

SVD za normalne matrice:  $L^*L = LL^*$

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

normalna  $\Leftrightarrow$  lastni vektorji  
 $A = PDP^{-1} =$  so ortogonalni  
 $= PDP^*$  in vertice se da  
 $\leftarrow$  diagonalizirati

$$AV_i = \lambda_i V_i \Rightarrow A^* A V_i = \lambda_i A^* V_i$$

$$A^* V_i = \bar{\lambda}_i V_i \quad A A^* V_i = \bar{\lambda}_i A V_i$$

$$A^* A V_i = \bar{\lambda}_i \lambda_i V_i$$

lastne vrednosti  $A^* A$  so  $|\lambda|^2$ ,  
 torej so singularne vrednosti normalne matrice kar  
 pomeni lastnih vrednosti:

Uredimo jih po velikosti:

$$|\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_r| \geq |\lambda_{r+1}| = \dots = |\lambda_n| = 0$$

$\rightarrow$  rang  $A$

$$V = [V_1 \dots V_n] \quad \left( \begin{array}{l} u_i = \frac{1}{\sigma_i} \cdot A V_i = \frac{\lambda_i}{|\lambda_i|} V_i \\ \rightarrow \forall i \in \{1, \dots, r\} \end{array} \right.$$

dopolnimo  $u_{r+1}, \dots, u_n$  so kar  $v_i$ ,  
 saj je  $V$  ONB in  $u_1, \dots, u_r$  so  
 le skalirani  $v_i$ :

$$U = \left[ \begin{array}{cccc} \frac{\lambda_1}{|\lambda_1|} v_1 & \dots & \frac{\lambda_r}{|\lambda_r|} v_r & v_{r+1} \dots v_n \end{array} \right]$$

Frobeniusova norma in skalarni produkt:

$$\langle B, C \rangle_F \quad \text{in} \quad \|B\|_F$$

$$\|B\| = \sqrt{\sum_{i=1}^m \sum_{j=1}^n |b_{ij}|^2} \quad \text{in} \quad \langle B, C \rangle_F = \sum_{i=1}^m \sum_{j=1}^n b_{ij} \overline{c_{ij}} =$$

$$= \text{sled}(BC^*)$$

$$Q_1 Q_2^* = I$$

a.) Pokaži, da  $Q_1, Q_2$  unitarni  $\Rightarrow$   
 $\|Q_1^* A Q_2\|^2 = \|A\|^2$

$$\langle Q_1^* A Q_2, Q_1^* A Q_2 \rangle = \text{s}(Q_1^* A Q_2 (Q_1^* A Q_2)^*) =$$

$$= \text{s}(Q_1^* A Q_2 Q_2^* A^* Q_1) = \text{s}(Q_1^* A A^* Q_1) = \text{s}(Q_1^* A^* A Q_1) =$$

$$= \text{s}(A A^*) = \langle A, A \rangle = \|A\|^2 \quad \square$$

dauno tega smo  
dokazali  
 $\text{sled}(xy) = \text{sled}(yx)$ ,  
če so elementi iz polja.  
za elemente iz obsega,  
tj. ni komutativnosti,  
to ne velja.

Izlet: (Eckart & Young, 1936):

Naj bo  $A \in M_{m \times n}(\mathbb{C})$  in  
 $\text{rang}(A) = r, \quad r \leq \min\{m, n\}$ .

$$A = U \Sigma V^* \quad (\text{SVD})$$

Def:  $\Sigma_r :=$  matrica  $\Sigma$ , tje  $\sigma_{r+1}, \dots, \sigma_n$  zanesajo z 0

$A_r := U \Sigma_r V^*$  se med vsemi matricami  
rang  $\leq r$  najbolj privlega  $A$ .

Privlečenost  $A$  glede na  
normo.

Uporaba: slikna velikosti  $m \times n \rightarrow m \cdot n$  podatkov.

let  $A \in M^{m \times n}(\mathbb{R})$ .

$$A = U \Sigma V^T \rightsquigarrow A_k = U \sum_k V^T = \sigma_1 u_1 v_1 + \dots + \sigma_k u_k v_k$$

Primer:  $m=n=200$   $k=50$

$$\frac{k(m+n+1)}{mn} = \frac{50401}{4000} \approx 0,5$$

$N$  poisci ortogonalno matrico, ki je najbližje dan: realni  $n \times n$  matrici  $A$ !

naj bo to  $Q$ .  $\|A - Q\|_F = \|U \Sigma V^T - Q\|_F =$

$$= \|u^T u \Sigma V^T V - u^T Q V\|_F = \|\Sigma - \underbrace{u^T Q V}_{\substack{\text{ortog.} \\ \text{ker je} \\ \text{produkt} \\ \text{ortog.}}}\|_F = \|\Sigma - \tilde{Q}\|_F$$

$$\|\Sigma - \tilde{Q}\|_F^2 = \langle \Sigma - \tilde{Q}, \Sigma - \tilde{Q} \rangle =$$

$$= \sum_{k=1}^n \sum_{j=1}^n \begin{cases} (\sigma_k + \tilde{q}_{kk})^2; & k=j \\ 0 - \tilde{q}_{kj}^2; & k \neq j \end{cases} =$$

$$= \sum_{k=1}^n \sum_{j=1}^n \begin{cases} \sigma_k^2 + 2\sigma_k \tilde{q}_{kk} + \tilde{q}_{kk}^2; & k=j \\ \tilde{q}_{kj}^2; & k \neq j \end{cases} =$$

$$= \sum_{k=1}^n \left( \underbrace{\sigma_k^2 + 2\sigma_k \tilde{q}_{kk}}_{\text{sled}(\Sigma^2)} + \underbrace{\sum_{j=1}^n \tilde{q}_{kj}^2}_{\text{sled}(Q Q^T) = n} \right) =$$

$$= \text{sled}(\Sigma^2) + n - 2 \sum_{k=1}^n \tilde{q}_{kk} \text{ čim večje}$$

$\downarrow$   
 kerže  $\tilde{Q}$  z  $0N$  stolpci,  
 je lahko  $\tilde{Q}$  enake  $I$ :  
 $\tilde{Q} = I$ , sedaj lahko izrazimo

$Q$ :

$$U \cdot \tilde{Q} = U^T Q V \quad / \cdot V^T$$

$$I = U^T Q V$$

$$\underbrace{UV = Q}_{\checkmark}$$

$N$  linearno preslitavo  $L: V \rightarrow U$

velja  $L^*L$  je sebiadjungirana:

$$L^*L = (L^*L)^* \Rightarrow L^*L \text{ normalna}$$

$\exists$  ONB za  $V$  iz lastnih vektorjev za  $L^*L$

$\text{Ker}(L^*L) =$  lastni podprostor za lastno 0

ONB za  $\text{Ker}(L^*L)$  obstaja:  $\{v_{r+1}, \dots, v_n\}$   
 $\downarrow = \text{Ker}(L)$

$\text{Im}(L^*L)$  je vsota vseh ostalih lapodpr.

ONB za  $\text{Im} L^*L$   $\exists$ :  $\{v_1, \dots, v_r\}$   
 $\stackrel{\text{LAP}}{=} \text{Ker}(L)^\perp \stackrel{\text{LAP}}{=} \text{Im}(L^*)$

$\forall i \in \{1, \dots, r\}$ :  $u_i = \frac{1}{\sqrt{\lambda_i}} L v_i$  + dopolnimo do ONB za prostoru  $U$ .

$$\Rightarrow u_1, \dots, u_r \text{ so ONB za } \operatorname{Im} L = \operatorname{Im} L L^*$$

$$u_{r+1}, \dots, u_n \text{ so ONB za } (\operatorname{Im} L)^\perp = \operatorname{Ker} L^*$$

torej če preslitavi  $L$  privedemo nazaj to glede na ONB, dobimo pravokotno diagonalno matriko s singularnimi vrednostmi po diagonali.

$\Rightarrow$  singularni vrednici ni evolucije.

## N POLARNA DEKOMPOZICIJA Matrike

1.) let  $A$  kvadratna

$$A = U \cdot P \Rightarrow U \text{ unitarna, } P \geq 0$$

$$A \in \mathbb{C}^{n \times n}$$

$$A = Q_1 \Sigma Q_2^* = \underbrace{Q_1 Q_2^*}_{U} \underbrace{\Sigma}_{P} Q_2^*$$

(produkt unitarnih  $\Rightarrow$  unitarna  $\checkmark$ )

$$\underline{P \geq 0}$$

$$P = Q_2 \Sigma Q_2^* \xrightarrow{\text{diag}}$$

$$P^* = Q_2 \Sigma^* Q_2^* = P$$

$$\text{torej } P \sim \Sigma \Rightarrow \operatorname{lav}(P) \text{ so } \geq 0$$

$$\hookrightarrow P \geq 0$$

2.  $A$   $m \times n$  natural: BSS  $m \geq n$

$$A = Q_1 \begin{bmatrix} \Sigma \\ 0 \end{bmatrix} Q_2^*$$

$$Q_1 = \begin{bmatrix} \underbrace{Q_1^1}_{n} & \underbrace{Q_1^2}_{m-n} \end{bmatrix}$$

bloco

partida por, da

$$P = V \Sigma V^*$$

$$A = U \cdot P$$

$$U = \underbrace{Q_1^1}_{\text{ni unitaria,}} V^*$$

toda fe  
ve o seu

$$U^* U = I$$

$$U U^* \neq I$$

