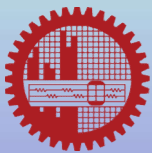


CSE6706: *Advanced Digital Image Processing*

Dr. Md. Monirul Islam



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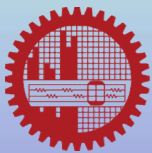
Types of Region Based Segmentation

- Region growing
- Region splitting and merging



Region Splitting and Merging

- Main Idea:
 - Subdivide an image into arbitrary regions
 - Split and merge the existing regions based some criteria

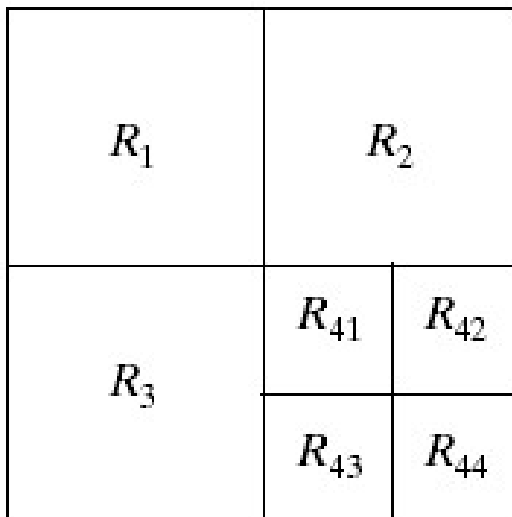


Region Splitting and Merging

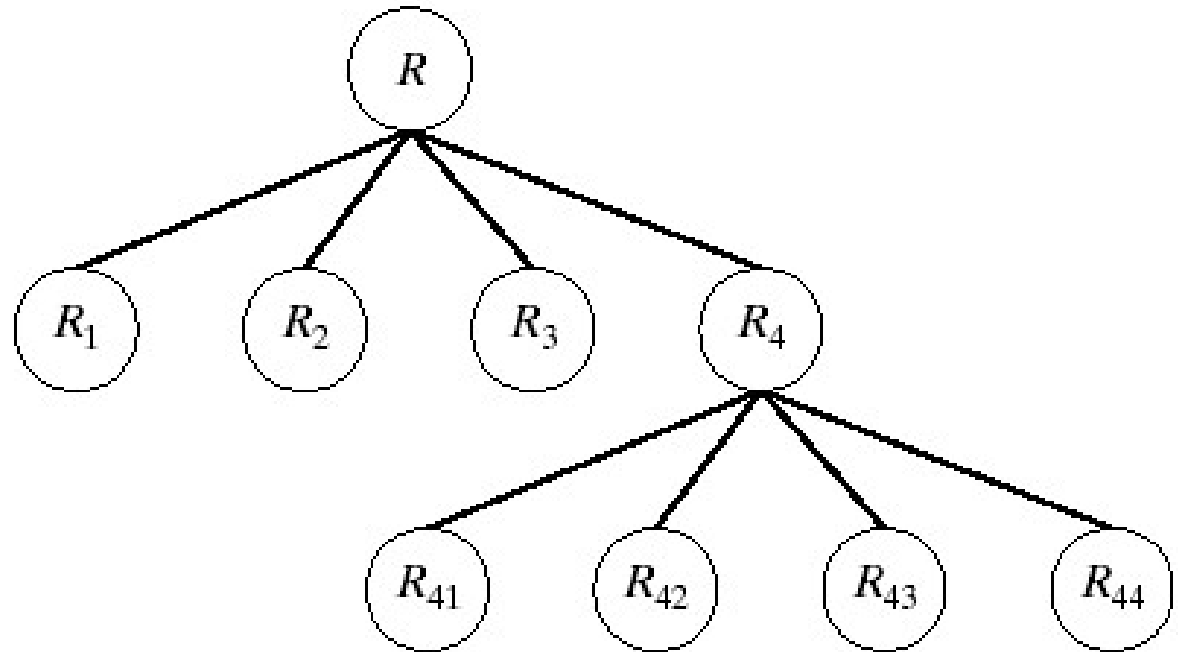
- Let,
 - R be the entire region/image
 - P be a predicate on R
- Splitting:
 - Start with R
 - If $P(R)=\text{FALSE}$
 - subdivide R successively into **smaller and smaller** quadrant regions R_i so that $P(R_i)=\text{TRUE}$
 - for any quadrant R_i , if $P(R_i)=\text{FALSE}$, divide it into sub-quadrants, and so on



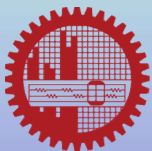
Region Splitting and Merging



An image, R



Quad Tree
representation



Region Splitting and Merging

- Merging:
 - Merge two regions R_j and R_k if $P(R_j \cup R_k) = \text{TRUE}$



Region Splitting and Merging (1)

1. **Split** into four disjoint quadrants any region R_i for which $P(R_i)=\text{FALSE}$
2. When no splitting possible, **merge** any **adjacent regions** R_j and R_k if $P(R_j \cup R_k)=\text{TRUE}$
3. **Stop** when no merging is possible

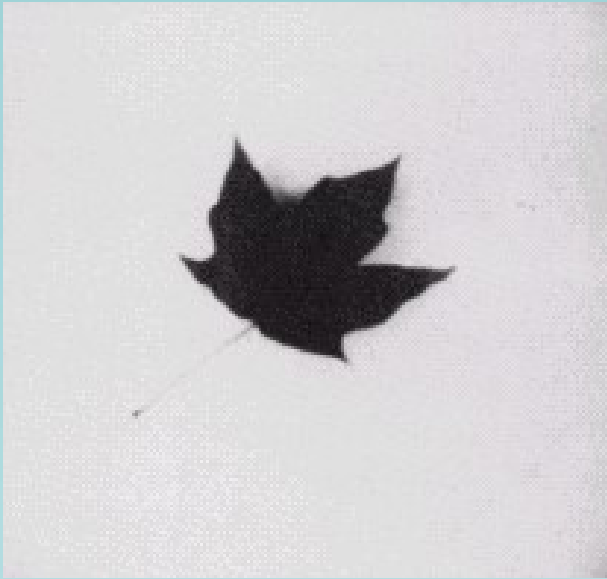


Region Splitting and Merging (2)

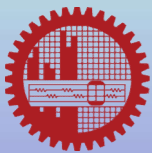
1. Split the image initially into set of blocks
2. Initial merging is limited to group of four blocks that are descendent and satisfy predicate
3. When no merging is possible as step 2, do final mergings as follows
 - 3.1 Merge any adjacent regions R_j and R_k if $P(R_j \cup R_k) = \text{TRUE}$



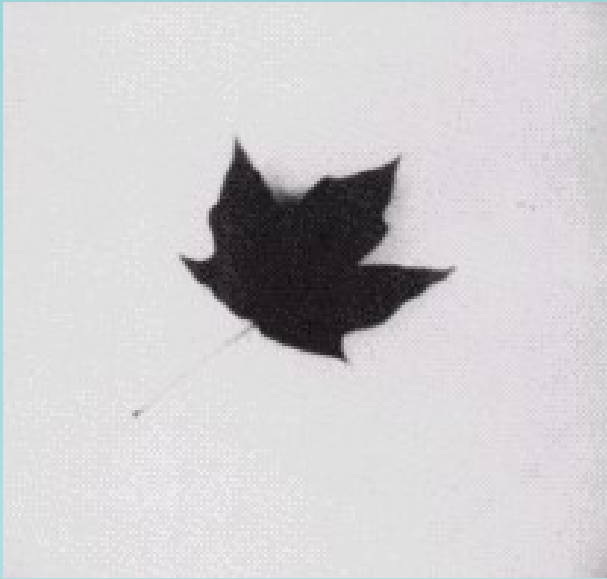
Region Splitting and Merging



- Predicate: $P(R_i)=\text{TRUE}$ if
 - At least 80% of pixels in R_i satisfy $|z-m_i|<2\sigma_i$, where
 - z : gray level of an pixel in R_i
 - m_i : average gray level in R_i
 - σ_i : standard deviation of gray level in R_i



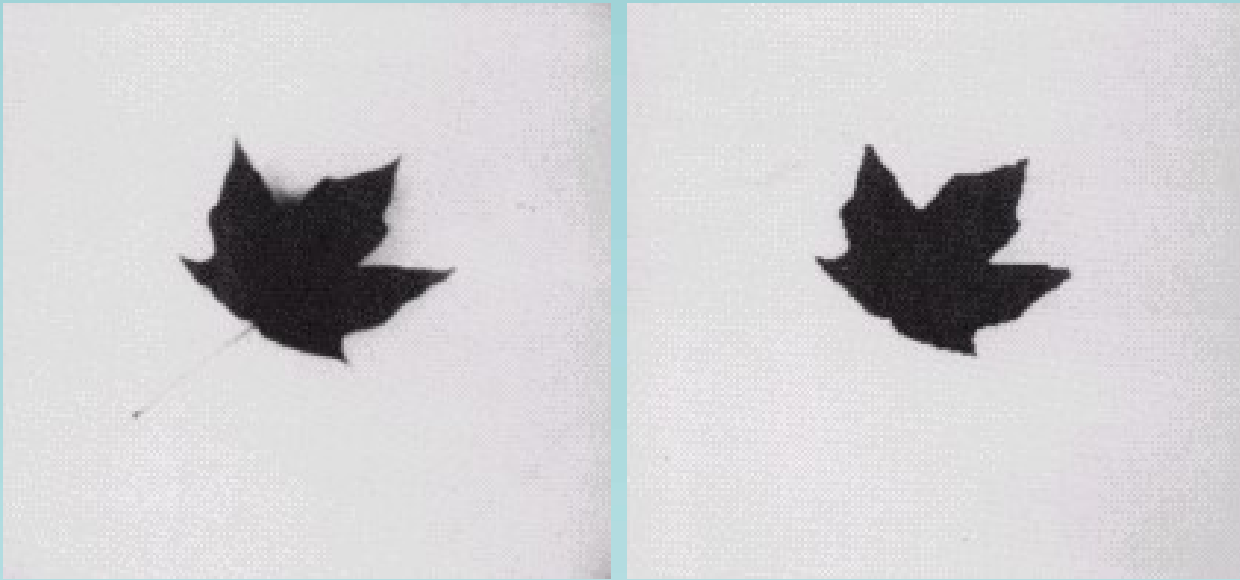
Region Splitting and Merging



- Predicate: $P(R_i)=\text{TRUE}$ if
 - At least 80% of pixels in R_i satisfy $|z-m_i|<2\sigma_i$, where
 - z : gray level of an pixel in R_i
 - m_i : average gray level in R_i
 - σ_i : standard deviation of gray level in R_i
- If satisfy, assign m_i to all pixel in R_i



Region Splitting and Merging



Using thresholding



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Region Splitting and Merging



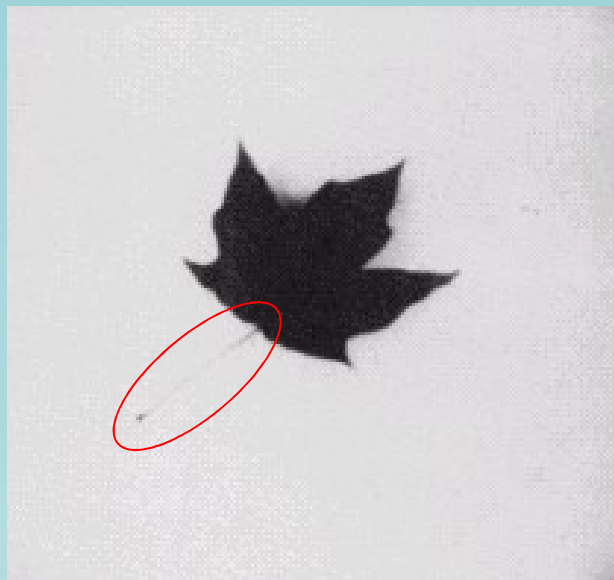
Using thresholding



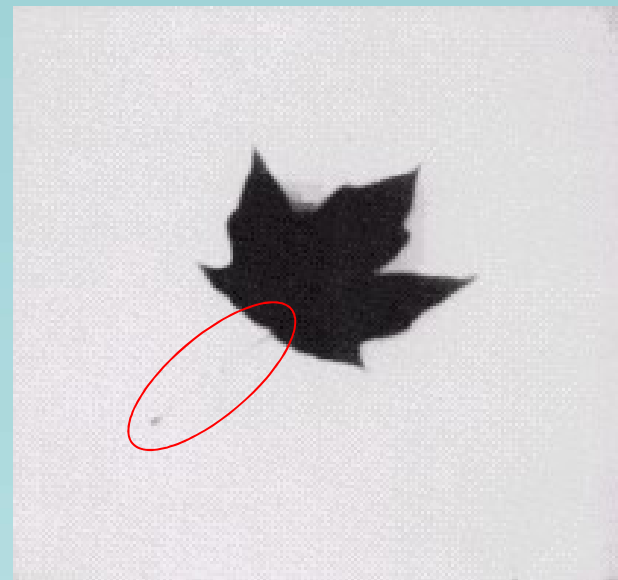
After region splitting
and merging



Region Splitting and Merging



Using thresholding



After region splitting
and merging



Motion in Segmentation

Motion

- relative displacement between imaging device and an object
- extracted from a sequence of image frames
- useful to find an object from a background of irrelevant detail



Motion in Segmentation

- Two way to extract motion:
 - Spatial domain technique
 - Frequency domain technique



Motion Extraction in Spatial Domain

- Motion is detected by monitoring the **changes in subsequent image frames**
- Compare **pixel by pixel gray differences** between two frames



Motion Extraction in Spatial Domain

- Let there be a **reference frame** consisting of **stationary objects**
- Other frames have objects with motion
- Find $diff_image = new_frame - ref_image$
 - Canceling everything except nonzero entries corresponding to moving object



Motion Extraction in Spatial Domain

- Let there be two frames $f(x, y, t_{t_i})$ and $f(x, y, t_{t_j})$
- The difference is given by,

$$d_{i,j}(x, y) = \begin{cases} 1 & \text{if } |f(x, y, t_{t_i}) - f(x, y, t_{t_j})| > T \\ 0 & \text{otherwise} \end{cases}$$



Motion Extraction in Spatial Domain

$$d_{i,j}(x,y) = \begin{cases} 1 & \text{if } |f(x,y,t_{t_i}) - f(x,y,t_{t_j})| > T \\ 0 & \text{otherwise} \end{cases}$$

- All pixels in $d_{i,j}(x,y)$ having value 1 correspond to moving object(s)



Motion Extraction in Spatial Domain

$$d_{i,j}(x,y) = \begin{cases} 1 & \text{if } |f(x,y,t_{t_i}) - f(x,y,t_{t_j})| > T \\ 0 & \text{otherwise} \end{cases}$$

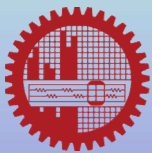
- All pixels in $d_{i,j}(x,y)$ having value 1 correspond to moving object(s)
- However, noise can also contribute to $d_{i,j}(x,y)$



Motion Extraction in Spatial Domain

$$d_{i,j}(x,y) = \begin{cases} 1 & \text{if } |f(x,y,t_{t_i}) - f(x,y,t_{t_j})| > T \\ 0 & \text{otherwise} \end{cases}$$

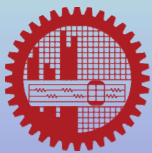
- All pixels in $d_{i,j}(x,y)$ having value 1 correspond to moving object(s)
- However, noise can also contribute to $d_{i,j}(x,y)$
- Remedy: use 4- or 8-connectivity check to find isolated noise



Motion Extraction in Spatial Domain

$$d_{i,j}(x, y) = \begin{cases} 1 & \text{if } |f(x, y, t_{t_i}) - f(x, y, t_{t_j})| > T \\ 0 & \text{otherwise} \end{cases}$$

- All pixels in $d_{i,j}(x, y)$ having value 1 correspond to moving object(s)
- However, noise can also contribute to $d_{i,j}(x, y)$
- Remedy: use 4- or 8-connectivity check to find isolated noise
- This can **cancel out small / slow moving object** as well



Motion Extraction in Spatial Domain

- Alternatively, monitor the change pattern of pixel values **over many frames**
- That means we need to use some **memory**



Motion Extraction in Spatial Domain

- Alternatively, monitor the change pattern of pixel values *over many frames*
- That means we need to use some *memory*
- The changes are accumulated in memory called *accumulative difference image (ADI)*



Motion Extraction in Spatial Domain

- The changes are accumulated in memory called *accumulative difference image* (ADI)
- Let there are frames $f(x, y, t_1), f(x, y, t_2), \dots, f(x, y, t_n)$
- For simplicity, let they are $f(x, y, 1), f(x, y, 2), \dots, f(x, y, n)$
- Let, $R(x, y) = f(x, y, 1)$



Motion Extraction in Spatial Domain

- The *accumulative difference image* (ADI) is calculated as

$$A_k(x, y) = \begin{cases} A_{k-1}(x, y) + 1 & \text{if } |R(x, y) - f(x, y, k)| > T \\ A_{k-1}(x, y) & \text{otherwise} \end{cases}$$



Motion Extraction in Spatial Domain

$$A_k(x, y) = \begin{cases} A_{k-1}(x, y) + 1 & \text{if } |R(x, y) - f(x, y, k)| > T \\ A_{k-1}(x, y) & \text{otherwise} \end{cases}$$

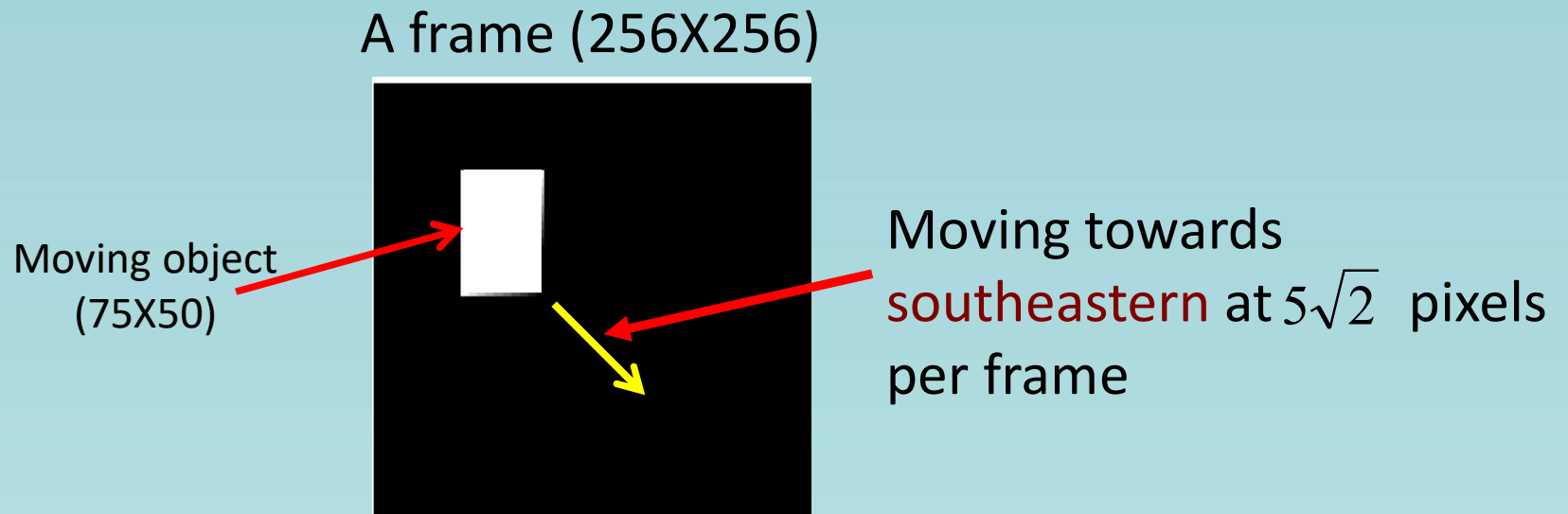
- Other two accumulators:

$$P_k(x, y) = \begin{cases} P_{k-1}(x, y) + 1 & \text{if } R(x, y) - f(x, y, k) > T \\ P_{k-1}(x, y) & \text{otherwise} \end{cases}$$

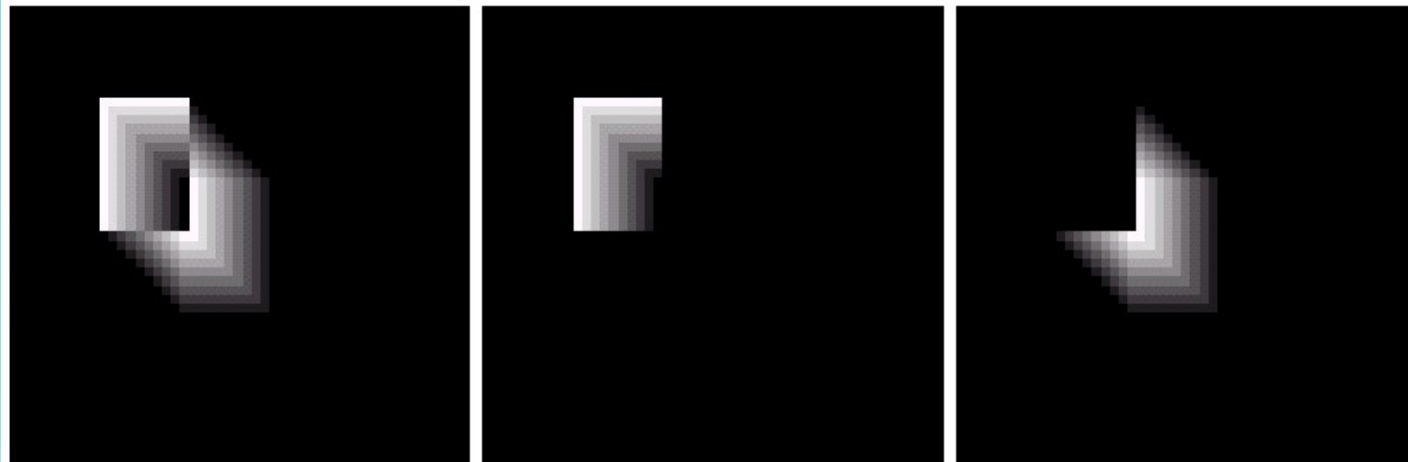
$$N_k(x, y) = \begin{cases} N_{k-1}(x, y) + 1 & \text{if } R(x, y) - f(x, y, k) < -T \\ N_{k-1}(x, y) & \text{otherwise} \end{cases}$$



Motion Extraction in Spatial Domain



Motion Extraction in Spatial Domain



Absolute

Positive

Negative

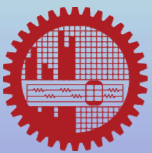
- Image 256X256
- Moving object: 75X50
- Moving towards **southeastern** at $5\sqrt{2}$ pixels per frame



Finding a reference image



- When the moving object in a frame completely moved out of its original position in a reference image,
 - Copy the background to the reference



Motion Extraction and Filtering in Spectral/Frequency Domain



Motion Extraction in Spectral/Frequency Domain

- Background:
 - Fourier transform



Fourier Series

‘Any function that *periodically repeats itself* can be expressed as the *sum of sines and/or cosines* of different frequencies’

- Joseph Fourier

$$f(t) = \sum_{n=-\infty}^{n=\infty} c_n e^{j \frac{2\pi n}{T} t} \quad \text{with } T = \text{period of } f(t)$$



Fourier Series

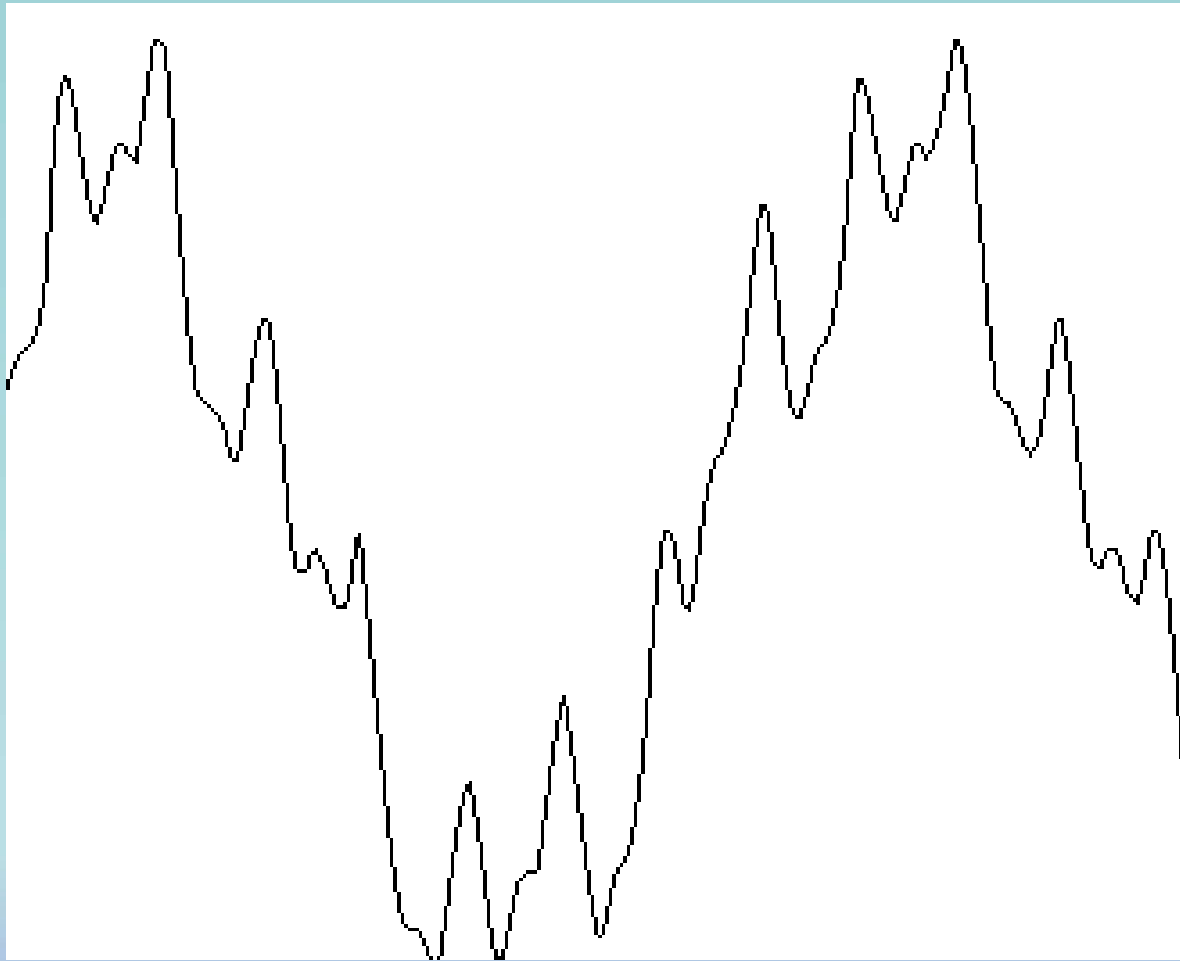
Fourier says:

Any function that *periodically repeats itself* can be expressed as the *sum of sines and/or cosines* of different frequencies

- Doesn't matter how complicated the function is!
- The summation is called *Fourier Series*

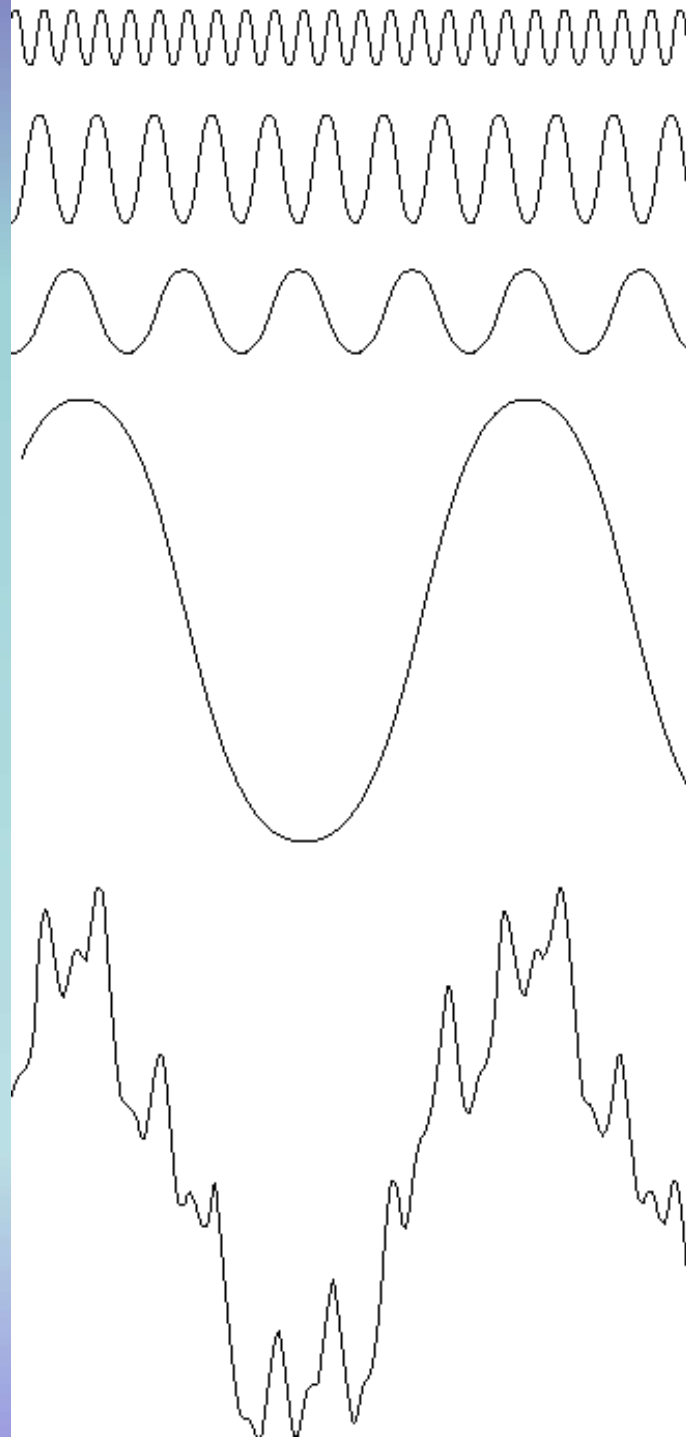


Fourier Series

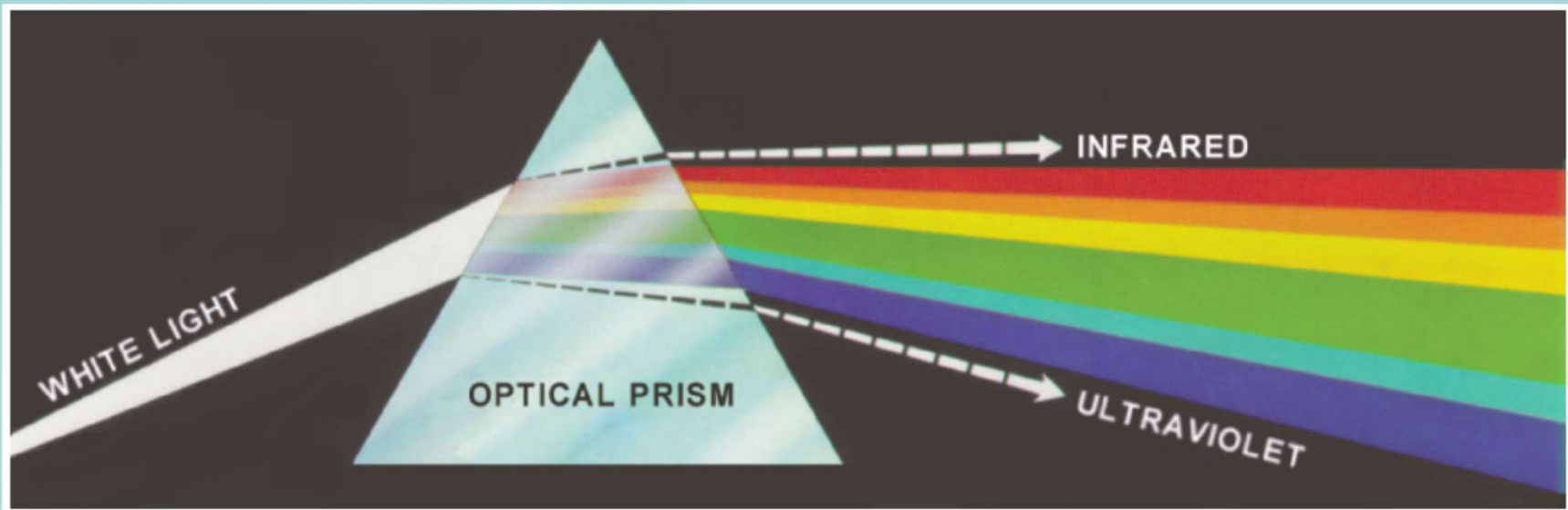


A complicated function



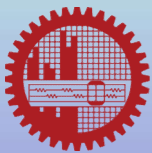


Fourier Series and Light Spectrum



A complicated
function

Individual
functions

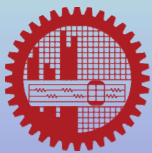


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Fourier for Aperiodic Function

- Even *aperiodic* function but *with finite area under curve* can be represented as *integral* of sines and cosines

$$f(t) = \int_{-\infty}^{\infty} F(u) e^{j2\pi ut} du$$



Fourier for Aperiodic Function

- Even *aperiodic* function but *with finite area under curve* can be represented as *integral* of sines and cosines
- This integral is called *Inverse Fourier Transform*



Fourier Transform

$$F(u) = \int_{-\infty}^{\infty} f(t) e^{-j2\pi ut} dt$$



Fourier Transform/Series

- Important property:
 - A function *represented in Fourier transform or Series* can be *reconstructed (recovered)* by an *inverse process*



1D Fourier Series and its Inverse

$$f(t) = \sum_{n=-\infty}^{n=\infty} c_n e^{j\frac{2\pi n}{T}t}$$

- The inverse Fourier series is

$$c_n = \frac{1}{T} \int_{-T/2}^{T/2} f(t) e^{-j\frac{2\pi n}{T}t} dt$$



1D Fourier Transform and its Inverse

$$F(u) = \int_{-\infty}^{\infty} f(t) e^{-j2\pi ut} dt$$

- The inverse Fourier transform is

$$f(t) = \int_{-\infty}^{\infty} F(u) e^{j2\pi ut} du$$



FT of Some Common Functions: Impulse Function

0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0



Continuous Impulse Function

$$\delta(t) = \begin{cases} \infty & \text{if } t = 0 \\ 0 & \text{if } t \neq 0 \end{cases}$$

with the constraint,

$$\int_{-\infty}^{\infty} \delta(t) dt = 1$$



Sifting Property of Continuous Impulse Function

$$\int_{-\infty}^{\infty} \delta(t) f(t) dt = f(0)$$



Sifting Property of Continuous Impulse Function

$$\int_{-\infty}^{\infty} \delta(t) f(t) dt = f(0)$$

- Evaluates the function at the location of the impulse



Sifting Property of Continuous Impulse Function

$$\int_{-\infty}^{\infty} \delta(t - t_0) f(t) dt = f(t_0)$$

- Evaluates the function at the location of the impulse

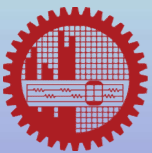
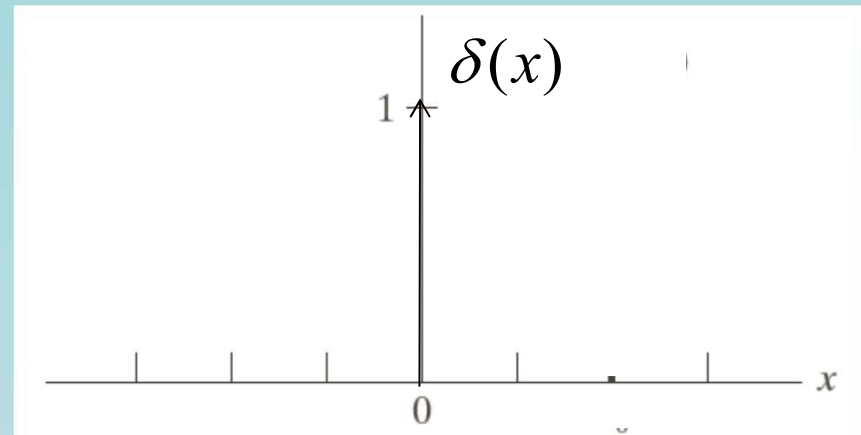


Discrete Unit Impulse Function

$$\delta(x) = \begin{cases} 1 & \text{if } x = 0 \\ 0 & \text{if } x \neq 0 \end{cases}$$

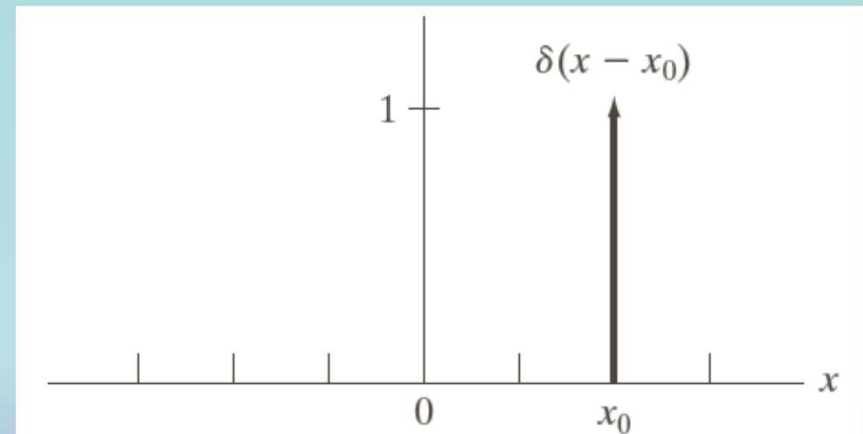
with the constraint,

$$\sum_{x=-\infty}^{x=\infty} \delta(x) = 1$$

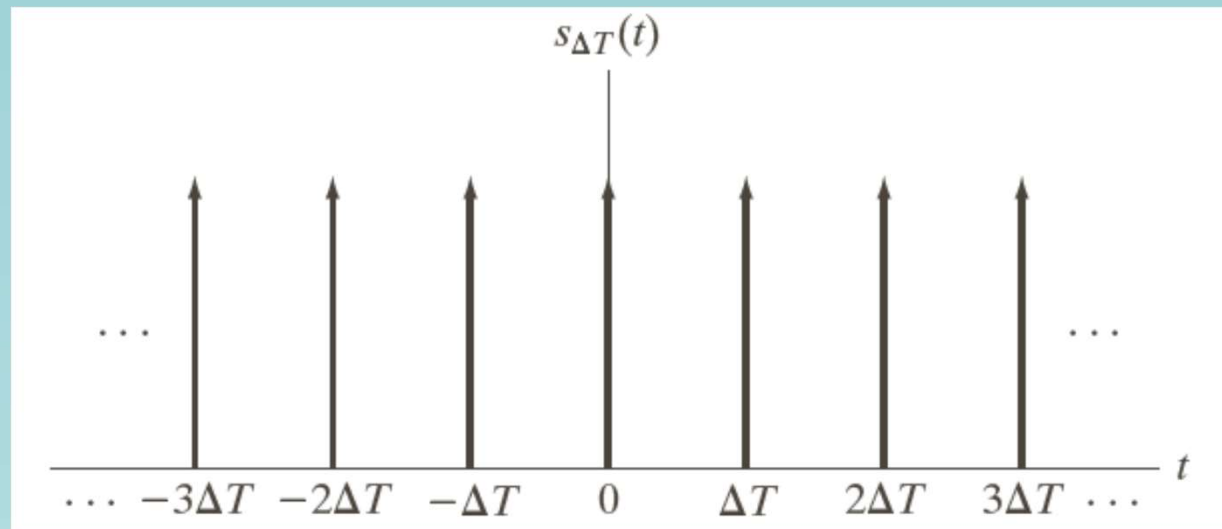


Sifting Property of Discrete Unit Impulse Function

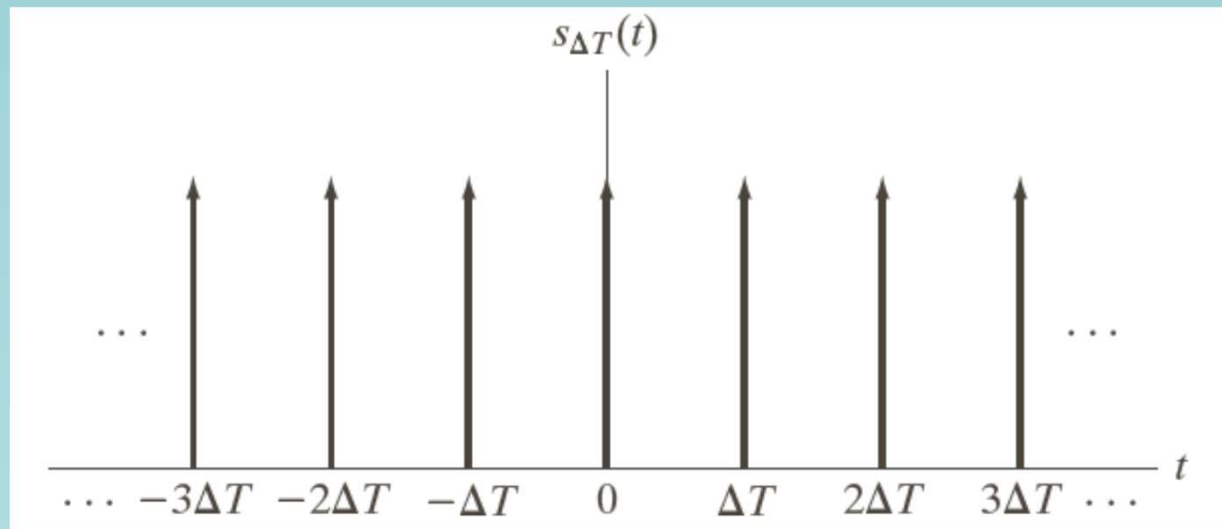
$$\sum_{x=-\infty}^{x=\infty} \delta(x - x_0) f(x) = f(x_0)$$



Impulse Train: $s_{\Delta T}(t)$



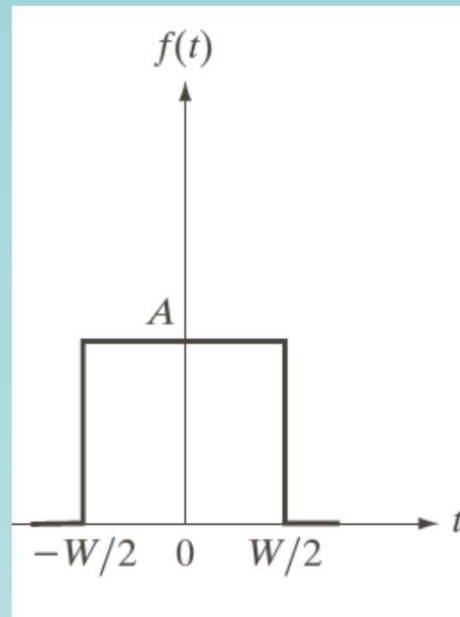
Impulse Train: $s_{\Delta T}(t)$



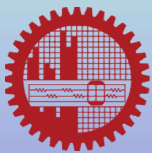
$$s_{\nabla T}(t) = \sum_{n=-\infty}^{n=\infty} \delta(t - n\nabla T)$$



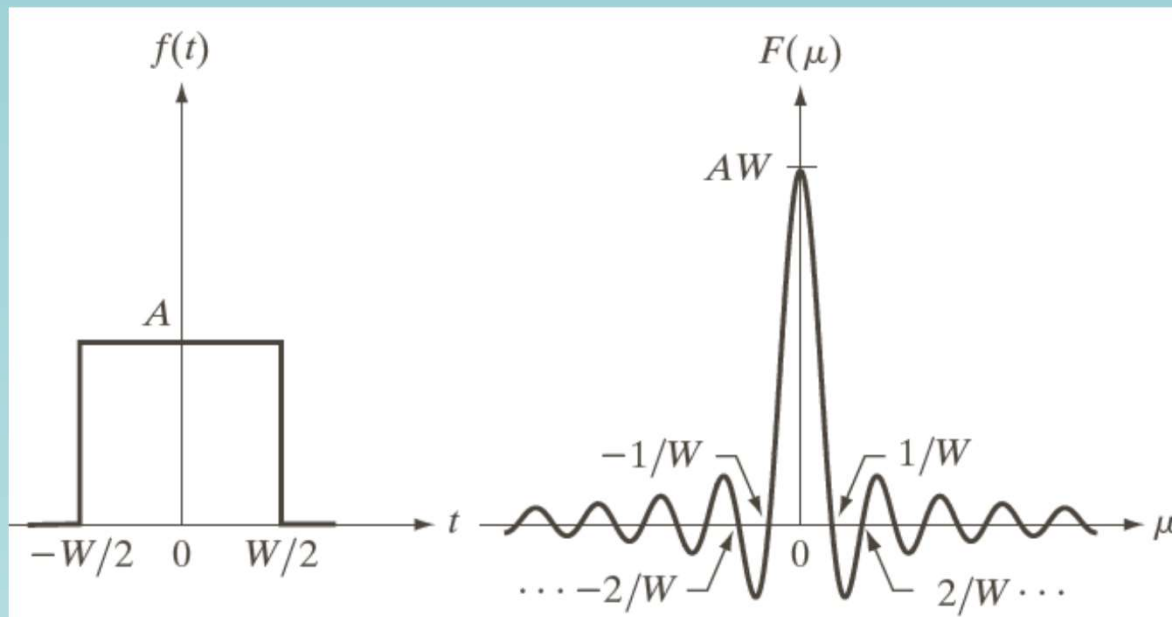
1D Fourier Transform and its Inverse



1D Box/Gate
Function

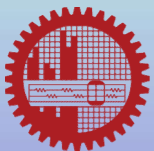


1D Fourier Transform and its Inverse

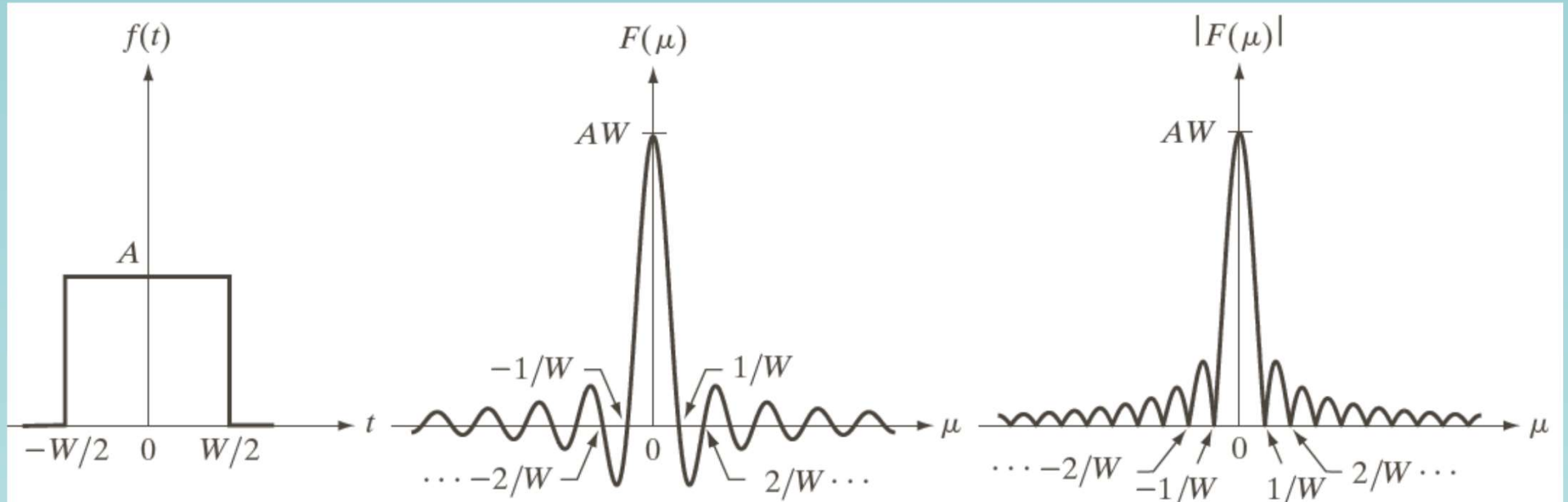


$$F(u) = \int_{-\infty}^{\infty} f(t) e^{-j2\pi ut} dt = \int_{-W/2}^{W/2} A e^{-j2\pi ut} dt = \dots$$

$$= \frac{A}{j2\pi u} \left[e^{j\pi u W} - e^{-j\pi u W} \right] = AW \frac{\sin(\pi u W)}{\pi u W} = AW \text{sinc}(\pi u W)$$



1D Fourier Transform and its Inverse



$$F(u) = \int_{-\infty}^{\infty} f(t) e^{-j2\pi ut} dt = \int_{-W/2}^{W/2} A e^{-j2\pi ut} dt = \dots$$

$$= \frac{A}{j2\pi u} \left[e^{j\pi u W} - e^{-j\pi u W} \right] = AW \frac{\sin(\pi u W)}{\pi u W} = AW \text{sinc}(\pi u W)$$

