

## q14

**Due** Oct 14 at 9am**Points** 5**Questions** 5**Available** Oct 13 at 9am - Oct 14 at 9am 1 day**Time Limit** 20 Minutes

## Attempt History

	Attempt	Time	Score
<b>LATEST</b>	<a href="#">Attempt 1</a>	4 minutes	4.8 out of 5

⚠ Correct answers will be available on Oct 14 at 9:01am.

Score for this quiz: **4.8** out of 5

Submitted Oct 13 at 11:04pm

This attempt took 4 minutes.

## Question 1

1 / 1 pts

The nonzero singular values of a matrix  $A$  are 0.1, 0.1, 0.2, 0.5, 1. Let  $\hat{x}$  denote the minimum-norm LS minimizer of  $\|Ax - y\|_2$ . Let  $\hat{x}_3$  denote the corresponding truncated SVD solution for  $K = 3$ . Let  $u_4$  denote the 4th left singular vector of  $A$ . Which of the following statements are true?

☒ The difference  $\hat{x} - \hat{x}_3$  is in a 2-dimensional subspace.

☒ The solution  $\hat{x}_3$  is in a 3-dimensional subspace.

☒  $u_4' A \hat{x}_3 = 0$

☒ The solution  $\hat{x}$  is in a 5-dimensional subspace.

☐  $A \hat{x} = A \hat{x}_3$

☐  $u_4' A \hat{x} = 0$

☐  $A\hat{x}_3 = y$

☐  $A\hat{x} = y$

☒ An upper bound on  $\|\hat{x} - \hat{x}_3\|_2$  is  $10\|y\|_2$ .

☐  $\hat{x} = \hat{x}_3$

The MNLS solution is  $\hat{x} = A^+ y = \sum_{k=1}^5 1/\sigma_k v_k u_k' y$ . And  $\hat{x}_3 = A_3^+ y = \sum_{k=1}^3 1/\sigma_k v_k u_k' y$ . So  $A\hat{x}_3 \perp u_4$  and  $\hat{x}_3 \in \text{span}(\{v_1, v_2, v_3\})$ . The difference is  $\hat{x} - \hat{x}_3 = (A^+ - A_3^+)y$  so  $\|\hat{x} - \hat{x}_3\|_2 \leq \|A^+ - A_3^+\|_2 \|y\|_2 = 10\|y\|_2$

## Question 2

1 / 1 pts

Every projection matrix has a unitary eigendecomposition.

☐ True

☒ False

No, consider  $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ . It is idempotent but not normal.

Partial

## Question 3

0.8 / 1 pts

Which of the following statements about an orthogonal projection matrix are always true?

☒ It has a unitary eigendecomposition.

☐ Its range is a subspace.

☒ Its range is the orthogonal complement of some subspace.

☒ It is diagonalizable.

☐ None of these.

☒

Its range and its null space are both subspaces of the same vector space.

Because it is Hermitian it is square so its range and nullspace are both subspaces of  $\mathbb{F}^N$ .

☐ It is invertible.

It is Hermitian so normal and thus diagonalizable. The range of any matrix is a subspace. The range of any matrix  $A$  is the orthogonal complement of the subspace  $\mathcal{R}^\perp(A)$ .

#### Question 4

1 / 1 pts

Let

$$A = \begin{bmatrix} U_r & U_0 \end{bmatrix} \Sigma \begin{bmatrix} V_r & V_0 \end{bmatrix}'$$

where  $U_r$  and  $V_r$  have  $r$  columns where  $r$  denotes the rank of  $A$ . The projection matrix for the orthogonal complement of  $\mathcal{R}(A')$  is:

☐  $I - U_0 U_0'$

☐  $U_0 U_0'$

☒  $I - V_r V_r'$

☐ None of these

☐  $I - U_r U_r'$

☐  $U_r U_r'$

☐  $I - V_0 V_0'$

☒  $V_0 V_0'$

☐  $V_r V_r'$

$$P_{\mathcal{R}(A')}^\perp = P_{\mathcal{R}(V_0)} = V_0 V_0' = I - V_r V_r'$$

### Question 5

1 / 1 pts

Let

$$A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

and  $b = \begin{bmatrix} -2 \\ 4 \\ 6 \end{bmatrix}$ . An SVD is  $A = U \Sigma V'$ . The projection matrix for  $\mathcal{R}(A)$  is:

☐ None of these

☒ The product  $U_r U_r'$ , where  $U_r = \begin{bmatrix} u_1 & \dots & u_r \end{bmatrix}$  and  $A$  has rank  $r$ .

☒  $I_3$

☒  $AA^+$

This  $A$  has full rank so  $U_r U_r' = UU' = AA^+ = I$ .

Quiz Score: **4.8** out of 5