

## Lecture 9

- Rayleigh Quotients
- Applications of SVD
- Principal components analysis

## Announcements

- Practice exam posted

## Rayleigh Quotient

The Rayleigh quotient for symmetric matrix  $S$  is

$$R(\vec{x}) = \frac{\vec{x}^T S \vec{x}}{\vec{x}^T \vec{x}}$$

Given a matrix  $A$ , maximizing the Rayleigh quotient  $\frac{\vec{x}^T S \vec{x}}{\vec{x}^T \vec{x}}$  where  $S = A^T A$  gives a maximum of  $\sigma_1$  (largest singular value of  $A$ ) at  $\vec{x} = \vec{v}_1$  (first right singular vector)

Suppose we want to maximize  $\frac{\vec{x}^T S \vec{x}}{\vec{x}^T \vec{x}} \dots$

$$\frac{\partial}{\partial x_i} (\vec{x}^T \vec{x}) = \frac{\partial}{\partial x_i} (x_1^2 + x_2^2 + \dots + x_n^2) = 2x_i$$

$$\frac{\partial}{\partial x_i} (\vec{x}^T S \vec{x}) = \frac{\partial}{\partial x_i} \left( \sum_{i=1}^n \sum_{j=1}^n S_{ij} x_i x_j \right) = 2 \sum_{j=1}^n S_{ij} x_j = 2(S\vec{x})_i$$

Use the quotient rule

$$\frac{\partial}{\partial x_i} \left( \frac{\vec{x}^T S \vec{x}}{\vec{x}^T \vec{x}} \right) = \frac{2(\vec{x}^T S \vec{x}) (S \vec{x})_i - 2(\vec{x}^T S \vec{x}) x_i}{\dots}$$

Critical values when

$$(\vec{x}^T S \vec{x}) (S \vec{x})_i = (\vec{x}^T S \vec{x}) x_i$$

$$(S \vec{x})_i = \left( \frac{\vec{x}^T S \vec{x}}{\vec{x}^T \vec{x}} \right) x_i$$

$$S \vec{x} = \left( \frac{\vec{x}^T S \vec{x}}{\vec{x}^T \vec{x}} \right) \vec{x}$$

eigenvectors of  $S$   
one critical value, ...

We can show that  $\vec{v}_1$  maximizes the Rayleigh quotient  
with value  $\sigma_1$ .

— also, the minimum value is the smallest eigenvalue.

Preview

Generalized Rayleigh Quotient

Given symmetric matrices  $S$  and  $M$

a generalized RQ is  $\frac{\vec{x}^T S \vec{x}}{\vec{x}^T M \vec{x}}$