

# LECTURE 5

MATH 504

FALL 2011

## SINGULAR VALUE DECOMPOSITION

### Orthogonality:

A matrix  $Q \in \mathbb{R}^{n \times n}$  ( $U \in \mathbb{C}^{n \times n}$ ) is called orthogonal (unitary) if

$$Q^T Q = I \quad (U^* U = I)$$

e.g.

$$\tilde{Q} = \begin{bmatrix} 1/\sqrt{5} & -2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix} \text{ is orthogonal.}$$

$$Q^T Q = I \iff q_i^T q_j = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$$

$$\iff q_i \perp q_j \text{ and } \|q_i\| = 1$$

$i \neq j$

$$\iff \{q_1, \dots, q_n\} \text{ is an orthonormal set}$$

Similarly for a unitary matrix  $U \in \mathbb{C}^{n \times n}$  the set  $\{u_1, \dots, u_n\}$  of columns is an orthonormal set.

9.

For the example above  $\left\{ \begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}, \begin{bmatrix} -2/\sqrt{5} \\ 1/\sqrt{5} \end{bmatrix} \right\}$   
is an orthonormal set in  $\mathbb{R}^2$

### THM (Unitary Invariance)

Let  $U_1 \in \mathbb{C}^{m \times m}$  and  $U_2 \in \mathbb{C}^{n \times n}$  be unitary matrices. For all  $A \in \mathbb{C}^{m \times n}$  we have

$$(i) \|A\|_2 = \|U_1 A\|_2 = \|A U_2\|_2$$

$$(ii) \|A\|_F = \|U_1 A\|_F = \|A U_2\|_F$$

### PROOF

(i)

$$\boxed{\|A\|_2 = \|U_1 A\|_2}$$

$$\|U_1 A\|_2 = \max_{\substack{x \in \mathbb{R}^n \\ \text{s.t.} \\ \|x\|_2 = 1}} \|U_1 A x\|_2$$

$$= \max_{\substack{x \in \mathbb{R}^n \\ \text{s.t.} \\ \|x\|_2 = 1}} \sqrt{(U_1 A x)^* (U_1 A x)}$$

$$= \max_{\substack{x \in \mathbb{R}^n \\ \text{s.t.} \\ \|x\|_2 = 1}} \sqrt{x^* A^* \underbrace{U_1^* U_1}_I A x}$$

$$= \max_{\substack{x \in \mathbb{R}^n \\ \text{s.t.} \\ \|x\|_2 = 1}} \|A x\|_2 = \|A\|_2$$

(2)

$$\boxed{\|A\|_2 = \|AU_2\|_2}$$

$$\|AU_2\|_2 = \max_{\substack{x \in \mathbb{R}^n \\ \text{s.t.} \\ \|x\|_2 = 1}} \|AU_2 x\|_2$$

Letting  $y := U_2 x$  we have

$$(1) \|y\|_2 = \sqrt{x^* U_2^* U_2 x} = \sqrt{x^* x} = \|x\|_2 = 1.$$

(2) Each  $x$  is mapped to unique  $y$ ,  
i.e.,  $x = U_2^* y$ .

Consequently

$$\|AU_2\|_2 = \max_{\substack{y \in \mathbb{R}^n \\ \text{s.t.} \\ \|y\|_2 = 1}} \|Ay\|_2 = \|A\|_2$$

(ii) Exercise

### Existence of SVD

Definition of Singular Values and Vectors

Let  $A \in \mathbb{R}^{2 \times 2}$  with SVD

$$A = \begin{bmatrix} u_1 & u_2 \end{bmatrix} \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{bmatrix} \begin{bmatrix} v_1^* \\ v_2^* \end{bmatrix}$$

③

Then

$$(1) u_1^* A = \sigma_1 v_1^* \quad \text{and} \quad A v_1 = \sigma_1 u_1$$

$$(2) u_2^* A = \sigma_2 v_2^* \quad \text{and} \quad A v_2 = \sigma_2 u_2$$

DEFN (Singular Values)

Let  $A \in \mathbb{C}^{m \times n}$ . Suppose that the equations

$$(i) A v = \sigma u, \quad \text{and}$$

$$(ii) u^* A = \sigma v^*$$

hold for some  $v \neq 0 \in \mathbb{C}^n$ ,  $u \neq 0 \in \mathbb{C}^m$  and real scalar  $\sigma \geq 0$ . Then

(i)  $\sigma$  is called a singular value,

(ii) and  $u, v$  are called the left and right singular vectors associated with  $\sigma$ .

e.g.

SVD for

$$\begin{bmatrix} 1 & -3 \\ 3 & 1 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} \sqrt{10} & 0 \\ 0 & \sqrt{10} \end{bmatrix} \begin{bmatrix} -1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix}$$

$\sigma_1 = \sqrt{10}$  is a singular value,

and  $u_1 = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$ ,  $v_1 = \begin{bmatrix} -1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}$  are the associated left/right singular vectors. (4)

# Proof of Existence of SVD

## Basic Idea

\* Find unitary transformations  $U$  and  $V$  s.t.

$U^* A V$  is diagonal

## EXAMPLE

Let

$$A = \begin{bmatrix} 1 & -3 \\ 3 & 1 \end{bmatrix}$$

For all  $x$  s.t.  $\|x\|_2 = 1$

$$\|Ax\|_2 = \left\| \begin{bmatrix} x_1 - 3x_2 \\ 3x_1 + x_2 \end{bmatrix} \right\|_2 = \sqrt{10x_1^2 + 10x_2^2} = \sqrt{10}$$

Consequently  $\|A\|_2 = \sqrt{10}$ . For instance

for  $v_1 = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$  we have

$$Av_1 = \sqrt{10} \underbrace{\begin{bmatrix} -1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}}_{u_1, \|u_1\|_2 = 1} \quad \left( \begin{array}{l} \text{Note} \\ u_1^T A = \sigma_1 v_1^T \end{array} \right)$$

Choose  $v_2$  so that

(1)  $\{v_1, v_2\}$  is orthonormal, equivalently

(2)  $V = [v_1 \ v_2]$  is orthogonal

⑤

Similarly choose  $u_2$  so that

$U = [u_1 \ u_2]$  is orthogonal

Specifically

$$V = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \text{ and } U = \begin{bmatrix} -1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix}$$

Now consider

$$u_1^T A v_1 = \sqrt{10} \ u_1^T u_1 = \sqrt{10}$$

$$u_2^T A v_1 = \sqrt{10} \ u_2^T u_1 = 0 \quad (\text{Because } u_2 \perp u_1)$$

That is

$$U^T A v_1 = \begin{bmatrix} \sqrt{10} \\ 0 \end{bmatrix}$$

whereas

$$U^T A v_2 = \begin{bmatrix} u_1^T A v_2 \\ u_2^T A v_2 \end{bmatrix} = \begin{bmatrix} \sqrt{10} \ v_1^T v_2 \\ u_2^T A v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ u_2^T A v_2 \end{bmatrix}$$

In other words

$$U^T A V = [U^T A v_1 \quad U^T A v_2] \text{ is diagonal.}$$

Indeed

$$\begin{bmatrix} -1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix} A \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} \sqrt{10} & 0 \\ 0 & \sqrt{10} \end{bmatrix}$$

(6)

Consequently

$$A = \underbrace{\begin{bmatrix} -1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix}}_U \underbrace{\begin{bmatrix} \sqrt{10} & 0 \\ 0 & \sqrt{10} \end{bmatrix}}_\Sigma \underbrace{\begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}}_{V^T}$$

THM (Existence of SVD)

Every matrix  $A \in \mathbb{C}^{n \times n}$  has an SVD of the form

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

where  $V, U \in \mathbb{C}^{n \times n}$  are unitary, and  $\Sigma$  is diagonal with non-negative entries.

PROOF

Let  $\sigma_1 := \|A\|_2$ . Since the maximization

$$\begin{aligned} & \max_{x \in \mathbb{C}^n} \|Ax\|_2 \\ & \text{s.t.} \\ & \|x\|_2 = 1 \end{aligned}$$

involves a

\* continuous objective function

$$x \longrightarrow \|Ax\|_2$$

\* over a compact domain

$$\{x \in \mathbb{C}^n \mid \|x\|_2 = 1\},$$

it will be attained at some  $v_1 \in \mathbb{C}^{n \times n}$ , i.e.,

(7)

$$Av_1 = \sigma_1 u_1$$

for some  $v_1, u_1 \in \mathbb{C}^n$ , where  $\|v_1\| = \|u_1\| = 1$ .

Choose vectors  $v_2, \dots, v_n \in \mathbb{C}^n$  s.t.

$$\hat{V} = [v_1 \ v_2 \ \dots \ v_n] \in \mathbb{C}^{n \times n}$$

is orthogonal. Similarly, choose  $u_2, \dots, u_n \in \mathbb{C}^n$  s.t.

$$\hat{U} = [u_1 \ u_2 \ \dots \ u_n] \in \mathbb{C}^{n \times n}$$

is orthogonal.

Since  $u_2^* u_1 = u_3^* u_1 = \dots = u_n^* u_1 = 0$ , we have

$$\hat{U}^* A \hat{V} = \begin{bmatrix} u_1^T \sigma_1 u_1 & \hat{U}^* A v_2 & \dots & \hat{U}^* A v_n \\ u_2^T \sigma_1 u_1 & & & \\ \vdots & & & \\ u_n^T \sigma_1 u_1 & & & \end{bmatrix}$$

$$= \begin{bmatrix} \underbrace{\sigma_1}_{1 \times 1} & \underbrace{w}_{1 \times n} \\ \underbrace{0}_{n \times 1} & \underbrace{\tilde{A}}_{(n-1) \times (n-1)} \end{bmatrix}$$

Because of unitary invariance

$$\|\hat{U}^* A \hat{V}\|_2 = \|A\|_2 = \sigma_1 \geq \frac{\left\| \begin{bmatrix} \sigma_1 & w \\ 0 & \tilde{A} \end{bmatrix} \begin{bmatrix} \sigma_1 \\ w^* \end{bmatrix} \right\|_2}{\left\| \begin{bmatrix} \sigma_1 \\ w \end{bmatrix} \right\|_2}$$

$$\geq (\sigma_1^2 + w^* w) / \sqrt{\sigma_1^2 + w^* w} \quad \textcircled{8}$$



therefore  $w=0$ , and  $U^T A V$  is of the form

$$(*) \hat{U}^* A \hat{V} = \begin{bmatrix} \overset{1 \times 1}{\sigma_1} & 0 \\ 0 & \underbrace{\tilde{A}}_{(n-1) \times (n-1)} \end{bmatrix}$$

Now the proof follows from induction. For a  $1 \times 1$  scalar, there is obviously an SVD. Let us suppose for the  $(n-1) \times (n-1)$  matrix  $\tilde{A}$  there is an SVD of the form

$$\tilde{A} = \tilde{U} \tilde{\Sigma} \tilde{V}^*$$

Then from (\*)

$$\hat{U}^* A \hat{V} = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \tilde{U} \tilde{\Sigma} \tilde{V}^* \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & \tilde{U} \end{bmatrix} \begin{bmatrix} \sigma_1 & 0 \\ 0 & \tilde{\Sigma} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & \tilde{V}^* \end{bmatrix}$$

$$\implies A = \underbrace{\hat{U} \begin{bmatrix} 1 & 0 \\ 0 & \tilde{U} \end{bmatrix}}_U \underbrace{\begin{bmatrix} \sigma_1 & 0 \\ 0 & \tilde{\Sigma} \end{bmatrix}}_{\Sigma} \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & \tilde{V}^* \end{bmatrix}}_{V^*}$$

where  $U := \hat{U} \begin{bmatrix} 1 & 0 \\ 0 & \tilde{U} \end{bmatrix}$  and  $V := \hat{V} \begin{bmatrix} 1 & 0 \\ 0 & \tilde{V} \end{bmatrix}$   $\square$