

Problem 1

$$u = \begin{bmatrix} 6 \\ 2 \\ 9 \\ 1 \end{bmatrix} \quad v = \begin{bmatrix} 0 \\ 3 \\ -5 \\ 2 \end{bmatrix}$$

(a) Inner Product:  $u^T v$ 

$$\begin{aligned} u^T v &= [6 \ 2 \ 9 \ 1] \cdot \begin{bmatrix} 0 \\ 3 \\ -5 \\ 2 \end{bmatrix} = 6 \times 0 + 2 \times 3 + 9 \times (-5) + 1 \times 2 \\ &= 0 + 6 + (-45) + 2 \\ &= -37 \end{aligned}$$

(b) Outer Product:  $u \otimes v = u v^T$ 

$$\begin{aligned} u \cdot v^T &= \begin{bmatrix} 6 \\ 2 \\ 9 \\ 1 \end{bmatrix}_{4 \times 1} \cdot \begin{bmatrix} 0 & 3 & -5 & 2 \end{bmatrix}_{1 \times 4} = \begin{bmatrix} 6 \times 0 & 6 \times 3 & 6 \times (-5) & 6 \times 2 \\ 2 \times 0 & 2 \times 3 & 2 \times (-5) & 2 \times 2 \\ 9 \times 0 & 9 \times 3 & 9 \times (-5) & 9 \times 2 \\ 1 \times 0 & 1 \times 3 & 1 \times (-5) & 1 \times 2 \end{bmatrix}_{4 \times 4} \\ &= \begin{bmatrix} 0 & 18 & -30 & 12 \\ 0 & 6 & -10 & 4 \\ 0 & 27 & -45 & 18 \\ 0 & 3 & -5 & 2 \end{bmatrix} \end{aligned}$$

(c) Hadamard Product:  $u \odot v$ 

$$u \odot v = \begin{bmatrix} 6 \\ 2 \\ 9 \\ 1 \end{bmatrix} \odot \begin{bmatrix} 0 \\ 3 \\ -5 \\ 2 \end{bmatrix} = \begin{bmatrix} 6 \times 0 \\ 2 \times 3 \\ 9 \times (-5) \\ 1 \times 2 \end{bmatrix} = \begin{bmatrix} 0 \\ 6 \\ -45 \\ 2 \end{bmatrix}$$

(d) Inner Product:  $a^H \cdot b$ , where  $^H$  denotes conjugate transpose

(2)

$$a = \begin{bmatrix} 1 \\ e^{i\pi/3} \\ e^{-i\pi/2} \\ e^{-i\pi/3} \end{bmatrix}$$

$$b = \begin{bmatrix} 0 \\ 1e^{i\pi} \\ 2e^{-i\pi/2} \\ 4e^{i\pi/2} \end{bmatrix}$$

$$a^H = \begin{bmatrix} 1 & e^{-i\pi/3} & e^{i\pi/2} & e^{i\pi/3} \end{bmatrix}$$

$$b = \begin{bmatrix} 0 \\ -1e^{-i\pi/2} \\ 2e^{i\pi/2} \\ 4e^{i\pi/2} \end{bmatrix}$$

$$\begin{aligned} a^H \cdot b &= \begin{bmatrix} 1 & e^{-i\pi/3} & e^{i\pi/2} & e^{i\pi/3} \end{bmatrix} \cdot \begin{bmatrix} 0 \\ -1e^{-i\pi/2} \\ 2e^{i\pi/2} \\ 4e^{i\pi/2} \end{bmatrix} = (1 \times 0) + (-1) \times e^{-i\pi/3} + (e^{i\pi/2} \times 2e^{-i\pi/2}) \\ &\quad + (e^{i\pi/3} \times 4e^{i\pi/2}) \\ &= 0 - e^{-i\pi/3} + 2e^{i(\frac{\pi}{2} - \frac{\pi}{2})} + 4e^{i(\frac{\pi}{3} + \frac{\pi}{2})} \\ &= -e^{-i\pi/3} + 2 + 4e^{i\frac{5\pi}{6}} \\ &= 2 - e^{-i\frac{\pi}{3}} + 4e^{i\frac{5\pi}{6}} \end{aligned}$$

(e) Inner Products:  $a^H \cdot a$  &  $b^H \cdot b$

$$\begin{aligned} a^H \cdot a &= \begin{bmatrix} 1 & e^{-i\pi/3} & e^{i\pi/2} & e^{i\pi/3} \end{bmatrix} \cdot \begin{bmatrix} 1 \\ e^{i\pi/3} \\ e^{-i\pi/2} \\ e^{-i\pi/3} \end{bmatrix} = (1 \times 1) + (e^{-i\pi/3} \times e^{i\pi/3}) \\ &\quad + (e^{i\pi/2} \times e^{-i\pi/2}) + (e^{i\pi/3} \times e^{-i\pi/3}) \\ &= 1 + e^{i(\frac{\pi}{3} - \frac{\pi}{3})} + e^{i(\frac{\pi}{2} - \frac{\pi}{2})} + e^{i(\frac{\pi}{3} - \frac{\pi}{3})} \\ &= 1 + 1 + 1 + 1 = 4 \end{aligned}$$

$$\begin{aligned} b^H \cdot b &= \begin{bmatrix} 0 & e^{-i\pi} & 2e^{i\pi/2} & 4e^{-i\pi/2} \end{bmatrix} \cdot \begin{bmatrix} 0 \\ e^{i\pi} \\ 2e^{i\pi/2} \\ 4e^{i\pi/2} \end{bmatrix} = (0 \times 0) + (e^{-i\pi} \times e^{i\pi}) \\ &\quad + (2e^{i\pi/2} \times 2e^{-i\pi/2}) + (4e^{-i\pi/2} \times 4e^{i\pi/2}) \\ &= 0 + e^{i(\pi - \pi)} + 4e^{i(\frac{\pi}{2} - \frac{\pi}{2})} + 16e^{i(\frac{\pi}{2} - \frac{\pi}{2})} \\ &= 0 + e^0 + 4e^0 + 16e^0 \\ &= 0 + 1 + 4 + 16 = 21 \end{aligned}$$

(3)

 $a^*$  - complex conjugate(f) Hadamard Product  $a \odot a^*$ 

$$a \odot a^* = \begin{bmatrix} 1 \\ e^{i\pi/3} \\ e^{-i\pi/2} \\ e^{-i\pi/3} \end{bmatrix} \odot \begin{bmatrix} 1 \\ e^{-i\pi/3} \\ e^{i\pi/2} \\ e^{i\pi/3} \end{bmatrix} = \begin{bmatrix} 1 \times 1 \\ e^{i\pi/3} \times e^{-i\pi/3} \\ e^{-i\pi/2} \times e^{i\pi/2} \\ e^{-i\pi/3} \times e^{i\pi/3} \end{bmatrix} = \begin{bmatrix} 1 \\ e^{i(\frac{\pi}{3} - \frac{\pi}{3})} \\ e^{i(\frac{\pi}{2} - \frac{\pi}{2})} \\ e^{i(\frac{\pi}{3} - \frac{\pi}{3})} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$I = \begin{bmatrix} 1 & 8 & 3 \\ 7 & 4 & 1 \\ 5 & 9 & 1 \end{bmatrix}$$

Write expression to calculate avg of 1st column using matrix operations

Let's say  $x = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$  - To calculate average of 1st column of  $I$  we can multiply  $I$  with  $x$  (Matrix Vector Product)

- This will give us the 1st column of Matrix  $I$

- We can then take the sum of all elements in this column vector & divide it by its shape.

$$I \cdot x = \begin{bmatrix} 1 & 8 & 3 \\ 7 & 4 & 1 \\ 5 & 9 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 7 \\ 5 \end{bmatrix}$$

$$\text{Average} = \frac{\text{SUM}(I \cdot x)}{\text{SHAPE}(I \cdot x)} = \frac{13}{3} = 4.33$$

Now repeat this after vectorizing  $I$

(4)

- Flattened  $I$ :

$$i = [1 \ 8 \ 3 \ 7 \ 4 \ 1 \ 5 \ 9]$$

- Now to get the same result using Flattened  $I$ , we can use the following column vector  $x$

$$x = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} \quad i \cdot x = [1 \ 8 \ 3 \ 7 \ 4 \ 1 \ 5 \ 9] \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} = 1 + 7 + 5 = 13$$

- Next we can divide this result by sum of column vector  $x$

$$\text{Average} = \frac{13}{\text{sum}(x)} = \frac{13}{3} = 4.33$$

Problem 2  $u = [1, -1]$   $v = [1, 2, 3, 4, 5, 6]$

(a) Discrete Convolution ( $W = u * v$ ) Flipped  $u = [-1, 1]$

$k$	-1	0	1	2	3	4	5	6
$v[k]$		1	2	3	4	5	6	
$u[-k]$	-1	1						
$u[1-k]$		-1	1					
$u[2-k]$			-1	1				
$u[3-k]$				-1	1			
$u[4-k]$					-1	1		
$u[5-k]$						-1	1	
$u[6-k]$							-1	1
	1	1	1	1	1	1	-6	

$$W = v * u = [1 \ 1 \ 1 \ 1 \ 1 \ 1 \ -6 \ ]$$



(6)  $W = U \cdot v$  - Matrix  $U \in \mathbb{R}^{7 \times 6}$  holds blur kernel

(5)

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

- To write a blur kernel  $u$  as a matrix, we can write it as a  $7 \times 6$

- Banded matrix
- (A matrix similar to diagonal matrix where most energy is concentrated towards diagonal)
- Every row of  $U$  corresponds to a shifted  $u$ .
- This blur kernel is helping us take a weighted avg of neighbouring pixels

### (C) Deconvolution

We can determine the Deconvolution matrix  $D$  by calculating the pseudoinverse of our convolution matrix  $U$

$$\hookrightarrow (U^T \cdot U)^{-1} \cdot U^T = D$$

$$U = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 1 & 0 & 0 & 0 \\ 0 & 0 & -1 & 1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 1 & 0 \\ 0 & 0 & 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & 0 & 0 & -1 \end{bmatrix}_{7 \times 6}$$

$$U^T = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}_{6 \times 7}$$

$$(U^T \cdot U) = \begin{bmatrix} 2 & -1 & 0 & 0 & 0 & 0 \\ -1 & 2 & -1 & 0 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 & 0 \\ 0 & 0 & -1 & 2 & -1 & 0 \\ 0 & 0 & 0 & -1 & 2 & -1 \\ 0 & 0 & 0 & 0 & -1 & 2 \end{bmatrix}_{6 \times 6}$$

$$(U^T \cdot U)^{-1} = \frac{1}{7} \begin{bmatrix} 6 & 5 & 4 & 3 & 2 & 1 \\ 5 & 10 & 8 & 6 & 4 & 2 \\ 4 & 8 & 12 & 9 & 6 & 3 \\ 3 & 6 & 9 & 12 & 8 & 4 \\ 2 & 4 & 6 & 8 & 10 & 5 \\ 1 & 2 & 3 & 4 & 5 & 6 \end{bmatrix}_{6 \times 6}$$

$$D = (U^T \cdot U)^{-1} \cdot U^T = \frac{1}{7} \begin{bmatrix} 6 & -1 & -1 & -1 & -1 & -1 \\ 5 & 5 & -2 & -2 & -2 & -2 \\ 4 & 4 & 4 & -3 & -3 & -3 \\ 3 & 3 & 3 & 3 & -4 & -4 \\ 2 & 2 & 2 & 2 & 2 & -5 \\ 1 & 1 & 1 & 1 & 1 & -6 \end{bmatrix}_{6 \times 7}$$

The matrix  $D$  is inverting the effects of convolution caused with matrix  $U$

(d) Bonus: Is  $D, D^T$  invertible?

(6)

Yes, this is an invertible matrix.  $D$  is  $6 \times 7$  &  $D^T$  is  $7 \times 6$  so  $D \cdot D^T$  is a  $6 \times 6$  matrix. Its determinant is non zero & its rank is 6, hence it is invertible.

This result is like a pseudo inverse of  $D$  & can help undo the convolution effect of the matrix.

Problem 3 Proof for FT theorem  $F[u(x) * v(x)] = F[u(x)] \cdot F[v(x)]$

Definition of Fourier Transform:  $F[f(x)] = \int_{-\infty}^{\infty} f(x) e^{-2\pi i f x} dx$  — (1)

Definition of convolution:  $u(x) * v(x) = \int_{-\infty}^{\infty} u(\tau) v(x-\tau) d\tau$  — (2)

Let's consider LHS of our eq<sup>n</sup> & substitute (2) in (1)

$$F[u(x) * v(x)] = \int_{-\infty}^{\infty} [u(x) * v(x)] e^{-2\pi i f x} dx$$
$$= \int_{-\infty}^{\infty} \left( \int_{-\infty}^{\infty} u(\tau) v(x-\tau) d\tau \right) e^{-2\pi i f x} dx \rightarrow \text{changing order of integration}$$

$$= \int_{-\infty}^{\infty} u(\tau) \int_{-\infty}^{\infty} v(x-\tau) e^{-2\pi i f x} dx d\tau$$

Let  $(x-\tau) = x'$   
so  $x = x' + \tau$  &  $dx = dx'$

$$\therefore = \int_{-\infty}^{\infty} u(\tau) \int_{-\infty}^{\infty} v(x') e^{-2\pi i f (x' + \tau)} dx' d\tau$$

$$= \int_{-\infty}^{\infty} u(\tau) \int_{-\infty}^{\infty} (v(x') e^{-2\pi i f x'}) (e^{-2\pi i f \tau}) d\tau$$

$$= \int_{-\infty}^{\infty} u(\tau) e^{-2\pi i f \tau} \left[ \int_{-\infty}^{\infty} v(x') e^{-2\pi i f x'} dx' \right] d\tau$$

$$= \int_{-\infty}^{\infty} u(\tau) e^{-2\pi i f \tau} d\tau \int_{-\infty}^{\infty} v(x') e^{-2\pi i f x'} dx'$$

$$= F[u(x)] \cdot F[v(x)]$$

= R.H.S Hence Proved.

Using property of definite integrals

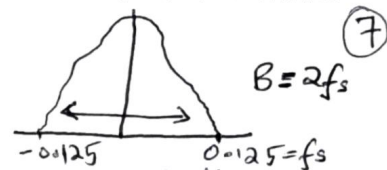
## Problem 4

(a) Optical field  $U(x, y)$

Max spatial frequency =  $0.125 \mu\text{m}^{-1}$

Image sensor is a cone function with pixel pitch =  $10 \mu\text{m}$

Can we recreate the original field from detected field.



According to Sampling theorem if continuous signal has a frequency of no more than  $B \text{ Hz}$  then we can recreate this signal if the samples are drawn at at least  $2B$  samples per second.

- So Max Frequency of continuous signal =  $0.125 \mu\text{m}^{-1}$
- So sampling should be done at at least  $2 \times 0.125 = 0.25 \mu\text{m}^{-1}$
- Distance b/w centers of Adjacent pixels =  $10 \mu\text{m}$   
 $\therefore$  it Samples at the rate of  $\frac{1}{10} \mu\text{m}$   
 $\therefore$  sampling frequency =  $0.1 \mu\text{m}^{-1}$

Since sampling rate  $<$  twice the frequency of continuous signal.

Hence original signal cannot be reconstructed.

If pixel pitch =  $6 \mu\text{m}$

then sampling frequency =  $\frac{1}{6} = 0.16$

$$0.16 < 0.25$$

$\therefore$  original signal cannot be reconstructed

If pixel pitch =  $4 \mu\text{m}$

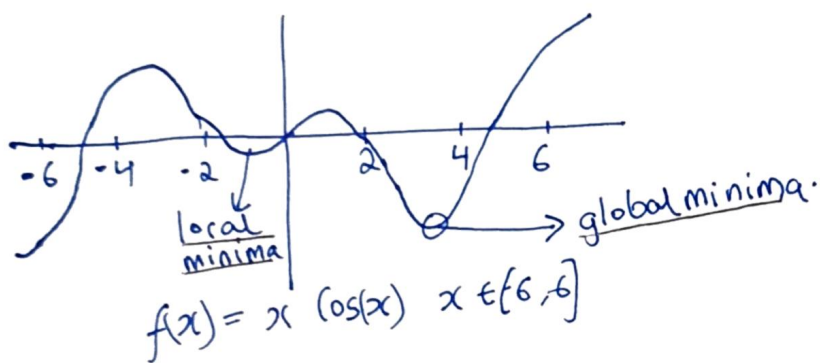
then sampling frequency =  $\frac{1}{4} = 0.25$

$$0.25 \leq 0.25$$

$\therefore$  original signal can be reconstructed



(6)  $f(x) = x \cos(x) \quad x \in [-6, 6]$  - Local & global minima?  
 - good value  
 - bad value



- we can find the minimas by taking a derivative of the function & finding critical points where derivative = 0
- 2nd derivative can be used to find if a critical point is a local minima or global minima.

- Gradient Descent does not always guarantee to find a global minima

It is possible to get stuck on a local minima in case the initial value is very far away from the global minima.

- A good value for initial state would be close to 3, somewhere b/w 2 & 5 as it would start close to the global minima

- A bad value for initial state could be close to 0, somewhere b/w -3 & 0 which is far away from global minima & it might get stuck in the local minima. The convergence would also be slow.

(C) Linear eq<sup>n</sup>  $f(x) = \text{sign}(W^T x)$   $w = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$   $x = (1, x_1, x_2)^T$   
 $\text{sign}(x) = \begin{cases} 1 & x > 0 \\ 0 & x = 0 \\ -1 & x < 0 \end{cases}$   $f(x)$  separates our data into 2 classes  $\begin{matrix} \rightarrow +1 \\ \rightarrow -1 \end{matrix}$

The line which separates our classes is the decision boundary  
 eq<sup>n</sup> of the line  $W^T x = 0$

$$\Rightarrow \begin{bmatrix} 1 & -2 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ x_1 \\ x_2 \end{bmatrix} = 0$$

$$\Rightarrow 1 - 2x_1 + x_2 = 0$$

$$\Rightarrow \boxed{2x_1 - x_2 - 1 = 0}$$

