_j v_{j-1} + (A - \alpha_j I) v_j \quad \neq 0$$

$$\beta_{j+1} = \|w\|$$

$$v_{j+1} = w / \beta_{j+1}$$

Symmetric Lanczos Algorithm

choose v_1 , $\|v_1\|=1$, $\beta_1=0$, $v_0=0$

for $j=1, \dots, k$

$$w = A v_j - \beta_j v_{j-1}$$

$$\alpha_j = (v_j, w)$$

$$w = w - \alpha_j v_j \quad (w = -\beta_j v_{j-1} + A v_j - \alpha_j v_j)$$

$$\beta_{j+1} = \|w\|_2$$

If $\beta_{j+1}=0$ invariant space

$$\text{else } v_{j+1} = w / \beta_{j+1}$$

end

T_n power method $T_n \rightarrow \lambda_1$

inverse power method $T_n \rightarrow \lambda_n$

$$T_n = L_n U_n$$

QR - RQ method

$$A \in \mathbb{R}^{n \times n}$$

$$A = A_1 = Q_1 R_1$$

$$A_2 = R_1 Q_1 = Q_1^T Q_1 R_1 Q_1 = Q_1^T (Q_1 R_1) Q_1 = Q_1^T A_1 Q_1 =$$

A_2 is similar to the matrix A_1

$$A_2 = Q_2 R_2$$

$$A_3 = R_2 Q_2 = Q_2^T Q_2 R_2 Q_2 = Q_2^T A_2 Q_2 = \underbrace{Q_2^T Q_1^T}_{Q^T} A_1 \underbrace{Q_1 Q_2}_Q$$

...

Theorem (Convergence Theorem)

Let $A \in \mathbb{R}^{n \times n}$ such that its eigenvalues have distinct modulus

$$0 < |\lambda_n| < |\lambda_{n-1}| < \dots < |\lambda_2| < |\lambda_1|$$

Let X the matrix containing the corresponding eigenvectors

$$A = X D X^{-1}$$

D : diagonal matrix with λ_j

on the diagonal

If X^{-1} admits a LU factorization then the matrices

$$A_{k+1} = Q_k^T A Q_k$$

($k \rightarrow \infty$)

tend to an upper triangular matrix whose diagonal elements are the eigenvalues of the matrix A in decreasing order

X^{-1} can not be factorized LU we lose the decreasing order of the eigenvalues

$$|\lambda_2| = |\lambda_3|$$

$A_{k+1} \rightarrow$

$$\begin{bmatrix} x & & & & \\ 0 & x & x & & \\ 0 & & x & x & \\ 0 & 0 & 0 & x & \\ 0 & 0 & 0 & 0 & x \\ 0 & 0 & 0 & 0 & 0 & x \end{bmatrix}$$

$$\begin{bmatrix} x & & & & \\ 0 & \boxed{} & & & \\ & 0 & x & & \\ & & 0 & \boxed{} & \\ & & & 0 & x \end{bmatrix}$$

$A \rightarrow$ Tridiagonal matrix

A symmetric

$$A^{(1)} = A = \begin{bmatrix} a_{11}^{(1)} & a_1^T \\ a_1 & B^{(1)} \end{bmatrix}$$

$$a_1 \in \mathbb{R}^{n-2 \times 1}$$

$$B^{(1)} \in \mathbb{R}^{n-1 \times n-1}$$

$$e_1 \in \mathbb{R}^{n-2 \times 1}$$

$$u_1 \in \mathbb{R}^{n-1 \times 1} \quad H_1 a_1 = \alpha_1 e_1$$

$$Q_1 = \begin{bmatrix} 1 & 0 \\ 0 & H_1 \end{bmatrix} \in \mathbb{R}^{n \times n}$$

$$A^{(2)} = Q_1^T A^{(1)} Q_1 = \begin{bmatrix} 1 & 0 \\ 0 & H_1^T \end{bmatrix} \begin{bmatrix} a_{11}^{(1)} & a_1^T \\ a_1 & B^{(1)} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & H_1 \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}^{(1)} & a_1^T \\ H_1^T a_1 & H_1^T B^{(1)} H_1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & H_1 \end{bmatrix} = \begin{bmatrix} a_{11}^{(1)} & a_1^T H_1 \\ H_1^T a_1 & H_1^T B^{(1)} H_1 \end{bmatrix}$$

$$\begin{aligned} H_1^T &= H_1 \\ a_1^T H_1 &= (H_1^T a_1)^T \end{aligned} \quad = \left[\begin{array}{c|ccc} a_{11}^{(2)} & \alpha_1 & \dots & 0 \dots 0 \\ \hline \alpha_1 & & & \\ \vdots & & H_1^T B^{(1)} H_1 & \\ 0 & & & \end{array} \right] = \left[\begin{array}{c|cc} a_{11}^{(2)} & a_{12}^{(2)} & 0^T \\ \hline a_{12}^{(2)} & a_{22}^{(2)} & a_2^T \\ \hline 0 & a_2 & B^{(2)} \end{array} \right] \quad \begin{aligned} 0 &\in \mathbb{R}^{n-2 \times 1} \\ a_2 &\in \mathbb{R}^{n-2 \times 1} \end{aligned}$$

$$u_2 \in \mathbb{R}^{n-2 \times 1} \quad H_2 a_2 = \alpha_2 e_1 \quad e_1 \in \mathbb{R}^{n-2 \times 1}$$

$$Q_2 = \begin{bmatrix} I & 0 \\ 0 & H_2 \end{bmatrix}$$

$$I \in \mathbb{R}^{2 \times 2}$$

$$0 \in \mathbb{R}^{n-2 \times 2}$$

$$H_2 \in \mathbb{R}^{n-2 \times n-2}$$

$$A^{(3)} = Q_2^T A^{(2)} Q_2 = \begin{bmatrix} a_{11}^{(2)} & a_{12}^{(2)} & 0^T \\ a_{12}^{(2)} & a_{22}^{(2)} & a_2^T \\ 0 & H_2^T a_2 & H_2^T B^{(2)} \end{bmatrix} \begin{bmatrix} I & 0 \\ 0 & H_2 \end{bmatrix}$$

$$H_2^T a_2 = H_2 a_2 = \alpha_2 e_1$$

$$= \begin{bmatrix} a_{11}^{(2)} & a_{12}^{(2)} & 0^T \\ a_{12}^{(2)} & a_{22}^{(2)} & a_2^T H_2 \\ 0 & H_2^T a_2 & H_2^T B^{(2)} H_2 \end{bmatrix} = \begin{bmatrix} T^{(3)} & 0^T \\ 0 & a_3 & B^{(3)} \end{bmatrix} \quad 0 \in \mathbb{R}^{n-3 \times 2}$$

$$T^{(3)} = \begin{bmatrix} a_{11}^{(2)} & a_{12}^{(2)} & 0 \\ a_{12}^{(2)} & a_{22}^{(2)} & \alpha_2 \\ 0 & \alpha_2 & x \end{bmatrix}$$

After n steps $T^{(n)}$ is a tridiagonal matrix

$$k: \quad U_k \in \mathbb{R}^{n-k \times 1} \Rightarrow H_k \in \mathbb{R}^{n-k \times n-k}$$

$$Q_k \in \mathbb{R}^{n \times n}$$

$$Q_k = \left[\begin{array}{c|c} I & 0^T \\ \hline 0 & H_k \end{array} \right]$$

$$O \in \mathbb{R}^{n-k \times k}$$

$$A^{(k+1)} = Q_k^T A^{(k)} Q_k = \left[\begin{array}{c|c} T^{(k)} & 0^T \\ \hline 0 & a_k \end{array} \right] \begin{array}{c} O^T \\ a_k^T \\ B^{(k+1)} \end{array}$$

$$T^{(k)} \in \mathbb{R}^{k \times k} \text{ tridiagonal}$$

$$a_k \in \mathbb{R}^{n-k-1 \times 1}$$

$$B^{(k+1)} \in \mathbb{R}^{n-k-1 \times n-k-1}$$

$$A^{(n)} = T^{(n)} \text{ tridiagonal similar to } A$$

$$U_k \quad k=1, \dots, n-1$$

Q can be "stored" in a triangular matrix

A generic matrix (non-symmetric)

$$A^{(1)} = A = \left[\begin{array}{c|c} a_{11} & a_1^{*T} \\ \hline a_1 & B^{(1)} \end{array} \right]$$

$$a_1 \in \mathbb{R}^{n-1 \times 1}$$

$$a_1^* \in \mathbb{R}^{n-1 \times 1} \quad a_1^* \neq a_1$$

$$H^{(1)}:$$

$$H_1 a_1 = \alpha_1 e_1$$

$$Q_1 = \left[\begin{array}{c|c} 1 & 0 \\ \hline 0 & H_1 \end{array} \right]$$

$$A^{(2)} = Q_1^T A^{(1)} Q_1 = \left[\begin{array}{c|c} a_{11}^{(2)} & a_1^{(2)*T} \\ \hline H_1 a_1 & H_1^T B^{(1)} H_1 \end{array} \right] = \left[\begin{array}{c|c} a_{11}^{(2)} & a_1^{(2)*T} H_1 \\ \hline 0 & H_1^T B^{(1)} H_1 \end{array} \right]$$

$$\uparrow$$

$$\alpha_1 e_1$$

$$a_1^{(2)*T} H_1 = (H_1^T a_1^*)^T = (H_1 a_1^*)^T \neq (\alpha_1 e_1)^T$$

$$A^{(2)} = \left[\begin{array}{c|c} a_{11}^{(2)} & a_{12}^{(2)} \\ \hline a_{21}^{(2)} & a_{22}^{(2)} \end{array} \right] \begin{array}{c} a_1^{(2)*T} \\ a_2^{(2)*T} \\ B^{(2)} \end{array}$$

$$[a_{12}^{(2)} \quad a_1^{(2)*T}] = a_1^{(2)*T} H_1$$

$$H_2 a_2 = \alpha_2 e_1$$

$$e_1 \in \mathbb{R}^{n-2 \times 1}$$

$$Q_2 = \left[\begin{array}{c|c} 1 & 0 \\ \hline 0 & 1 \end{array} \right] \begin{array}{c} 0 \\ H_2 \end{array}$$

$$A^{(n)} = H^{(n)} \text{ Hessenberg matrix}$$

$$H^{(n)} = \left[\begin{array}{c} \diagup \quad \times \quad \diagdown \end{array} \right]$$

A symmetric \rightarrow Tridiagonal matrix

A non-symmetric \rightarrow Hessenberg matrix

A strongly sparse (many zeros under the main diagonal)

Givens method instead of the Householder method

$H_{(k)}$

$G(q, k)$

$(q, k): A^{(k)}(q, k) \neq 0$

QR factorization of $T^{(n)}$ or $H^{(n)}$ is much more convenient than the factorization of A both are converging to a triangular matrix with eigenvalues of A on the diagonal.

A non-symmetric

Preliminary step $A \rightarrow H^{(n)}$ is $\mathcal{O}(n^3)$

each QR factorization A is $\mathcal{O}(n^3)$

" " " $H^{(n)}$ is $\mathcal{O}(n^2)$

Stopping criterion

$$H^{(n)} = QR$$

QR-RQ \rightarrow Hessenberg matrix

$$H^{(n)} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ 0 & a_{32} & a_{33} & a_{34} \\ 0 & 0 & a_{43} & a_{44} \end{bmatrix}$$

$$G_1 = \begin{bmatrix} c_1 & s_1 & 0 & 0 \\ -s_1 & c_1 & 0 & 0 \\ 0 & 0 & & \\ 0 & 0 & I & \end{bmatrix}$$

G_1, G_2, G_3 tridiagonal

$$G_1 H^{(n)} = \begin{bmatrix} a_{11}^{(1)} & a_{12}^{(1)} & a_{13}^{(1)} & a_{14}^{(1)} \\ 0 & a_{22}^{(1)} & a_{23}^{(1)} & a_{24}^{(1)} \\ 0 & a_{32} & a_{33} & a_{34} \\ 0 & 0 & a_{43} & a_{44} \end{bmatrix}$$

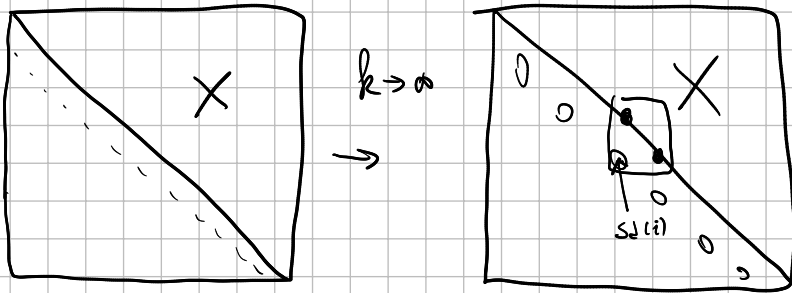
$$G_2 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & c_2 & s_2 & 0 \\ 0 & -s_2 & c_2 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$G_3 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & c_3 & s_3 \\ 0 & 0 & -s_3 & c_3 \end{bmatrix}$$

$G_3 G_2 G_1 =$ Hessenberg (lower)

$$G_2 G_1 = \begin{bmatrix} c_1 & s_1 & 0 & 0 \\ -s_1 c_2 & c_1 c_2 & s_2 & 0 \\ s_1 s_2 & -s_2 c_2 & c_2 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$G_3 G_2 G_1 = \begin{bmatrix} c_1 & s_1 & 0 & 0 \\ -s_1 c_2 & c_1 c_2 & s_2 & 0 \\ s_1 s_2 c_3 & -s_2 c_2 c_3 & c_2 c_3 & s_3 \\ -s_1 s_2 s_3 & s_2 c_2 s_3 & -c_2 s_3 & c_3 \end{bmatrix}$$



$$G = G_3 G_2 G_1$$

$$G H^{(n)} = R$$

$$H^{(n)} = Q R$$

$$Q = G^T = G_1^T G_2^T G_3^T$$

↑ ↑ Hessenberg Hessenberg triangular

$$R Q =$$

$$|s_d(i)| \sim |d(i)|, |d(i-1)|$$

$$|s_d(i)| \ll |d(i)| + |d(i-1)|$$

$$|s_d(i)| < (|d(i)| + |d(i-1)|) \tan$$

$$k \sim n$$

$G = G_3 G_2 G_1 =$ Hessenberg matrix (lower)

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & c_3 & s_3 \\ 0 & 0 & -s_3 & c_3 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & c_2 & s_2 & 0 \\ 0 & -s_2 & c_2 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} c_1 & s_1 & 0 & 0 \\ -s_1 & c_1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & c_3 & s_3 \\ 0 & 0 & -s_3 & c_3 \end{bmatrix} \begin{bmatrix} c_1 & s_1 & 0 & 0 \\ -s_1 c_2 & c_1 c_2 & s_2 & 0 \\ s_1 s_2 & -c_1 s_2 & c_2 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} =$$

$$\begin{bmatrix} c_1 & s_1 & 0 & 0 \\ -s_1 c_2 & c_1 c_2 & s_2 & 0 \\ s_1 s_2 c_3 & -c_1 s_2 c_3 & c_2 c_3 & s_3 \\ -s_1 s_2 s_3 & c_1 s_2 s_3 & -c_2 s_3 & c_3 \end{bmatrix}$$

G : lower Hessenberg matrix | $H^{(3)}$: upper Hessenberg matrix

$$G H^{(3)} = R$$

$$H^{(3)} = G^T R = Q R$$

$G^T = Q$ upper Hessenberg matrix

$R Q$ is still upper Hessenberg