

Ellipsoids

- ▶ ellipsoids
- ▶ ellipsoids in estimation

Ellipsoids

if $A = A^T > 0$, the set

$$\mathcal{E} = \{ x \mid x^T A x \leq 1 \}$$

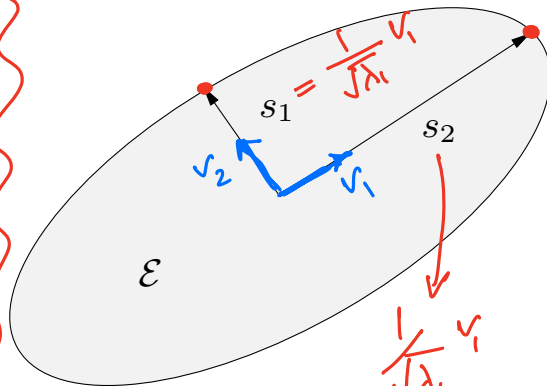
is an *ellipsoid* in $\mathbb{R}^n$, centered at 0

$x = \alpha v$: eigenvector ($\|v\|=1$)

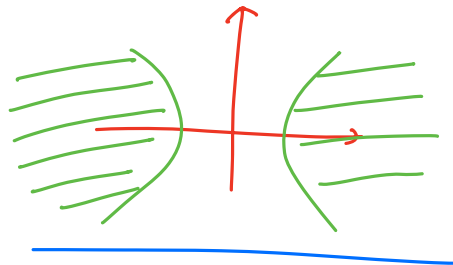
$$\begin{aligned} x^T A x &= \alpha^2 \cdot \underbrace{v^T A v}_{\lambda v} \\ &= \alpha^2 \lambda \underbrace{v^T v}_{\|v\|=1} = \alpha^2 \lambda \end{aligned}$$

Boundary: $x^T A x = 1$

$$\alpha \lambda^2 = 1 \rightarrow \alpha = \frac{1}{\sqrt{\lambda}}$$



$$\lambda_i < 0$$



$$A \leq 0$$

$\Downarrow$

$$\mathcal{E} = \{ x \mid x^T A x \leq 1 \} = \mathbb{R}^n$$

Ellipsoids

semi-axes are given by $s_i = \lambda_i^{-1/2} q_i$, i.e.:

- ▶ eigenvectors determine directions of semiaxes
- ▶ eigenvalues determine lengths of semiaxes

note:

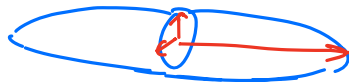
- ▶ in direction q_1 , $x^T A x$ is *large*, hence ellipsoid is *thin* in direction q_1
- ▶ in direction q_n , $x^T A x$ is *small*, hence ellipsoid is *fat* in direction q_n
- ▶ $\sqrt{\lambda_{\max}/\lambda_{\min}}$ gives maximum *eccentricity*

if $\tilde{\mathcal{E}} = \{ x \mid x^T B x \leq 1 \}$, where $B > 0$, then $\mathcal{E} \subseteq \tilde{\mathcal{E}} \iff A \geq B$

$\mathbb{R}^3$:

λ_1, λ_2 large
 λ_3 small

cigar



λ_1 large

λ_2, λ_3 small

pancake

