

Math 2940 Worksheet Least Squares, SVD

Week 14 December 5th, 2019

This worksheet covers material from **Sections 6.4**, **6.5**, and **7.4**. Please work in collaboration with your classmates to complete the following exercises - this means sharing ideas and asking each other questions.

Question 1.

(a) Apply the Gram-Schmidt process to the columns of $A = \begin{bmatrix} 5 & 9\\ 1 & 7\\ -3 & -5\\ 1 & 5 \end{bmatrix}$

(b) Use your answer from part (a) to compute a QR factorization of A.

(c) Use your answer from part (b) to find a least-squares solution of $A\boldsymbol{x} = \boldsymbol{b}$ for $\boldsymbol{b} = \begin{bmatrix} -1\\ 2\\ 1\\ 6 \end{bmatrix}$.

Question 2. Suppose A = QR, where Q is an $m \times n$ matrix with orthogonal columns and R is an $n \times n$ matrix. Show that if the columns of A are linearly dependent, then R cannot be invertible.

Question 3. Let
$$A = \begin{bmatrix} 1 & -3 & -3 \\ 1 & 5 & 1 \\ 1 & 7 & 2 \end{bmatrix}$$
 and $b = \begin{bmatrix} 5 \\ -3 \\ -3 \end{bmatrix}$.

Find a least-squares solution of $A\mathbf{x} = \mathbf{b}$, and compute the associated least-squares error.

Question 4.

Consider the set of vectors with unit length in \mathbb{R}^2 , denoted by $D = \{ v \in \mathbb{R}^2 : ||v|| = 1 \}$. Describe the image of D under the linear transformation

$$T(oldsymbol{v}) = egin{bmatrix} 3 & 2 \ 2 & 3 \ 2 & -2 \end{bmatrix} oldsymbol{v}, \qquad oldsymbol{v} \in D$$

Sketch the set D and its image T(D). [Hint: use the SVD of the matrix of T.]

Answer to Question 1.

(a) We want to take the vectors $\boldsymbol{x}_1 = \begin{bmatrix} 5\\1\\-3\\1 \end{bmatrix}$ and $\boldsymbol{x}_2 = \begin{bmatrix} 9\\7\\-5\\5 \end{bmatrix}$, and convert them into orthogonal

vectors \boldsymbol{v}_1 and \boldsymbol{v}_2 .

For the first vector, we choose

$$\boldsymbol{v}_1 = \boldsymbol{x}_1 = egin{bmatrix} 5\\1\\-3\\1\end{bmatrix}$$

To make sure that v_2 is orthogonal to v_1 , we want to take x_2 and subtract the component in the v_1 direction.

$$\boldsymbol{v}_{2} = \boldsymbol{x}_{2} - \operatorname{proj}_{\boldsymbol{v}_{1}}(\boldsymbol{x}_{2}) = \boldsymbol{x}_{2} - \frac{\boldsymbol{v}_{1} \cdot \boldsymbol{v}_{2}}{\boldsymbol{v}_{1} \cdot \boldsymbol{v}_{1}} \boldsymbol{v}_{1}$$

$$= \begin{bmatrix} 9\\7\\-5\\5 \end{bmatrix} - \frac{\begin{bmatrix} 5\\1\\-3\\1 \end{bmatrix} \cdot \begin{bmatrix} 9\\7\\-5\\5 \end{bmatrix}}{\begin{bmatrix} -5\\5 \end{bmatrix}} = \begin{bmatrix} 9\\7\\-5\\5 \end{bmatrix} - \frac{72}{36} \begin{bmatrix} 5\\1\\-3\\1 \end{bmatrix} = \begin{bmatrix} -1\\5\\1\\3 \end{bmatrix}$$

Now that we have orthogonal vectors v_1 and v_2 , we can find their magnitudes:

$$\|\boldsymbol{v}_1\| = \sqrt{5^2 + 1^2 + (-3)^2 + 1^2} = \sqrt{36} = 6$$
$$\|\boldsymbol{v}_2\| = \sqrt{(-1)^2 + 5^2 + 1^2 + 3^2} = \sqrt{36} = 6$$

So the orthonormal basis is

$$egin{aligned} m{u}_1 &= rac{m{v}_1}{\|m{v}_1\|} = egin{bmatrix} 5/6 \ 1/6 \ -3/6 \ 1/6 \end{bmatrix} \ m{u}_2 &= rac{m{v}_2}{\|m{v}_2\|} = egin{bmatrix} -1/6 \ 5/6 \ 1/6 \ 3/6 \end{bmatrix} \end{aligned}$$

(b) To find Q, we can just combine the colums we found in part (a)

$$Q = \begin{bmatrix} 5/6 & -1/6\\ 1/6 & 5/6\\ -3/6 & 1/6\\ 1/6 & 3/6 \end{bmatrix}$$

To find R, we note that

$$A = QR$$

Multiplying both sides on the left by Q^T ,

$$Q^T A = Q^T Q R = I R = R$$

So we can compute

$$R = Q^T A = \begin{bmatrix} 5/6 & 1/6 & -3/6 & 1/6 \\ -1/6 & 5/6 & 1/6 & 3/6 \end{bmatrix} \begin{bmatrix} 5 & 9 \\ 1 & 7 \\ -3 & -5 \\ 1 & 5 \end{bmatrix} = \begin{bmatrix} 6 & 12 \\ 0 & 6 \end{bmatrix}$$

Therefore the QR factorization of A is:

A = QR =	$\begin{bmatrix} 5/6\\ 1/6 \end{bmatrix}$	$-1/6 \\ 5/6$	[6	12]
	$\begin{bmatrix} -3/6\\ 1/6 \end{bmatrix}$	$\frac{1/6}{3/6}$	0	$\begin{bmatrix} 12\\ 6 \end{bmatrix}$

(c) First we note that the least-squares system is equivalent to solving

$$A \boldsymbol{x} = Q Q^T \boldsymbol{b}$$

Using the QR factorization,

$$QR\boldsymbol{x} = QQ^T \boldsymbol{b}$$

Multiplying both sides on the left by Q^T ,

$$Q^T Q R \boldsymbol{x} = Q^T Q Q^T \boldsymbol{b}$$

 $R \boldsymbol{x} = Q^T \boldsymbol{b}$

First we compute

$$Q^{T}\boldsymbol{b} = \begin{bmatrix} 5/6 & 1/6 & -3/6 & 1/6 \\ -1/6 & 5/6 & 1/6 & 3/6 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \\ 1 \\ 6 \end{bmatrix} = \begin{bmatrix} \frac{-5+2-3+6}{6} \\ \frac{1+10+1+18}{6} \end{bmatrix} = \begin{bmatrix} 0 \\ 5 \end{bmatrix}$$

So we want to solve the system

$$R\boldsymbol{x} = Q^T \boldsymbol{b}$$
$$\begin{bmatrix} 6 & 12 \\ 0 & 6 \end{bmatrix} \boldsymbol{x} = \begin{bmatrix} 0 \\ 5 \end{bmatrix}$$

Setting up the augmented matrix and row reducing,

$$\begin{bmatrix} 6 & 12 & 0 \\ 0 & 6 & 5 \end{bmatrix} \xrightarrow{\text{R1}-2 \cdot \text{R2}} \begin{bmatrix} 6 & 0 & | & -10 \\ 0 & 6 & | & 5 \end{bmatrix} \xrightarrow{\frac{1}{6} \cdot \text{R2}} \begin{bmatrix} 6 & 0 & | & -10 \\ 0 & 1 & | & 5/6 \end{bmatrix} \xrightarrow{\frac{1}{6} \cdot \text{R1}} \begin{bmatrix} 1 & 0 & | & -10/6 \\ 0 & 1 & | & 5/6 \end{bmatrix}$$

-10/6Therefore the least-squares solution is

5/6

Answer to Question 2. First, we know that

$$A = QR$$

Multiplying both sides on the left by Q^T ,

$$Q^T A = Q^T Q R = I R = R$$

Since the columns of A are linearly dependent, we know that there exists a non-zero vector \boldsymbol{x} such that $A\boldsymbol{x} = \boldsymbol{0}$. However, this means

$$R\boldsymbol{x} = Q^T A \boldsymbol{x} = Q^T (A \boldsymbol{x}) = Q^T \boldsymbol{0} = \boldsymbol{0}$$

Since there is a non-zero vector \boldsymbol{x} such that $R\boldsymbol{x} = \boldsymbol{0}$, this implies that R is not invertible.

Answer to Question 3. We can find least-squares solutions as solutions of the system $A^T A \boldsymbol{x} = A^T \boldsymbol{b}$.

We can compute

$$A^{T}A = \begin{bmatrix} 1 & 1 & 1 \\ -3 & 5 & 7 \\ -3 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & -3 & -3 \\ 1 & 5 & 1 \\ 1 & 7 & 2 \end{bmatrix} = \begin{bmatrix} 3 & 9 & 0 \\ 9 & 83 & 28 \\ 0 & 28 & 14 \end{bmatrix}$$

and

$$A^{T}\boldsymbol{b} = \begin{bmatrix} 1 & 1 & 1 \\ -3 & 5 & 7 \\ -3 & 1 & 2 \end{bmatrix} \begin{bmatrix} 5 \\ -3 \\ -5 \end{bmatrix} = \begin{bmatrix} -3 \\ -65 \\ -28 \end{bmatrix}$$

Setting up the augmented matrix for $A^T A \boldsymbol{x} = A^T \boldsymbol{b}$ and row reducing,

$$\begin{bmatrix} 3 & 9 & 0 & | & -3 \\ 9 & 83 & 28 & | & -65 \\ 0 & 28 & 14 & | & -28 \end{bmatrix} \xrightarrow{\frac{1}{3} \cdot \text{R1}} \begin{bmatrix} 1 & 3 & 0 & | & -1 \\ 9 & 83 & 28 & | & -65 \\ 0 & 28 & 14 & | & -28 \end{bmatrix} \xrightarrow{\frac{1}{4} \cdot \text{R3}} \begin{bmatrix} 1 & 3 & 0 & | & -1 \\ 9 & 83 & 28 & | & -65 \\ 0 & 2 & 1 & | & -2 \end{bmatrix}$$

$$\underbrace{ \begin{array}{c} \text{R2} - 9 \cdot \text{R1} \\ \text{R2} - 9 \cdot \text{R1} \\ \text{R2} \\ \text{R3} \\ \text{R3} \\ \text{R3} \\ \text{R2} \\ \text{R3} \\ \text{R3} \\ \text{R2} \\ \text{R3} \\$$

Since there is a free variable, we see that there are actually inifinitely many least-squares solutions to this problem.

To choose a specific one, we can set $x_3 = 0$ to get a least squares solution $\begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}$.

We can also check that $A\hat{x} = b$ exactly, so the associated least-squares error is zero.

Answer to Question 4. First we want to compute the SVD of the matrix $A = \begin{bmatrix} 3 & 2 \\ 2 & 3 \\ 2 & -2 \end{bmatrix}$. We compute

$$A^{T}A = \begin{bmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{bmatrix} \begin{bmatrix} 3 & 2 \\ 2 & 3 \\ 2 & -2 \end{bmatrix} = \begin{bmatrix} 17 \\ 8 \end{bmatrix}$$

8 17

To find the singular values of A, we can compute the eigenvalues of $A^T A$ as follows

$$\det(A^T A - \lambda I) = \det\left(\begin{bmatrix} 17 - \lambda & 8\\ 8 & 17 - \lambda \end{bmatrix}\right) = (17 - \lambda)^2 - 8^2 = 0$$
$$(17 - \lambda)^2 = 8^2$$
$$17 - \lambda = \pm 8$$
$$\lambda = 25, 9$$

So the singular values are $\sigma_1 = \sqrt{25} = 5$ and $\sigma_2 = \sqrt{9} = 3$, and $\Sigma = \begin{bmatrix} 5 & 0 \\ 0 & 3 \\ 0 & 0 \end{bmatrix}$.

To find the right singular vectors, we want to find the eigenvectors of $\vec{A^T}A$. For the first eigenvector,

$$A - 25I = \begin{bmatrix} -8 & 8\\ 8 & -8 \end{bmatrix} \sim \begin{bmatrix} 1 & -1\\ 0 & 0 \end{bmatrix}$$

Therefore the first right singular vector is proportional to $\begin{bmatrix} 1\\1 \end{bmatrix}$. Normalizing, we get

$$\boldsymbol{v}_1 = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

For the second eigenvector,

$$A - 9I = \begin{bmatrix} 8 & 8 \\ 8 & 8 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$$

Therefore the first right singular vector is proportional to $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$. Normalizing, we get

$$\boldsymbol{v}_2 = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

and therefore $V = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}$. We can compute the first two singular vectors as

$$\boldsymbol{u}_{1} = \frac{1}{\sigma_{1}}A\boldsymbol{v}_{1} = \frac{1}{5} \begin{bmatrix} 3 & 2\\ 2 & 3\\ 2 & -2 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2}\\ 1/\sqrt{2}\\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2}\\ 1/\sqrt{2}\\ 0 \end{bmatrix} \boldsymbol{u}_{2} = \frac{1}{\sigma_{2}}A\boldsymbol{v}_{2} = \frac{1}{3} \begin{bmatrix} 3 & 2\\ 2 & 3\\ 2 & -2 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2}\\ -1/\sqrt{2}\\ -1/\sqrt{2}\\ 4/\sqrt{2} \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 1/\sqrt{2}\\ -1/\sqrt{2}\\ 4/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1/3\sqrt{2}\\ -1/\sqrt{2}\\ 4/\sqrt{2} \end{bmatrix}$$

For the third singular vector, we need to find a vector orthogonal to u_1 and u_2 . This is equivalent to the system of equations

$$\begin{bmatrix} 1/3\sqrt{2} & -1/3\sqrt{2} & 4/3\sqrt{2} & | & 0 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 & | & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 4 & | & 0 \\ 1 & 1 & 0 & | & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 4 & | & 0 \\ 0 & 2 & -4 & | & 0 \end{bmatrix}$$
$$\sim \begin{bmatrix} 1 & 0 & 2 & | & 0 \\ 0 & 1 & -2 & | & 0 \end{bmatrix}$$

So we see that $x_1 = -2x_3$ and $x_2 = 2x_3$. Setting $x_3 = 1$, we see that $\begin{vmatrix} -2\\2\\1\end{vmatrix}$ is orthogonal to u_1 and u_2 . Normalizing,

$$oldsymbol{u}_3 = egin{bmatrix} -2/3 \ 2/3 \ 1/3 \end{bmatrix}$$

And therefore $U = \begin{bmatrix} 1/\sqrt{2} & 1/3\sqrt{2} & -2/3 \\ 1/\sqrt{2} & -1/3\sqrt{2} & 2/3 \\ 0 & 4/3\sqrt{2} & 1/3 \end{bmatrix}$. So the singular value decomposition is

$$A = U\Sigma V^{T} = \begin{bmatrix} 1/\sqrt{2} & 1/3\sqrt{2} & -2/3\\ 1/\sqrt{2} & -1/3\sqrt{2} & 2/3\\ 0 & 4/3\sqrt{2} & 1/3 \end{bmatrix} \begin{bmatrix} 5 & 0\\ 0 & 3\\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2}\\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}$$

We recall that the set $D = \{ v \in \mathbb{R}^2 : ||v|| = 1 \}$, which is just the unit circle in \mathbb{R}^2 :



To determine what the transformation T does to the set D, we will figure out what each individual part of the SVD of T does to the set D.

First, since $V^T = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}$ corresponds to a rotation, this won't change the set D. Then, $\Sigma = \begin{bmatrix} 5 & 0 \\ 0 & 3 \\ 0 & 0 \end{bmatrix}$ will transform D from a circle in \mathbb{R}^2 to an ellipse in \mathbb{R}^3 . The ellipse will actually lie entirely in the *xy*-plane, with a major axis of $\begin{bmatrix} 5\\0\\0 \end{bmatrix}$ and a minor axis of $\begin{bmatrix} 0\\3\\0 \end{bmatrix}$:



You can find an interactive version of this plot at https://www.math3d.org/6PgGSXbQ

Finally, $U = \begin{bmatrix} 1/\sqrt{2} & 1/3\sqrt{2} & -2/3 \\ 1/\sqrt{2} & -1/3\sqrt{2} & 2/3 \\ 0 & 4/3\sqrt{2} & 1/3 \end{bmatrix}$ will rotate the major axis to the vector $\boldsymbol{u}_1 = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix}$, and rotate the minor axis to the vector $\boldsymbol{u}_1 = \begin{bmatrix} 1/3\sqrt{2} \\ -1/3\sqrt{2} \\ 4/3\sqrt{2} \end{bmatrix}$:



You can find an interactive version of this plot at https://www.math3d.org/Vq5qsBSx