

Math 218D Problem Session

Week 12

1. Rules of vector SVD

Which of the following $A = \sigma_1 u_1 v_1^T + \cdots + \sigma_r u_r v_r^T$ are valid singular value decompositions? Why/why not?

a) $A = 1(1, 0)(1, 0)^T + 3(0, 1)(0, 1)^T$

b) $A = 4(1, 0)(0, 1)^T + 3(0, 1)(1, 0)^T$

c) $A = 3(1, -1)(1, 0)^T + 2(1, 1)(0, 1)^T$

d) $A = -3(1/\sqrt{2}, -1/\sqrt{2}, 0)(1, 0)^T + 2(0, 0, 1)(0, 1)^T$

e) $A = 3(-1/\sqrt{2}, 1/\sqrt{2}, 0)(1, 0)^T + 2(0, 0, 1)(0, 1)^T$

f) $A = 5(1, 0, 0)(0, 1)^T + 3(0, 1, 0)(1, 0)^T + 2(0, 0, 1)(0, 1)^T$

3. Computing the vector SVD

To find the vector SVD of a matrix A :

- (1) Find the non-zero eigenvalues $\lambda_1 \geq \lambda_2 \geq \dots \lambda_r > 0$ of $A^T A$.
- (2) Find an orthonormal basis of each of the λ_i eigenspace of $A^T A$. Listed in order of decreasing eigenvalue, these are the right singular vectors v_1, \dots, v_r .
- (3) For $i = 1, \dots, r$, set $\sigma_i = \sqrt{\lambda_i}$ and $u_i = \frac{Av_i}{\sigma_i}$. These are the singular values and left singular vectors.
- (4) Write $A = \sigma_1 u_1 v_1^T + \dots + \sigma_r u_r v_r^T$.

Compute the vector SVD of each of the following matrices:

a) $A = \begin{pmatrix} 0 & -1 \\ 3 & 0 \end{pmatrix}$

b) $A = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & -3 & 0 & 0 \end{pmatrix}$

c) $A = \begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix}$

4. Computing the matrix SVD

To find the matrix SVD $A = U\Sigma V^T$ of a matrix A :

- (1) Find the symmetric diagonalization VDV^T of $A^T A$, where the eigenvalues are listed in decreasing order: $\lambda_1 \geq \dots \geq \lambda_n$. The rank r of A is the same as the number of non-zero eigenvalues of $A^T A$ (counted with multiplicity).
- (2) The columns of V are the right singular vectors v_1, \dots, v_r , followed by an orthonormal basis v_{r+1}, \dots, v_n of $\text{Nul}(A)$.
- (3) For $i = 1, \dots, r$, set $\sigma_i = \sqrt{\lambda_i}$ and $u_i = \frac{Av_i}{\sigma_i}$. These are the singular values and left singular vectors.
- (4) We still need the vectors u_{r+1}, \dots, u_m : find these by computing an orthonormal basis of $\text{Nul}(A^T)$ (using RREF to find a basis, Gram–Schmidt to replace it with an orthonormal basis).
- (5) Finally, the matrix U is the matrix with columns u_1, \dots, u_m , the matrix V was found in (1), and Σ has its first r diagonal entries as $\sigma_1, \dots, \sigma_r$ and the remaining entries of Σ being zero.

Compute the matrix SVD of each of the following matrices:

a) $A = \begin{pmatrix} 0 & -1 \\ 3 & 0 \end{pmatrix}$

b) $A = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & -3 & 0 & 0 \end{pmatrix}$

c) $A = \begin{pmatrix} 1 & 2 \\ 1 & 2 \\ 2 & 4 \end{pmatrix}$

5. Sums of rank 1 matrices

This final problem is not about SVDs, but just about sums of rank one matrices.

- a) Without computing A , explain why

$$A = (1, 2, 1)(1, 1)^T + (1, -1, 1)(-1, 1)^T$$

is a rank 2 matrix.

Hint: compute $A(1, 1)$ and $A(-1, 1)$, and use this to show that $(1, 2, 1)$ and $(1, -1, 1)$ are in the column space of A .

- b) If $A = u_1 v_1^T + \cdots + u_r v_r^T$ for some vectors $u_i \in \mathbf{R}^m$ and $v_j \in \mathbf{R}^n$, explain why the rank of A is at most r .

Hint: Show that the subspace $\text{Col}(A)$ is contained in the span $\text{Span}\{u_1, \dots, u_r\}$, which is at most r -dimensional.

- c) If the vectors $u_1, \dots, u_r \in \mathbf{R}^m$ are a linearly independent set of vectors, and the vectors $v_1, \dots, v_r \in \mathbf{R}^n$ are also linearly independent, prove that

$$A = u_1 v_1^T + \cdots + u_r v_r^T$$

has rank equal to r .

Hint: Show that there is a vector $v \in \mathbf{R}^n$ which is orthogonal to v_2, \dots, v_r , but $v_1^T v \neq 0$. Compute Av , and use this to show that $u_1 \in \text{Col}(A)$. The same idea shows that u_2, \dots, u_r are all in $\text{Col}(A)$.