

Claim: If $S = S^T$ and $S \in \mathbb{R}^{n \times n}$

λ is eigval of S , then

λ is real.

$$\lambda = \bar{\lambda}$$

$$\Downarrow$$

$$b = 0$$

$$\lambda = a + bi$$

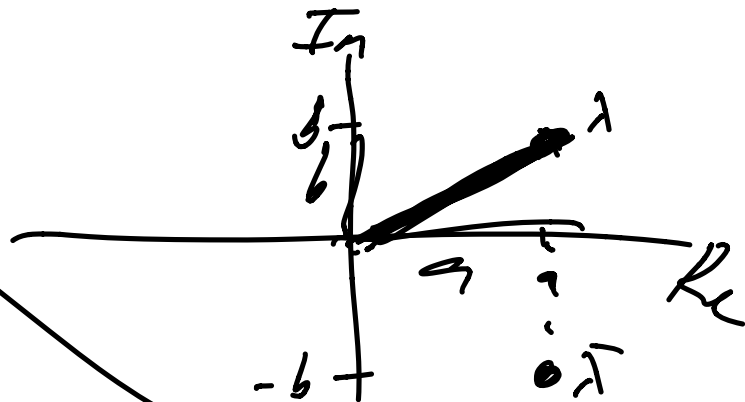
$$a = \operatorname{Re}(\lambda)$$

$$b = \operatorname{Im}(\lambda)$$

$$\bar{\lambda} = a - bi$$

$$i = \sqrt{-1}$$

$$\overline{\sqrt{z}} = \sqrt{\bar{z}}$$



$$Sx = \lambda x$$

$$x \neq 0$$

$$S \bar{x} = \bar{\lambda} \bar{x}$$

$$\bar{x}^T S^T = \bar{x}^T \bar{\lambda}$$

$$\overline{\begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}} = \begin{bmatrix} \bar{u}_1 \\ \bar{u}_2 \\ \bar{u}_3 \end{bmatrix}$$

$$\bar{x}^T S = \bar{x}^T \lambda$$

$$\downarrow$$

$$\boxed{\bar{x}^T S x} = \bar{x}^T \lambda x$$

$$= \lambda \bar{x}^T x$$

$$Sx = \lambda x$$

$$\boxed{\bar{x}^T S x} = \lambda \bar{x}^T x$$

$$\boxed{\lambda \bar{x}^T x = \lambda \bar{x}^T x}$$

$$\bar{x}^T x = \sum_{i=1}^n \bar{x}_i x_i = \sum_{i=1}^n |x_i|^2 \neq 0$$

$$> 0$$

$$\boxed{\bar{\lambda} = \lambda}$$

$$(a+bi) =$$

$$\sqrt{a^2+b^2}$$

Hence λ is real.

Claim: $S = S^T$

Assume $\lambda_1 \neq \lambda_2$

$Sx = \lambda_1 x$ $x, y \neq 0$

$Sy = \lambda_2 y$

$y^T S x = \lambda_1 y^T x$

$x^T S y = \lambda_2 x^T y$

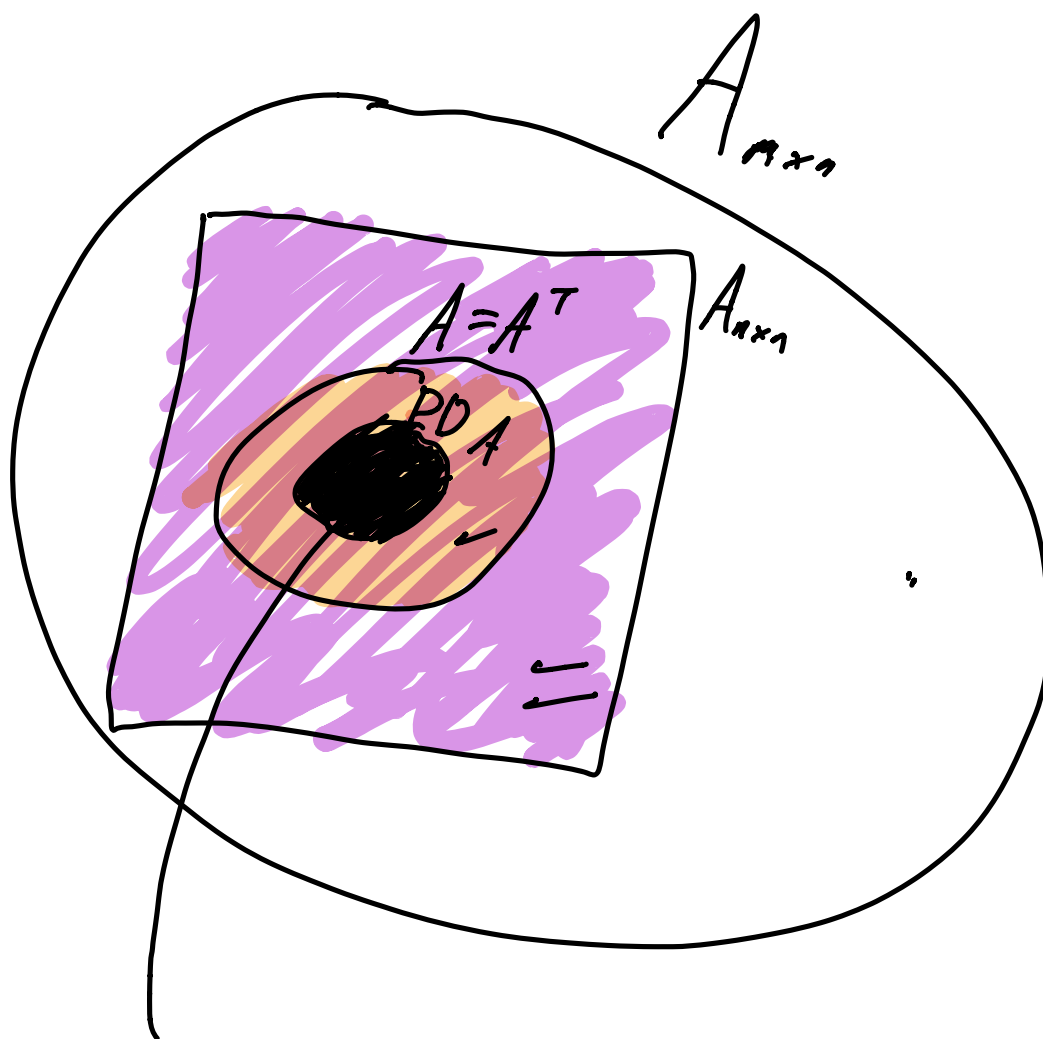
$y^T S x = \lambda_2 y^T x$

$\lambda_1 y^T x = \lambda_2 y^T x$

$y^T x = 0$

Hence $x \perp y$.

□



$S \succ 0$ PSD

1) Eigvals > 0

2) $x^T S x > 0 \quad \forall x \neq 0$

PSD

$$x^T S x \geq 0 \quad \forall x$$

$$Sx = \lambda x$$

$$x \neq 0$$

$$\lambda > 0$$

λ real

$$\begin{aligned} x^T S x &= \lambda x^T x \\ &= \lambda \|x\|^2 > 0 \end{aligned}$$

Claim:

If A PD

and B PD

$C = A + B$ is PD

$$\underline{x^T C x} = x^T (A + B) x$$

$$= (x^T A + x^T B) x$$

$$= \underbrace{x'Ax}_{\geq 0} + \underbrace{x' \lambda x}_{\geq 0} \geq 0$$

C is PD

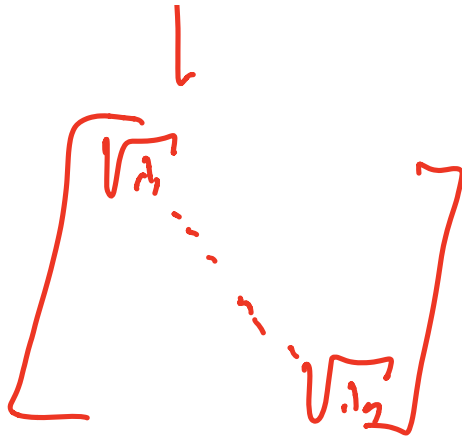
Criterion 3: $S = A^T A$

for A with indep cols.

$$S = A^T A$$

$$S = Q \Lambda Q^T$$

$$= Q \underbrace{\sqrt{\Lambda} \sqrt{\Lambda}}_{\uparrow} Q^T$$



$$A = Q \sqrt{\Lambda} Q^T$$

$$A^T A = \underbrace{Q \sqrt{\Lambda} Q^T}_{A^T} Q \sqrt{\Lambda} Q^T$$

$$= Q \Lambda Q^T$$

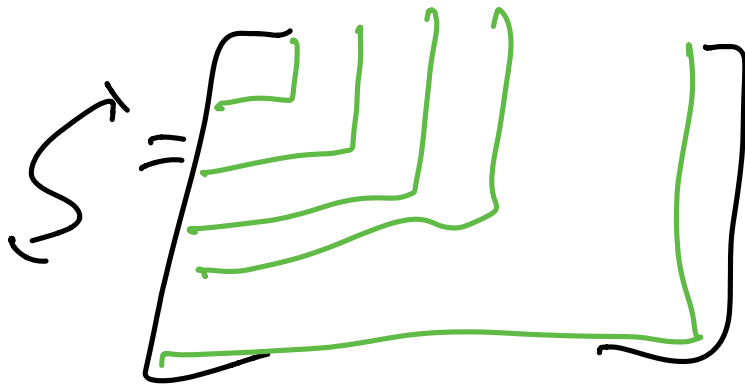
Cholesky decomposition

$$S = A^T A$$

Λ , \dots

A is lower triangular
(using LU decomp)

Sylvester Criterion



Least Squares

$$\boxed{Ax \equiv b}$$

$$A_{n \times n} \quad n > 1$$

$$\text{null}(A) = \{0\}$$

$$A^{\dagger} = (A^T A)^{-1} A^T$$

$$A^{\dagger} A x = A^{\dagger} b$$

$$\boxed{\hat{x} = A^{\dagger} b}$$

$$\arg \min_{x \in \mathbb{R}^n} \|Ax - b\|^2 = \arg \min_{x \in \mathbb{R}^n} \|Ax - b\|$$

$$L(x) = \|Ax - b\|^2 = (Ax - b)^T (Ax - b)$$

$$= \underbrace{(x^T A^T - b^T)}_{\text{row vector}} \underbrace{(Ax - b)}_{\text{column vector}}$$

$$= \underline{x^T A^T A x} - \underline{b^T A x} - \underbrace{x^T A^T b}_{+ b^T b}$$

$$= x^T A^T A x - b^T A x - b^T A x + b^T b$$

$$= x^T \underbrace{A^T A}_{\text{matrix}} - \underline{2 b^T A x} + \underline{b^T b}$$

$$- 2 \underbrace{(A^T b^T)^T}_{\text{row vector}} x + \underline{b^T b}$$

$$= \underline{x^T S x} + \underline{c^T x} + \underline{d}$$

$$ax^2 + bx + c$$

$$HL = \underline{S} = A^T A \text{ is } \underline{\underline{PD}}$$

Quiz 3

1) Projections

$$\underline{Ax = b}$$

$$\underline{P} = A \underbrace{(A^T A)^{-1} A^T}$$

$$\underline{A = QR}$$

2) Least Squares with $A \neq$

3) Eigen stuff

4) Diagonalization

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

$$A = \underline{\underline{1 \ 1 \ 1 \ 1}}$$

5) Sym.

$$6) \ x^T (\underline{\underline{A+B}}) x \Rightarrow 0$$

7) Hessian Matrix

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$f(u, v) = \underline{\underline{\begin{bmatrix} u + e^v \\ uv^2 \end{bmatrix}}}$$

SVD

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NO Q

$$A^+ = U \Sigma^* V^T$$

$$\hat{x} = A^+ b$$

$$\begin{bmatrix} \frac{1}{\sigma_1} & \frac{1}{\sigma_2} & \dots & \frac{1}{\sigma_n} \end{bmatrix}$$

$$A \hat{x} = b$$

$$A^+ = (A A^T)^{-1} A^T$$