

## Overview

$$x^2 + \alpha x + \gamma$$

Low-rank approximation

Iteration

Equation solving

# SVD as Sum of Outer Products

- What's another way of writing the SVD?

$$\Sigma = \begin{pmatrix} \sigma_1 \\ \vdots \\ 0 \\ \vdots \end{pmatrix} \Leftrightarrow \begin{pmatrix} \sigma_1 \\ \vdots \\ 0 \\ \vdots \end{pmatrix}$$

$$\text{rank}(u v^T) = 1$$

$$A = U \Sigma V^T = \begin{pmatrix} | & & | \\ u_1 & \dots & u_n \\ | & & | \end{pmatrix} \begin{pmatrix} \sigma_1 & & \\ & \ddots & \\ & & \sigma_n \end{pmatrix}$$

$$= \begin{pmatrix} | & & | \\ u_1 & \dots & u_n \\ | & & | \end{pmatrix} \begin{pmatrix} \sigma_1 v_1^T \\ \vdots \\ \sigma_n v_n^T \end{pmatrix}$$

inner product  
 $\vec{u}^T \vec{v} = \text{one number}$

outer product  
 $\vec{u} v^T =$

$$\Sigma \rightarrow \begin{pmatrix} | & & | \\ \sigma_1 & & \\ | & & | \end{pmatrix} \begin{pmatrix} | & & | \\ v_1^T & & \\ | & & | \end{pmatrix}$$

$\begin{pmatrix} 1 \\ 0 \\ 0 \\ 3 \end{pmatrix}$	$\begin{pmatrix} 10 & 20 & 30 \\ 10 & 20 & 30 \\ 20 & 40 & 60 \\ 30 & 60 & 90 \end{pmatrix}$	$\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$
--------------------------------------------------	----------------------------------------------------------------------------------------------	---------------------------------------------

$$= \sigma_1 \vec{u}_1 \vec{v}_1^T + \sigma_2 \vec{u}_2 \vec{v}_2^T + \dots + \sigma_n \vec{u}_n \vec{v}_n^T$$

## Low-Rank Approximation (I)

- What is the *rank* of  $\sigma_1 \underline{u_1 v_1^T}$ ? 1
- What is the *rank* of  $\sigma_1 \underline{u_1 v_1^T} + \sigma_2 \underline{u_2 v_2^T}$ ? 2

**Demo:** Image Compression ↩

## Low-Rank Approximation

- What can we say about the low-rank approximation

$$A_k = \sigma_1 \mathbf{u}_1 \mathbf{v}_1^T + \dots + \sigma_k \mathbf{u}_k \mathbf{v}_k^T$$

to

$$\rightarrow A = \sigma_1 \mathbf{u}_1 \mathbf{v}_1^T + \sigma_2 \mathbf{u}_2 \mathbf{v}_2^T + \dots \left( + \sigma_{k+1} \mathbf{u}_{k+1} \mathbf{v}_{k+1}^T \right) \leftarrow \text{SVD}$$

For simplicity, assume  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n > 0$ .

Then among all rank- $k$  matrices  $B$ ,  $A_k$  is the one that satisfies

$$\min_B \|A - B\|_2 = \|A - A_k\|_2$$

$$\text{Also: } \|A - A_k\|_2 = \left\| \underbrace{\sigma_{k+1} \mathbf{u}_{k+1} \mathbf{v}_{k+1}^T + \dots}_{\text{rest of terms}} + \dots + \sigma_n \mathbf{u}_n \mathbf{v}_n^T \right\|_2 = \sigma_{k+1}$$

[Spring 17: Add PCA]

$$\|A - A_k\|_F = \sqrt{\sigma_{k+1}^2 + \dots + \sigma_n^2}$$

**Part 3:**  
**Approximation—When the  
Exact Answer is Out of Reach**

# 14 Iteration and Convergence



- What is linear convergence? quadratic convergence?

for power it:

$$\|e_{k+1}\| = \left(\frac{\lambda_2}{\lambda_1}\right) \|e_k\|$$

$$\|e_{k+1000}\| = \left(\frac{\lambda_2}{\lambda_1}\right)^{1000} \|e_k\| \leftarrow \text{linear convergence}$$

- gains a fixed  
possibly fractional  
number of digits  
every time

$$\|e_{k+1}\| \lesssim C \cdot \|e_k\|$$

$$\|e_{k+1}\| < \underline{C} \cdot \|e_k\|^2 \leftarrow \text{quadratic convergence}$$

Example for quadratic:

$$\|e_1\| = 0.1 \sim 1$$

$$\|e_2\| = 0.01 = 10^{-2} \quad \sim 2 \text{ digits}$$

$$\|e_3\| = 10^{-4} \quad \rightarrow 4 \text{ digits}$$

$$\|e_4\| = (10^{-2})^2 = 10^{-4} = 10^{-8} \quad \sim 8 \text{ digits}$$

$C=1$

An iterative method converges with rate  $r$  if

$$\lim_{k \rightarrow \infty} \frac{\|e_{k+1}\|}{\|e_k\|^r} = c \begin{cases} > 0 \\ < \infty \end{cases}$$

convergent with rate 1 : linear

2 : quadratic

## About Convergence Rates

### Demo: Rates of Convergence

- Characterize linear, quadratic convergence in terms of the 'number of accurate digits'.

# 15 Solving One Equation

## Solving Nonlinear Equations

- What is the goal here?

Given: equation

$$f(x) = 0$$

Find  $x$  so that the equation is true.

$$f(x) = y$$

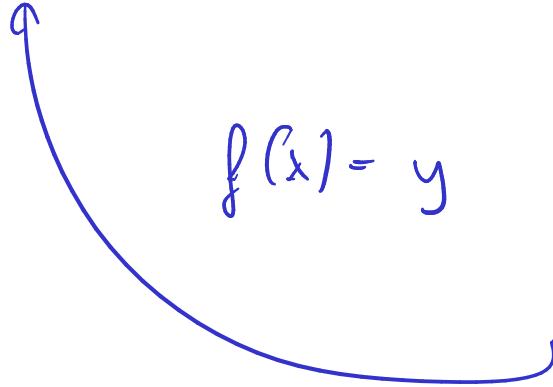
$$\hat{f}(x) = f(x) - y$$

↓ solve

$$\tilde{f}(x) = 0$$

$$ax + b = 0 \leftarrow$$

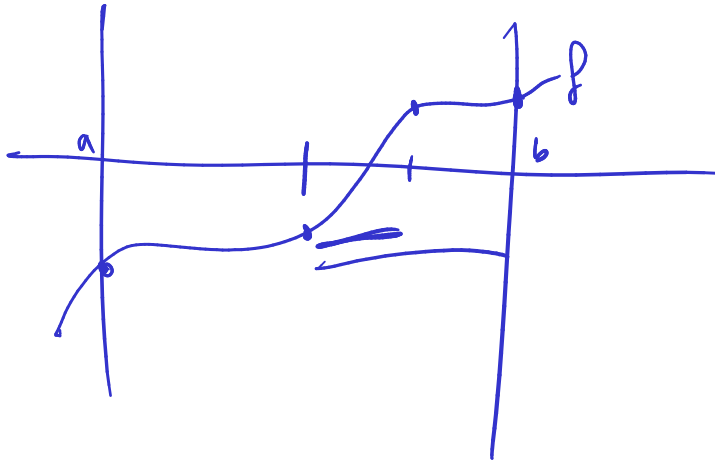
$$\underline{ax^2 + bx + c = 0}$$



# Bisection Method

## Demo: Bisection Method

- What's the rate of convergence? What's the constant?



$\epsilon_i$

$f(a)$  has a different  
sign than  $f(b)$

$$f(a) > 0$$
$$f(b) < 0$$

## Newton's Method

- Derive Newton's method.