

Name: **Solutions**

1. Let $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$. Find the full SVD for A . Find the pseudoinverse A^+ of A .

Solution:

Step 1. The first step is to compute $A^T A$ and find its eigenvalues and eigenvectors.

$$A^T A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 1 \end{bmatrix}.$$

To compute the eigenvalues, we find the characteristic polynomial (note that it will have root 0, since the matrix is singular, since it has the same rank as A , that is 2).

$$\begin{aligned} \det(A^T A - \lambda I) &= \begin{vmatrix} 1-\lambda & 1 & 0 \\ 1 & 2-\lambda & 1 \\ 0 & 1 & 1-\lambda \end{vmatrix} = (\text{cofactors in first row}) \\ &= (1-\lambda)((2-\lambda)(1-\lambda) - 1) - 1((1-\lambda) - 0) \\ &= (1-\lambda)(\lambda^2 - 3\lambda + 2 - 1 - 1) = (1-\lambda)(3-\lambda)\lambda. \end{aligned}$$

Nonzero eigenvalues therefore are $\lambda_1 = 3$ and $\lambda_2 = 1$. Respectively, the singular values are $\sigma_1 = \sqrt{\lambda_1} = \sqrt{3}$, $\sigma_2 = \sqrt{\lambda_2} = 1$. Now we find the respective eigenvectors. They are going to be singular vectors v_1, v_2 for A .

For $\lambda_1 = 3$, we get

$$\begin{bmatrix} 1-3 & 1 & 0 \\ 1 & 2-3 & 1 \\ 0 & 1 & 1-3 \end{bmatrix} v_1 = 0,$$

$$\begin{bmatrix} -2 & 1 & 0 \\ 1 & -1 & 1 \\ 0 & 1 & -2 \end{bmatrix} v_1 = 0,$$

solving which we get v_1 proportional to $\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$. Since we want a vector of unit

length, we take $v_1 = \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$.

For $\lambda_1 = 1$, we get

$$\begin{bmatrix} 1-1 & 1 & 0 \\ 1 & 2-1 & 1 \\ 0 & 1 & 1-1 \end{bmatrix} v_2 = 0,$$

$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -2 & 1 \\ 0 & 1 & 0 \end{bmatrix} v_2 = 0,$$

solving which we get v_2 proportional to $\begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$. Since we want a vector of unit

length, we take $v_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$.

Step 2. Now u_1, u_2 are computed from $Av_i = \sigma_i u_i$.

$$u_1 = \frac{1}{\sigma_1} Av_1 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = \frac{1}{3\sqrt{2}} \begin{bmatrix} 3 \\ 3 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix},$$

$$u_2 = \frac{1}{\sigma_2} Av_2 = \frac{1}{1} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}.$$

At this point we now have equality

$$\begin{aligned} A &= \sigma_1 u_1 v_1^T + \sigma_2 u_2 v_2^T \\ &= \sqrt{3} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \frac{1}{\sqrt{6}} \begin{bmatrix} 1 & 2 & 1 \end{bmatrix} + 1 \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 0 & -1 \end{bmatrix}. \end{aligned}$$

If we want the full SVD $A = U\Sigma V^T$ (which we do), then we proceed to step 2.

Step 3. We have to complete u_1, u_2 to an orthonormal basis of \mathbb{R}^2 (2 is the respective dimension of A). However, we already have two vectors u_1, u_2 , so there is nothing to further to do with u_i 's.

Step 4. We have to complete v_1, v_2 to an orthonormal basis of \mathbb{R}^3 (3 is the respective dimension of A). Solve system given by $v \cdot v_1 = 0$, $v \cdot v_2 = 0$ (we omit square roots since they don't change equality of the respective dot products to 0):

$$\begin{aligned} x + 2y + z &= 0, \\ x - z &= 0. \end{aligned}$$

Generally, we have to solve this and construct an orthonormal basis of the solution space. In this case, since the solution space is 1-dimensional, we only

have to normalize a solution. Solving the system, we get solution space $\begin{bmatrix} z \\ -z \\ z \end{bmatrix}$.

For a length 1 vector v_3 we can take $v_3 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$.

Organizing vectors u_1, u_2 into columns of a matrix U , and vectors v_1, v_2, v_3 into columns of a matrix V , we obtain full SVD:

$$A = U\Sigma V^T = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{6} & 2/\sqrt{6} & 1/\sqrt{6} \\ 1/\sqrt{2} & 0 & -1/\sqrt{2} \\ 1/\sqrt{3} & -1/\sqrt{3} & 1/\sqrt{3} \end{bmatrix}.$$

To find the pseudoinverse A^+ , we find Σ^+ (we invert singular values and transpose A):

$$\Sigma^+ = \begin{bmatrix} 1/\sqrt{3} & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

and obtain the product:

$$\begin{aligned} A^+ &= V\Sigma^+U^T = \\ &= \begin{bmatrix} 1/\sqrt{6} & 1/\sqrt{2} & 1/\sqrt{3} \\ 2/\sqrt{6} & 0 & -1/\sqrt{3} \\ 1/\sqrt{6} & -1/\sqrt{2} & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} 1/\sqrt{3} & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}. \end{aligned}$$

Out of curiosity, actually perform the multiplication.

$$\begin{aligned} A^+ &= \begin{bmatrix} 1/\sqrt{6} & 1/\sqrt{2} & 1/\sqrt{3} \\ 2/\sqrt{6} & 0 & -1/\sqrt{3} \\ 1/\sqrt{6} & -1/\sqrt{2} & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} 1/\sqrt{3} & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \\ &= \begin{bmatrix} 1/3\sqrt{2} & 1/\sqrt{2} \\ 2/3\sqrt{2} & 0 \\ 1/3\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{6} + \frac{1}{2} & \frac{1}{6} - \frac{1}{2} \\ \frac{1}{3} & \frac{1}{3} \\ \frac{1}{6} - \frac{1}{2} & \frac{1}{6} + \frac{1}{2} \end{bmatrix} = \begin{bmatrix} \frac{2}{3} & -\frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} \\ -\frac{1}{3} & \frac{2}{3} \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 2 & -1 \\ 1 & 1 \\ -1 & 2 \end{bmatrix}. \end{aligned}$$

REMARK 1. In Steps 1 and 2, we could hold off on normalizing: for example, take

$v'_1 = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$ and $u'_1 = Av_1 = \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix}$, and then normalize after that: $v_1 = \frac{v'_1}{\|v'_1\|}$ and $u_1 = \frac{u'_1}{\|u'_1\|}$. The result is the same, but that way we avoid dragging square roots around.

REMARK 2. We also could have worked with AA^T in Step 1, obtaining u_1, u_2 ; in Step 2 we would then get v_i by applying A^T to u_i . For illustration purposes, let's the first two steps that way now.

Step 1. $A^T A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix},$

$$\det(AA^T - \lambda I) = \begin{vmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{vmatrix} = \lambda^2 - 4\lambda + 3 = (\lambda - 3)(\lambda - 1).$$

(Note the nonzero eigenvalues of $A^T A$ and AA^T are the same.) We get eigenvectors:

$$\begin{aligned} \begin{bmatrix} 2 - 3 & 1 \\ 1 & 2 - 3 \end{bmatrix} u_1 &= 0, & \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} u_1 &= 0, & u_1 &= \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}. \\ \begin{bmatrix} 2 - 1 & 1 \\ 1 & 2 - 1 \end{bmatrix} u_2 &= 0, & \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} u_2 &= 0, & u_2 &= \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}. \end{aligned}$$

Notice the same result in Step 2 in the original solution.

Step 2. Now find v_i from $A^T u_i = \sigma_i v_i$:

$$v_1 = \frac{1}{\sigma_1} A^T u_1 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}.$$

$$v_2 = \frac{1}{\sigma_2} A^T u_2 = \frac{1}{1} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}.$$

Again, note the same result as in Step 1 in the original solution. From this point, Step 3 and Step 4 are the same.