

# Linear System Theory

## Solution to Homework 2

1. Suppose that  $x \in \mathcal{N}(A)$ . Then  $Ax = 0 \Rightarrow EAx = 0 \Rightarrow x \in \mathcal{N}(EA)$ . Therefore,  $\mathcal{N}(A) \subseteq \mathcal{N}(EA)$ .

Conversely, suppose that  $x \in \mathcal{N}(EA)$ , i.e.  $EAx = 0$ . Since  $E$  is invertible, there exists  $E^{-1}$  such that  $E^{-1}EAx = Ax = 0$ . This implies  $x \in \mathcal{N}(A)$ , and therefore  $\mathcal{N}(EA) \subseteq \mathcal{N}(A)$ . Combine the previous result and we have  $\mathcal{N}(A) = \mathcal{N}(EA)$ .

Suppose that  $y \in \mathcal{R}(A)$ , i.e.  $\exists x \in \mathbb{R}^n$  such that  $y = Ax$ . Let  $z = F^{-1}x$  and  $y = AFF^{-1}x = AFz$ . This implies  $y \in \mathcal{R}(AF)$  and therefore,  $\mathcal{R}(A) \subseteq \mathcal{R}(AF)$ .

Conversely, suppose that  $y \in \mathcal{R}(AF)$ , i.e.  $\exists z \in \mathbb{R}^n$  such that  $y = AFz$ . Define  $x = Fz$  and  $y = Ax$ . This implies that  $y \in \mathcal{R}(A)$ , and therefore  $\mathcal{R}(AF) \subseteq \mathcal{R}(A)$ . Combine the previous result and we have  $\mathcal{R}(A) = \mathcal{R}(AF)$ .

2. (a) Suppose that  $\mathcal{N}(A) = \{0\}$ . If  $f$  is not one-to-one, then there exist  $x, y \in \mathbb{R}^n$ ,  $x \neq y$ , such that  $f(x) = Ax = f(y) = Ay$ . Therefore  $A(x - y) = 0$ . This is a contradiction to  $\mathcal{N}(A) = \{0\}$  because  $(x - y) \in \mathcal{N}(A)$  and  $x - y \neq 0$ . Therefore,  $\mathcal{N}(A) = \{0\}$  implies that  $f$  is one-to-one.

Conversely, suppose that  $f$  is one-to-one. If there exists  $z \neq 0$  such that  $z \in \mathcal{N}(A)$ , then for any  $x \in \mathbb{R}^n$ ,  $f(x + z) = A(x + z) = Ax = f(x)$  and  $x \neq x + z$ . This contradicts the hypothesis that  $f$  is one-to-one. Therefore, if  $f$  is one-to-one, then  $\mathcal{N}(A) = \{0\}$ .

- (b) Suppose that  $\mathcal{R}(A) = \mathbb{R}^m$ . Then for each  $y \in \mathbb{R}^m$ , there exists  $x \in \mathbb{R}^n$  such that  $y = Ax = f(x)$ . In other words,  $f$  is onto.

Conversely, if  $f$  is onto, then for every  $y \in \mathbb{R}^m$ , there exists  $x \in \mathbb{R}^n$  such that  $y = f(x) = Ax$ . Therefore  $y \in \mathcal{R}(A)$  and  $\mathcal{R}(A) = \mathbb{R}^m$ .

3. (a) The subspace spanned by  $x$  is  $\text{span}(x) = \{\alpha x | \alpha \in \mathbb{R}\}$ . The distance between  $y$  and any point in  $\text{span}(x)$  is  $D = \|y - \alpha x\|$ . Then the projection  $p \in \text{span}(x)$  is the point that  $D$  is minimal. Note that

$$D^2 = \|y - \alpha x\|^2 = (y - \alpha x)^T (y - \alpha x) = y^T y - 2\alpha x^T y + \alpha^2 x^T x$$

Since  $D^2$  is a convex quadratic function of  $\alpha$ , which has minimum at the point that the first derivative of  $D^2$  w.r.t.  $\alpha$  vanishes, i.e.

$$0 = \frac{dD^2}{d\alpha} = 2\alpha x^T x - 2x^T y$$

which implies  $\alpha = \frac{x^T y}{x^T x}$  and  $p = \alpha x = \frac{x^T y}{x^T x} x$ .

(b)

$$\begin{aligned}
0 &\leq (y-p)^T(y-p) \\
&= \left(y - \frac{x^T y}{x^T x} x\right)^T \left(y - \frac{x^T y}{x^T x} x\right) = y^T y - 2 \frac{x^T y}{x^T x} x^T y + \left(\frac{x^T y}{x^T x}\right)^2 x^T x \\
&= y^T y - \frac{(x^T y)^2}{x^T x}
\end{aligned}$$

Hence

$$|x^T y| \leq \sqrt{x^T x} \sqrt{y^T y} = \|x\| \|y\|$$

4. Since the columns of  $U$  are orthonormal, we have  $n \geq k$ . Let  $V \in \mathbb{R}^{n \times (n-k)}$  such that  $Q = [U, V]$  is an orthogonal matrix. Then

$$\|x\|^2 = \|Q^T x\|^2 = \left\| \begin{bmatrix} U^T x \\ V^T x \end{bmatrix} \right\|^2 = \|U^T x\|^2 + \|V^T x\|^2, \quad \forall x \in \mathbb{R}^n$$

Hence  $\|U^T x\| \leq \|x\|$  for all  $x \in \mathbb{R}^n$ .

Based on the previous analysis, the equality holds when  $V^T x = 0$ . In other words,  $x$  is orthogonal to each column of  $V$ . However, the columns of  $U$  are orthogonal to the columns of  $V$ . This implies that  $x$  is a linear combination of the columns of  $U$ , i.e.  $x \in \mathcal{R}(U)$ . Notice that if  $n = k$ , then  $\mathcal{R}(U) = \mathbb{R}^n$ . Hence for all  $x \in \mathbb{R}^n$ ,  $\|U^T x\| = \|x\|$ .

5. (a) Let  $P$  be a projection matrix. Then  $(I - P)^T = I - P^T = I - P$ , and  $(I - P)^2 = (I - P)(I - P) = I - 2P + P^2 = I - P$ . Hence  $I - P$  is also a projection matrix.
- (b)  $(UU^T)^T = UU^T$  and  $(UU^T)^2 = UU^T UU^T = UU^T$ . This is because  $U^T U = I$ . Hence  $UU^T$  is a projection matrix.
- (c)  $(A(A^T A)^{-1} A^T)^T = A(A^T A)^{-1} A^T$  and  $(A(A^T A)^{-1} A^T)^2 = (A(A^T A)^{-1} A^T)(A(A^T A)^{-1} A^T) = A(A^T A)^{-1} A^T$ . Hence  $A(A^T A)^{-1} A^T$  is a projection matrix.
- (d) Suppose that  $z \in \mathcal{R}(P)$ , i.e. there exists  $w \in \mathbb{R}^n$  such that  $z = Pw$ . Then

$$\|x - z\|^2 = \|x - Px + Px - z\|^2 = \|x - Px\|^2 + \|Px - z\|^2 + 2(x - Px)^T(Px - z).$$

Notice that

$$(x - Px)^T(Px - z) = (x - Px)^T P(x - w) = x^T (I - P)P(x - w) = x^T (P - P^2)(x - w) = 0$$

Thus

$$\|x - z\|^2 = \|x - Px\|^2 + \|Px - z\|^2 \geq \|x - Px\|^2, \quad \forall z \in \mathcal{R}(P)$$

This implies that  $Px$  is the closest point in  $\mathcal{R}(P)$  to  $x$ , i.e.  $Px$  the projection of  $x$  on  $\mathcal{R}(P)$ .