

direct methods

iterative methods

eigenvalues
eigenvectors

$n \times n$ matrix

$$\det(A - \lambda I) = 0$$

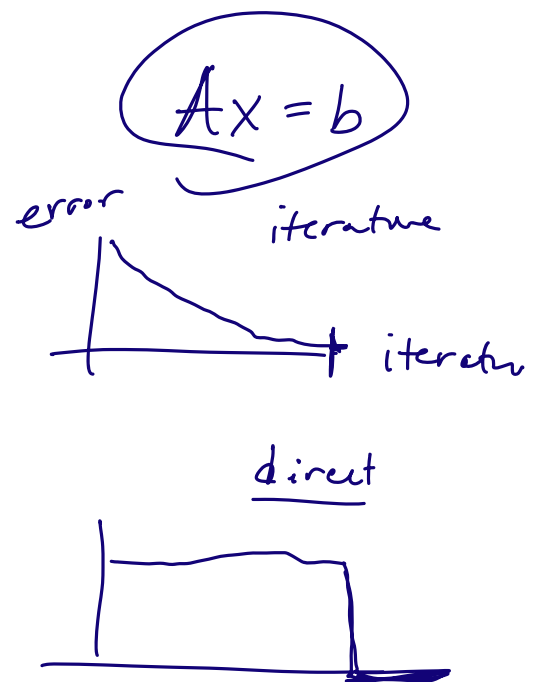
$$A\vec{x} = \lambda\vec{x} \Rightarrow$$

$$(A - \lambda I)\vec{x} = 0$$

$p(\lambda)$ of degree n

n eigenvalues

x_0
 \downarrow
 x_1
 \downarrow
 \vdots
 \downarrow
 x_n



$$\underline{A\vec{x} = \lambda \vec{x}}$$

eigenvalues? eigenvectors?

shifts

$$\begin{aligned} \underline{(A + sI)} \vec{x} &= A\vec{x} + s\vec{x} \\ &= \lambda \vec{x} + s\vec{x} \\ &= (\lambda + s) \vec{x} \end{aligned}$$

- eigenvalues shifted
- eigenvectors same

inverse

$$\begin{aligned} &A^{-1} \text{ exists} \\ &\Rightarrow \lambda \neq 0 \\ &A\vec{x} = 0 \cdot \vec{x} = \vec{0} \end{aligned}$$

$$A\vec{x} = \lambda \vec{x}$$

$$\underline{A^{-1}A} \vec{x} = \lambda A^{-1} \vec{x}$$

$$\underline{I} \quad A^{-1} \vec{x} = \frac{1}{\lambda} \vec{x}$$

$$\frac{1}{\lambda} \vec{x} = \cancel{A^{-1}} \vec{x}$$

A^{-1}

- eigenvalues are the reciprocals

- eigenvectors same

Powers of A

$$\boxed{A\vec{x} = \lambda \vec{x}}$$

$$\underline{AA}\vec{x} = \lambda \underline{A}\vec{x}$$

$$A^2 \vec{x} = \lambda \lambda \vec{x} = \lambda^2 \vec{x}$$

$$A \quad A^2 \vec{x} = \underline{A} \quad \lambda^2 \vec{x}$$

$$\boxed{A^k \vec{x} = \lambda^k \vec{x}}$$

polynomials of A $A \text{ } n \times n$ $Ax = \lambda x$

$$p(t) = c_0 + c_1 t + c_2 t^2 + \dots + c_k t^k$$

$$p(A) = \underline{c_0 I} + c_1 A + c_2 \underline{A^2} + \dots + c_k \underline{A^k}$$

$n \times n$

$$\begin{aligned} p(A) \underline{\vec{x}} &= c_0 \vec{x} + c_1 Ax + c_2 \underline{A^2 x} + \dots + c_k A^k x \\ &= \underline{c_0 \vec{x}} + c_1 \lambda \vec{x} + c_2 \lambda^2 \vec{x} + \dots + c_k \lambda^k \vec{x} \end{aligned}$$

$$= \underbrace{(c_0 + c_1 \lambda + \dots + c_k \lambda^k)}_{p(\lambda)} \underline{\vec{x}}$$

• eigenvalues $p(\lambda)$

• \vec{x} unchanged

Similarity $A, B \text{ } n \times n$ are "similar"

\exists invertible $T \text{ } n \times n$

$$\textcircled{A} = \underline{T \textcircled{B} T^{-1}}$$

$$Ax = \lambda x$$

$$Ax = \underbrace{T B T^{-1} x}_{= \lambda x}$$

$$B T^{-1} x = T^{-1} (\lambda x)$$

$$B \underbrace{(T^{-1} x)}_{\text{"}} = \lambda (T^{-1} x)$$

$$B y = \lambda y$$

- same eigenvalues λ
- eigenvectors transformed by T^{-1}

If A has a full set of eigenvectors

$$A \begin{bmatrix} | & & | \\ v_1 & \dots & v_n \\ | & & | \end{bmatrix} = \begin{bmatrix} | & & | \\ v_1 & \dots & v_n \\ | & & | \end{bmatrix} \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$$

$\{v_i \mid i=1, \dots, n\}$ are lin. indep. set, then
 V is invertible

$$A V = V \Lambda$$

$$A = V \Lambda V^{-1}$$

eigen
decomposition
of A

$$\begin{pmatrix} d_1 & & \\ & d_2 & \\ & & \ddots \\ & & & d_n \end{pmatrix}$$

$$\lambda_i': d_1, d_2, \dots, d_n$$

$$\begin{pmatrix} u_{11} & \dots & u_{1n} \\ & u_{22} & \\ & & \ddots \\ & & & u_{nn} \end{pmatrix}$$

$$\lambda_i': u_{11}, u_{22}, \dots, u_{nn}$$

Special A for which

$$A = Q \Lambda Q^T$$

λ_i' are real

A real symmetric $A = A^T$

A complex Hermitian $A = \bar{A}^T = A^H = A^*$

$A = Q \Lambda Q^H$

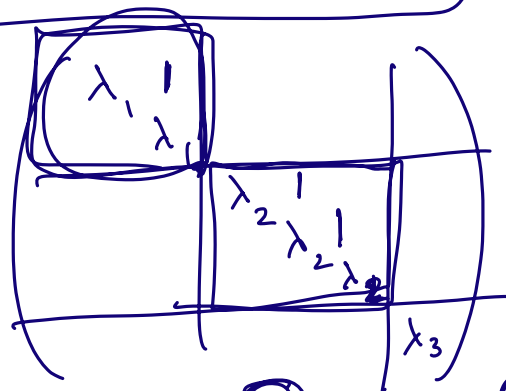
$$U \Sigma V^T$$

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n \geq 0$$

$$A = V \Lambda V^{-1} \leftarrow$$

Jordan normal form

$$A = V J V^{-1}$$



defective

$$p(\lambda) = (\lambda - \lambda_1)^{\textcircled{2}} (\lambda - \lambda_2)^{\textcircled{3}} (\lambda - \lambda_3)$$

λ_1 algebraic multiplicity 2

λ_2 " 3

λ_3 1

$$\left[\begin{array}{l} \text{geometric multiplicity } \lambda \\ \dim(S_\lambda) \end{array} \right]$$

$$\vec{v}_1, \vec{v}_2$$

$$\vec{v} = \begin{pmatrix} x \\ y \end{pmatrix}$$

$$S_\lambda = \{ \vec{v} \mid S\vec{v} = \lambda \vec{v} \}$$

defective

$$\begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = 2 \begin{pmatrix} x \\ y \end{pmatrix}$$

$$2x + y = 2x$$

$$2y = 2y \checkmark$$

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} \alpha \\ 0 \end{pmatrix}$$

$$\dim(S_2) = 1 < 2 \text{ (alg. mult.)}$$

(geom. mult.)

$$2x + 1 = 2x$$

$$1 = 0$$

defective matrices = non diagonalizable matrices

Normal Matrix

$$A A^H = A^H A$$

$$A = \underline{Q} \Lambda \underline{Q}^H$$

Examples:	A real symm	Λ real
	A Hermitian	Λ real
	A skew-Hermitian $A^H = -A$	Λ complex
	A unitary	Λ complex

Spectrum of A

$$= \{ \lambda_i \mid Ax_i = \lambda_i x_i \}$$

spectral radius

$$\rho(A) = \max_i |\lambda_i|$$

)

power iteration

$$x_1 = Ax_0$$

$$x_2 = Ax_1$$

\vdots

converges to an eigenvector v_1

s.t.

$$Av_1 = \lambda_1 v_1$$

λ_1 largest magnitude eigenvalues

Why? Assume A has a full set of eigenvectors: v_1, \dots, v_n form a basis for \mathbb{R}^n

$$x_0 = \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n$$

$$\underline{A^k x_0} = \alpha_1 \underline{A^k v_1} + \alpha_2 A^k v_2 + \dots + \alpha_n A^k v_n$$

$$= \alpha_1 (\lambda_1^k) v_1 + \alpha_2 \lambda_2^k v_2 + \dots + \alpha_n \lambda_n^k v_n$$

$$= \underbrace{(\lambda_1^k)}_{\text{circled}} \underbrace{(\alpha_1 v_1 + \alpha_2 \lambda_2^k v_2 + \dots + \alpha_n \lambda_n^k v_n)}_{\text{bracketed}}$$

assume

$$|\lambda_1| > |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_n|$$

Algorithm normalized power iteration

x_0
for $k = 1, 2, \dots$
 $y_k = A x_{k-1}$
 $x_k = y_k / \|y_k\|_\infty$
end

$$A v = \lambda v$$

$$A(\alpha v) = \alpha A v = \alpha \lambda v = \lambda (\alpha v)$$