

1. How to solve $Ax = b$ if we have
 $A = U \Sigma V^T$ (SVD)

$$Ax = b \iff U \Sigma V^T x = b$$

$$U y = b \quad (y = \Sigma V^T x)$$

$$y = U^T b$$

$$\Sigma V^T x = y$$

$$z = V^T x = \Sigma^{-1} y$$

$$\Sigma^{-1} = \begin{pmatrix} 1/\sigma_1 & & \\ & \ddots & \\ & & 1/\sigma_n \end{pmatrix}$$

$$V^T x = z \quad \Rightarrow \quad x = Vz$$

1.	$y = U^T b$
2.	$z = \Sigma^{-1} y$
3.	$x = Vz$

$$U \Sigma V^T x = b$$

$$\Sigma V^T x = U^T b = y$$

$$V^T x = \Sigma^{-1} y = z$$

$$x = Vz$$

2. $A = U \Sigma V^T$

Proof 1 $A^{-1} = (V^T)^{-1} \Sigma^{-1} U^{-1}$

$$= V \begin{pmatrix} 1/\sigma_1 & & \\ & \ddots & \\ & & 1/\sigma_n \end{pmatrix} U^T$$

Proof 2 $A = U \Sigma V^T$

$$A (V \Sigma^{-1} U^T) = (U \Sigma V^T) (V \Sigma^{-1} U^T)$$

$$= U \Sigma \cancel{(V^T V)} \Sigma^{-1} U^T$$

$$= U \cancel{\Sigma} \Sigma^{-1} U^T$$

$$= U U^T = I$$

Continue : Application 4

solve $Ax = b$ for LS solution

or $A^T A x = A^T b$ (A is tall)

Use non-fullspan version

$$A = U_1 \Sigma_1 V_1^T$$

$$U_1^T U_1 = I \quad \text{but } U_1 U_1^T \neq I$$

$$V_1^T V_1 = I \quad \text{but } V_1 V_1^T \neq I$$

$$\Sigma_1^{-1} = \begin{pmatrix} 1/\sigma_1 & & \\ & \ddots & \\ & & 1/\sigma_r \end{pmatrix} \quad \Sigma_1^T = \Sigma$$

$$A^T A = V_1 \Sigma_1 \cancel{U_1^T U_1} \Sigma_1 V_1^T$$

$$A^T A x = A^T b \quad V_1 \Sigma_1 \Sigma_1 V_1^T x = V_1 \Sigma_1 U_1^T b$$

$$\Sigma_1 \Sigma_1 V_1^T x = \Sigma_1 U_1^T b$$

$$\Sigma_1 V_1^T x = U_1^T b = y$$

$$V_1^T x = \Sigma_1^{-1} U_1^T b = z$$

$$x = V_1 z$$

Algorithm

$$1. \quad A = U_1 \Sigma_1 V_1^T$$

$$2. \quad y = U^T b$$

$$3. \quad z = \Sigma_1^{-1} y$$

$$4. \quad x = V_1 z$$

Consider fullspan SVD

$$A = [U_1, U_2] \begin{bmatrix} \Sigma_1 & 0 \\ 0 & 0 \end{bmatrix} [V_1, V_2]^T$$

$$= U_1 \Sigma_1 V_1^T$$

$$[V_1, V_2]^T [U_1, U_2] = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}$$

$$\begin{bmatrix} V_1^T \\ V_2^T \end{bmatrix} [V_1, V_2] = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}$$

$$\begin{bmatrix} V_1^T V_1 & V_1^T V_2 \\ V_2^T V_1 & V_2^T V_2 \end{bmatrix} = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}$$

$$V_1^T V_2 = 0, \quad V_2^T V_1 = 0$$

Revisit $V_1^T x = z$

$$V_1^T (V_1 x + V_2 u) = z + 0$$

The solution to $V_1^T x = z$

$$x_0 + V_2 u \quad \forall u$$

($x_0 = V_1 z$ is the minimum norm solution)

Application 5

Image processing & approximation

1. Image file \rightarrow matrix A (Image2Matrix.mpl)

$$2. A = \underbrace{[U_1, U_2]}_r \begin{bmatrix} \underbrace{\Sigma}_r \\ \underbrace{\Sigma}_r \end{bmatrix} \underbrace{[V_1, V_2]^T}_r$$

$$\approx U_1 \Sigma_1 V_1^T = B$$

3. matrix $B \rightarrow$ image array Q
(Matrix2Image.mpl)

4. ImageTools :- Preview (Q)

need much less storage

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Exam 2

1. Proof problems

2. Ask some "how to" question
on computation

Final exam (optional)