

Announcements

1) HW 1 due today, 5:00

2) HW 2 up either tonight
or tomorrow

Recall

Theorem: (unitary diagonalization)

If $A \in \mathbb{C}^{m \times m}$ satisfies

$$A = A^*$$

is unitarily diagonalizable;

i.e., \exists diagonal matrix

$D \in \mathbb{C}^{m \times m}$ and Q unitary,

$Q \in \mathbb{C}^{m \times m}$, with

$$A = Q D Q^*$$

Consequence: A^*A (for
 $A \in \mathbb{C}^{m \times n}$) is always equal
to its adjoint. Moreover,
if λ is an eigenvalue for
 A^*A and v is an associated
eigenvector, then

$$(A^* A v)^* v$$

$$= (\lambda v)^* v$$

$$= \bar{\lambda} v^* v$$

$v^* v$ is a positive scalar.

However, by adjoint

properties,

$$(A^*Av)^*$$

$$v$$

$$= (Av)^* (Av) \quad (\text{adjoint property})$$

$$\geq 0.$$

If zero, then

$$\lambda v^*v = 0 \Rightarrow \lambda = 0.$$

If bigger than zero,

$$\bar{\lambda} v^*v > 0,$$

$$\bar{\lambda} > 0.$$

This shows that the eigenvalues of A^*A are non-negative real numbers.

We can write

$$A^*A = Q D Q^*$$

with the diagonal entries of D nonnegative real numbers.

Definition: (absolute value)

Let $A \in \mathbb{C}^{m \times n}$. We

define the absolute value

of A to be the matrix

$$|A| = Q \sqrt{D} Q^*$$

where

$$A^*A = QDQ^*$$

and \sqrt{D} is the diagonal matrix whose entries are the entries of D , square-rooted.

Fact about complex numbers

$$z \in \mathbb{C}.$$

Can write

$$z = e^{i\theta} |z| \text{ (polar decomposition)}$$

for some θ , $0 \leq \theta < 2\pi$.

For complex numbers,

$$|e^{i\theta}| = 1 \quad \forall \theta \in \mathbb{R}.$$

Example 1:

$$\text{Let } z = \frac{5\sqrt{3}}{2} + \frac{5}{2}i.$$

$$\begin{aligned}\text{Then } |z| &= 5 \left| \frac{\sqrt{3}}{2} + \frac{i}{2} \right| \\ &= 5.\end{aligned}$$

$$\theta = \frac{\pi}{6} = \arctan\left(\frac{\frac{1}{2}}{\frac{\sqrt{3}}{2}}\right)$$

We write

$$z = 5 e^{\frac{\pi i}{6}} .$$

Theorem: (polar decomposition)

$A \in \mathbb{C}^{m \times n}$. Then there
exists a matrix $Q \in \mathbb{C}^{m \times n}$
with orthonormal columns

such that

$$A = Q |A|$$

(polar decomposition for
matrices)

Singular value decomposition

(reduced)

Given $A \in \mathbb{C}^{m \times n}$, write

$A = Q |A|$ in polar decomposition. Diagonalize

$$|A| = R \sqrt{D} R^*$$

Let $P \in \mathbb{C}^{n \times n}$ be the permutation unitary that rearranges the entries on the diagonal of \sqrt{D} so that in the (diagonal) matrix

$$P \sqrt{D} P^* = \hat{\Sigma},$$

$$\left(\hat{\Sigma} \right)_{1,1} \geq \left(\hat{\Sigma} \right)_{2,2} \geq \dots \geq \left(\hat{\Sigma} \right)_{n,n}$$

We write

$$\hat{U} = QR P^* \in \mathbb{C}^{m \times n}$$

(orthonormal columns)

and

$$\hat{V} = RP^* \in \mathbb{C}^{n \times n}$$

(unitary).

Then $A = \hat{U} \hat{\Sigma} (\hat{V})^*$ the

reduced singular value decomposition
of A .

Facts

- 1) Every matrix has a singular value decomposition, unlike eigenvalue decompositions.
- 2) A real matrix has a real singular value decomposition:
 \hat{U} and \hat{V} are real matrices.

Matlab Calling Command

$\text{svd}(A)$ gives the diagonal entries of $\hat{\Sigma}$.

These are called the **singular values** of A .

$$[U, D, V] = \text{svd}(A, 0)$$

zero
↓

gives the reduced singular value decomposition of A , with

$$D = \hat{\Sigma}$$

Example 2 :

$$A = \begin{bmatrix} 1 & 3 \\ 5 & i \\ -6i & 10 \end{bmatrix}$$

Singular value decomposition of
A (4 decimal places)

$$\hat{U} = \begin{bmatrix} -.0550 + .1972i & .1817 + .42i \\ -.2951 - .003i & .8654 + .0065i \\ -.0303 + .9328i & -.0646 + .1936i \end{bmatrix}$$

$$\hat{\Sigma} = \begin{bmatrix} 12.4641 & 0 \\ 0 & 4.0799 \end{bmatrix}$$

$$\hat{\Sigma}^{-1/2} = \begin{bmatrix} -.5718 & .8204 \\ -.0378 + .8195i & -.0264 + .5712i \end{bmatrix}$$

Full SVD

Start with the reduced

SVD

$$A = \hat{U} \hat{\Sigma} (\hat{V})^*$$

Let $\hat{\Sigma}$ be the $m \times n$ diagonal matrix with rows 1 through n the rows of $\hat{\Sigma}$ and rows $n+1$ to m all zero.

Let U be the $m \times m$ unitary matrix obtained from \hat{U} by augmenting by $(m-n)$ orthonormal m -vectors which are orthogonal to the columns of \hat{U} (Gram-Schmidt process)

$$V = \hat{V}.$$

Then

$$A = U \Sigma V$$
 is the

full singular value decomposition
of A .

Matlab: $[U, D, V] = \text{svd}(A)$