



# Principle Component Analysis



## Rayleigh Quotients

Physical meaning of SVD!

Given a symmetric matrix  $A \in S^n$ ,

$$\lambda_{\min}(A) \leq \frac{x^T A x}{x^T x} \leq \lambda_{\max}(A), \forall x \neq 0$$

$$\lambda_{\max}(A) = \max_{x: \|x\|_2=1} x^T A x$$

$$\lambda_{\min}(A) = \min_{x: \|x\|_2=1} x^T A x$$

The maximum and minimum are attained for  $x = u_1$  and for  $x = u_n$ , respectively, where  $u_1$  and  $u_n$  are the largest and smallest eigenvector of  $A$ , respectively.



## Rayleigh Quotients – Proof:

- Apply the [spectral theorem](#),  $U$  is orthogonal,  $\Lambda$  is diagonal

$$x^T A x = x^T U \Lambda U^T x = \bar{x}^T \Lambda \bar{x} = \sum_{i=1}^n \lambda_i \bar{x}_i^2$$

- Obviously,

$$\lambda_{\min} \sum_{i=1}^n \bar{x}_i^2 \leq \sum_{i=1}^n \lambda_i \bar{x}_i^2 \leq \lambda_{\max} \sum_{i=1}^n \bar{x}_i^2$$

- Also, orthogonal matrix  $U$  doesn't change the norm of any vector

$$\sum_{i=1}^n x_i^2 = x^T x = x^T U U^T x = (U^T x)^T (U^T x) = \bar{x}^T \bar{x} = \sum_{i=1}^n \bar{x}_i^2$$

- Combining the above 3 equations,

$$\lambda_{\min} x^T x \leq x^T A x \leq \lambda_{\max} x^T x$$