

## Problem 1: Convolution as Linea Map

### Solution

- (a) Obviously the size of matrice  $T$  must be  $(N + 1) \times (N + 1)$ . Let  $T_{i1}, T_{i2}, \dots, T_{i(N+1)}$  be the  $i$ -th row of matrice  $T$ . Since  $y = Tx$ , we can get

$$v(i - 1) = \sum_{j=1}^{N+1} T_{ij} u(j - 1) = T_{i1} u(0) + T_{i2} u(1) + \dots + T_{i(N+1)} u(N) \quad (1)$$

Accoring to the convolution function,

$$v(i - 1) = \sum_{k=-\infty}^{+\infty} h(k) u(i - 1 - k) = \sum_{k=(i-1-N)}^{i-1} h(k) u(i - 1 - k) \quad (2)$$

Comparing function (1) and function (2), it is easy to conclude that

$$T_{ij} = h(i - j), \text{ where } 1 \leq i \leq (N + 1), 1 \leq j \leq (N + 1)$$

So the matrice is a matrice where each element could be represented as  $T_{ij} = h(i - j)$ , where  $1 \leq i \leq (N + 1)$ ,  $1 \leq j \leq (N + 1)$ .

- (b) The structure of matrice  $T$  could be described as follow:

$$T_{i,j} = T_{i+1,j+1}$$

## Problem 2: Affine Funciton

### Solution

- (a) *Proof.* For any  $\alpha, \beta \in \mathbb{R}$  and any  $x, y \in \mathbb{R}^n$

$$\alpha f(x) + \beta f(y) = \alpha(Ax + b) + \beta(Ay + b) = \alpha Ax + \beta Ay + (\alpha + \beta)b = \alpha Ax + \beta Ay + b$$

$$f(\alpha x + \beta y) = A(\alpha x + \beta y) + b = \alpha Ax + \beta Ay + (\alpha + \beta)b = \alpha Ax + \beta Ay + b$$

Hence,  $\alpha f(x) + \beta f(y) = f(\alpha x + \beta y)$ . So function  $f(x) = Ax + b$  is affine.  $\square$

- (b) *Proof.* First we can show that  $b$  is unique. Because  $f(0) = b$ , so  $b$  must be unique, otherwise  $f(0)$  will be mapped as different values in  $\mathbb{R}^m$ , which conflicts with the function definition. Then we can show that  $A$  is unique. Let function

$$g(x) = f(x) - f(0) = Ax + b - b = Ax$$

Suppose  $b_1, b_2, \dots, b_n$  are the basis of  $\mathbb{R}^n$  and  $B = \sum_{i=1}^n \alpha_i b_i$ ,  $\alpha_i \in \mathbb{R}$ .

$$g(B) = g\left(\sum_{i=1}^n \alpha_i b_i\right) = A\left(\sum_{i=1}^n \alpha_i b_i\right) = \sum_{i=1}^n \alpha_i Ab_i = \sum_{i=1}^n \alpha_i g(b_i)$$

Suppose  $b_i = e_i$ ,  $i = 1, 2, \dots, n$ , then  $g(B)$  above could be represented as

$$\begin{aligned} g(B) &= [g(b_1) \ g(b_2) \ \dots \ g(b_n)] \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{bmatrix} \\ &= [g(e_1) \ g(e_2) \ \dots \ g(e_n)] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \\ &= Ax \end{aligned}$$

where

$$x \in \mathbb{R}^n.$$

Hence  $A$  is unique. So any affine function  $f$  could be represented uniquely as  $f(x) = Ax + b$  for some  $A \in \mathbb{R}^{m \times n}$  and  $b \in \mathbb{R}^{m \times 1}$   $\square$

## Problem 3: Matrix Multification

### Solution

(a) Suppose

$$A = \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} \neq 0$$

$$B = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \neq 0$$

Then

$$AB = 0$$

Hence the statement is incorrect.

(b) Suppose

$$A = \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} \neq 0$$

Then

$$A^2 = 0$$

Hence the statement is incorrect.

(c) Suppose

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

Then

$$A^T A = \begin{bmatrix} a_{11}^2 & a_{12}^2 & \dots & a_{1n}^2 \\ a_{21}^2 & a_{22}^2 & \dots & a_{2n}^2 \\ \vdots & \vdots & & \vdots \\ a_{n1}^2 & a_{n2}^2 & \dots & a_{nn}^2 \end{bmatrix}$$

So if  $A^T A = 0$ , then  $a_{ij} = 0$ ,  $1 \leq i \leq n$ ,  $1 \leq j \leq n$ .  $A = 0$ . So the statement is correct.

## Problem 4: Linear Maps and Differentiation of polynomials

### Solution

(a) For any  $p_1(x)$ ,  $p_2(x) \in \mathcal{P}_n$ ,

$$T(p_1(x) + p_2(x)) = \frac{d(p_1(x) + p_2(x))}{dx} = \frac{dp_1(x)}{dx} + \frac{dp_2(x)}{dx} = T(p_1(x)) + T(p_2(x)) \quad (3)$$

$$T(\alpha p_1(x)) = \frac{d(\alpha p_1(x))}{dx} = \alpha \frac{dp_1(x)}{dx} = \alpha T(p_1(x)) \quad (4)$$

Since equations (3) and (4) are valid,  $T$  is linear.

(b) For