

# Linear Algebra

## Problem Set 6 Solution

Spring 2016

### 1. (20pts)

(a)

$$f = ae^x + be^{-x}, f' = ae^x - be^{-x}$$

$$[f]_{\beta} = \begin{bmatrix} a \\ b \end{bmatrix}, [f']_{\beta} = \begin{bmatrix} a \\ -b \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} a \\ -b \end{bmatrix}, \quad [D]_{\beta} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

(b)

$$f = ae^x + be^{-x} = \frac{a+b}{2}(e^x + e^{-x}) + \frac{a-b}{2}(e^x - e^{-x})$$

$$[f]_{\gamma} = \begin{bmatrix} (a+b)/2 \\ (a-b)/2 \end{bmatrix}$$

$$[f]_{\beta} = \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} (a+b)/2 \\ (a-b)/2 \end{bmatrix} = P[f]_{\gamma},$$

$$P = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}, Q = P^{-1} = \begin{bmatrix} 0.5 & 0.5 \\ 0.5 & -0.5 \end{bmatrix}$$

(c)

$$[D]_{\beta}[f]_{\beta} = [f']_{\beta}$$

$$[D]_{\beta}P[f]_{\gamma} = P[f']_{\gamma} = P[D]_{\gamma}[f]_{\gamma}$$

$$[D]_{\gamma} = Q[D]_{\beta}P$$

(d)

$$\begin{array}{ccc}
 f_{\beta} & \xrightarrow{D_{\beta}} & D(f)_{\beta} \\
 \downarrow & & \downarrow \\
 [f]_{\beta} & \xrightarrow{[D]_{\beta}} & [D(f)]_{\beta} \\
 \\ 
 Q_{\beta} \downarrow \uparrow P_{\beta} & & P_{\beta} \uparrow \downarrow Q_{\beta} \\
 [f]_{\gamma} & \xrightarrow{[D]_{\gamma}} & [D(f)]_{\gamma}
 \end{array}$$

**2.(25pts)**

(a)

$S$  and  $T$  are subspaces of  $V$ , so  $0 \in S, 0 \in T$ .

$$0 = 0 + 0 \in S + T$$

$$x \in S, y \in T$$

Let  $x_1, x_2 \in x$  and  $y_1, y_2 \in y$  and scalar  $\alpha$ .

$$\alpha x_1 + x_2 \in x$$

$$\alpha y_1 + y_2 \in y$$

$$x_1 + y_1 \in S + T$$

$$x_2 + y_2 \in S + T$$

$$\alpha(x_1 + y_1) + (x_2 + y_2) = (\alpha x_1 + x_2) + (\alpha y_1 + y_2) \in S + T$$

(b)

If  $V \in S + T, v = s_1 + t_1$ , so  $v \in \text{span}\{S \cup T\} \rightarrow S + T \subseteq \text{span}\{S \cup T\}$

If  $V \in \text{span}\{S \cup T\}$

$$v = s_1 + t_1 + s_2 + t_2 + \dots + s_n + t_n = (s_1 + s_2 + \dots + s_n) + (t_1 + t_2 + \dots + t_n) \in S + T$$

$$\rightarrow \text{span}\{S \cup T\} \subseteq S + T$$

Hence,  $S + T = \text{span}\{S \cup T\}$

(c)

$$x \in S, y \in S^\perp$$

Let  $x_1, x_2 \in x$  and  $y_1, y_2 \in y$

$$v = x_1 + y_1 = x_2 + y_2 \rightarrow x_1 - x_2 = y_2 - y_1$$

$$x_1 - x_2 \in S, y_2 - y_1 \in S^\perp \rightarrow (x_1 - x_2) \text{ and } (y_2 - y_1) \in S \cap S^\perp = \{0\}$$

So,  $x_1 = x_2, y_1 = y_2$ , there exists a unique vector  $x$  and vector  $y$ .

(d)

$$S = \text{span}\{x_1, x_2, \dots, x_m\}$$

$$T = \text{span}\{y_1, y_2, \dots, y_n\}$$

$$\text{Suppose } A = [x_1 \ \dots \ x_m \ y_1 \ \dots \ y_n]$$

$$\dim(S + T) = \dim(C(A)) = \text{rank}(A)$$

Consider vector  $c$  in  $N(A), Ac = 0$

$$c_1 x_1 + c_2 x_2 + \dots + c_m x_m + c_{m+1} y_1 + c_{m+2} y_2 + \dots + c_{m+n} y_n = 0$$

$$\text{Let } z = c_1 x_1 + c_2 x_2 + \dots + c_m x_m = -c_{m+1} y_1 - c_{m+2} y_2 - \dots - c_{m+n} y_n$$

Hence,  $z \in S$  and  $z \in T, z \in S \cap T$

反向推論亦成立，故 $S \cap T$ 為 $N(A)$ ,  $\dim(S \cap T) = \dim N(A)$

由秩零維數定理得

$$\dim N(A) = m + n - \text{rank}(A) \rightarrow \dim(S \cap T) = \dim S + \dim T - \dim(S + T)$$

(e)

From (c), (d)

$$\dim(S + S^\perp) = \dim S + \dim S^\perp - \dim(S \cap S^\perp) = \dim S + \dim S^\perp = \dim V$$

### 3.(35pts)

(a)

$$S = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\} = C(R), S^\perp = N(R^T)$$

$$\text{Let } R = \begin{bmatrix} 1 & 1 \\ 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}, R^T = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 0 & -1 & 1 \end{bmatrix}$$

$$S^\perp = N(R^T) = \text{span} \left\{ \begin{bmatrix} -1 \\ 0 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} \right\}$$

(b)

$$S = N(A) = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\} \rightarrow A = \begin{bmatrix} -1 & \alpha & 1 & 1 \\ 0 & \beta & 0 & 0 \end{bmatrix}, \alpha \text{ and } \beta \text{ are arbitrary constants.}$$

(c)

$$U = \begin{bmatrix} 1 & 1 \\ 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}, V = \begin{bmatrix} -1 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 0 \end{bmatrix}$$

$$P = U(U^T U)^{-1} U^T = \begin{bmatrix} 2/3 & 0 & 1/3 & 1/3 \\ 0 & 0 & 0 & 0 \\ 1/3 & 0 & 2/3 & -1/3 \\ 1/3 & 0 & -1/3 & 2/3 \end{bmatrix}$$

$$Q = V(V^T V)^{-1} V^T = \begin{bmatrix} 1/3 & 0 & -1/3 & -1/3 \\ 0 & 1 & 0 & 0 \\ -1/3 & 0 & 1/3 & 1/3 \\ -1/3 & 0 & 1/3 & 1/3 \end{bmatrix}$$

(d)

$$P = \begin{bmatrix} 2/3 & 0 & 1/3 & 1/3 \\ 0 & 0 & 0 & 0 \\ 1/3 & 0 & 2/3 & -1/3 \\ 1/3 & 0 & -1/3 & 2/3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$C(P) = \left\{ \begin{bmatrix} 2/3 \\ 0 \\ 1/3 \\ 1/3 \end{bmatrix}, \begin{bmatrix} 1/3 \\ 0 \\ 2/3 \\ -1/3 \end{bmatrix} \right\} = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\} = S$$

$$N(P) = C(P^T)^\perp = C(P)^\perp = S^\perp$$

(e)

$$v = x + y = a \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} + c \begin{bmatrix} -1 \\ 0 \\ 1 \\ 1 \end{bmatrix} + d \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$a = \frac{2}{3}, b = \frac{2}{3}, c = \frac{1}{3}, d = 1$$

$$x = \begin{bmatrix} 4/3 \\ 0 \\ 2/3 \\ 2/3 \end{bmatrix}, y = \begin{bmatrix} -1/3 \\ 1 \\ 1/3 \\ 1/3 \end{bmatrix}$$

(f)

$$C(B) \perp N(B^T)$$

$$C(B) = (N(B^T))^\perp = N(B)^\perp$$

$$\therefore B = B^T$$

From(d),  $S = C(P) = N(P)^\perp$ , and  $P^T = P$

$$\text{Hence, let } B = P = \begin{bmatrix} 2/3 & 0 & 1/3 & 1/3 \\ 0 & 0 & 0 & 0 \\ 1/3 & 0 & 2/3 & -1/3 \\ 1/3 & 0 & -1/3 & 2/3 \end{bmatrix}$$

(g)

$$(x - Mx)^T My = x^T My - x^T M^T My = x^T My - x^T My = 0$$

Hence,  $(x - Mx) \perp My$

**4.(20pts)**

(a)

$$x_n \in N(A)$$

$$\|x\|^2 = \|x_r + x_n\|^2 = \|x_r\|^2 + \|x_n\|^2 + 2x_r^T x_n$$

$$\text{If } x_r \in C(A^T), x_r^T x_n = 0$$

$$\|x\|^2 = \|x_r\|^2 + \|x_n\|^2 \geq \|x_r\|^2$$

$$Ax = A(x_r + x_n) = Ax_r$$

When  $x = x_r$ ,  $\|x\|$  is minimized.

(b)

$$A^T A \hat{x} = A^T b$$

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

$$\hat{x} = \begin{bmatrix} -2 \\ 3 \end{bmatrix}$$

(c)

When  $x_r \in C(A^T)$ , we have minimized  $\|x\|^2$ .

$$Ax = A(x_r + x_n) = Ax_r = b$$

$$\text{Let } A^T y = x_r$$

$$AA^T y = b$$

$$x = x_r = A^T (AA^T)^{-1} b = \begin{bmatrix} -7/6 \\ 1/3 \\ 11/6 \end{bmatrix}$$

(d)

$$A^T A \hat{x} = A^T b$$

$$\begin{bmatrix} 5 & 10 \\ 10 & 20 \end{bmatrix} \hat{x} = \begin{bmatrix} 7 \\ 14 \end{bmatrix}$$

We find  $\hat{x}$  has infinite solutions.

$$C\hat{x} = [5 \ 10]\hat{x} = 7 = d$$

From (c),

$$\hat{x} = C^T (CC^T)^{-1} d = \begin{bmatrix} 7/25 \\ 14/25 \end{bmatrix}$$