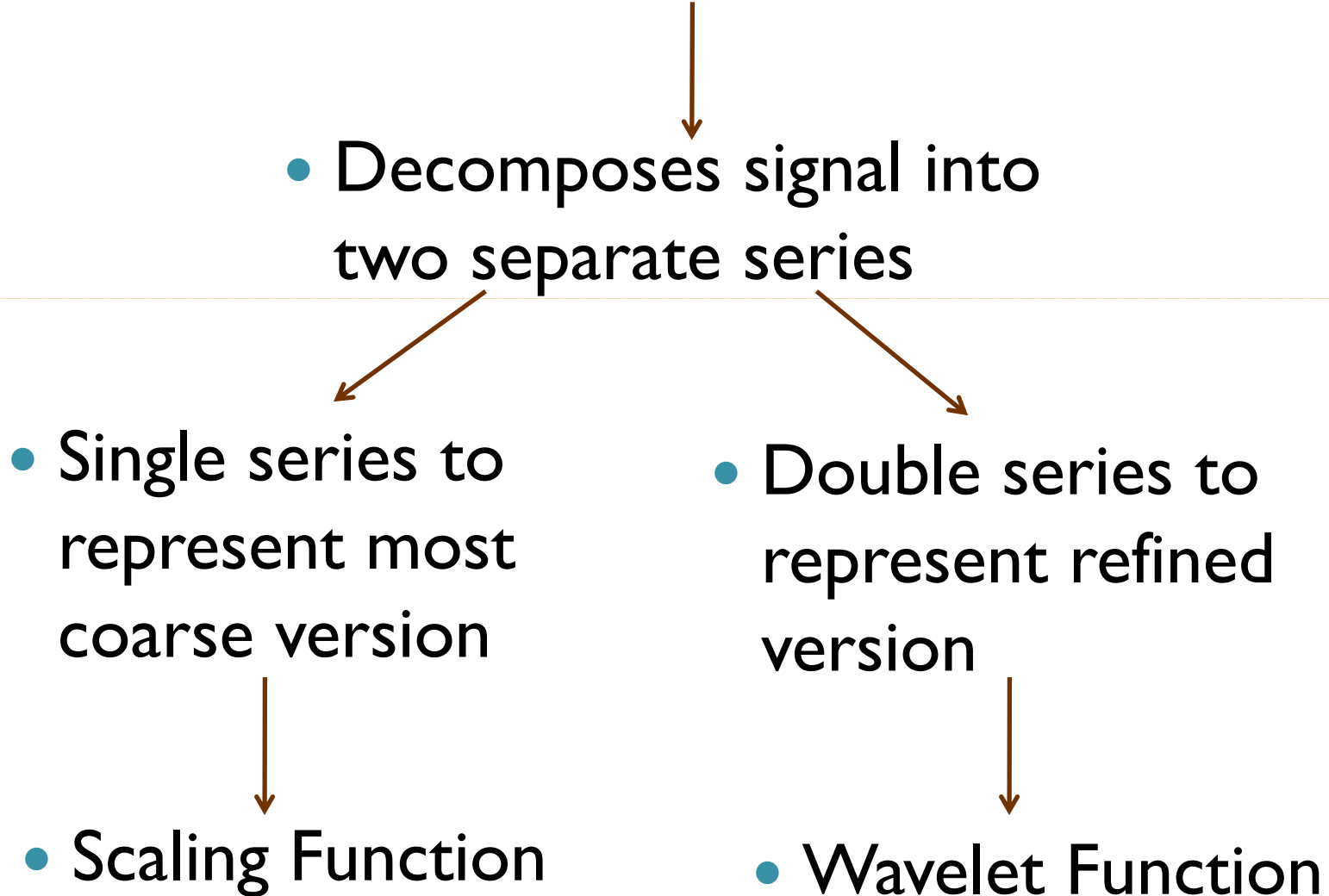




Lecture 50 – Wavelet Applications

Dr. Aditya Abhyankar

Wavelet Transform

- 
- ```
graph TD; A[Wavelet Transform] --> B[Decomposes signal into two separate series]; B --> C[Single series to represent most coarse version]; B --> D[Double series to represent refined version]; C --> E[Scaling Function]; D --> F[Wavelet Function]
```
- Decomposes signal into two separate series

- Single series to represent most coarse version

- Scaling Function

- Double series to represent refined version

- Wavelet Function



# Framework

- Gave us power to move up or down the ladder
- We can now indeed zoom-in or zoom-out of any part of the signal
- This makes the entire analysis ‘scalable’!!
- Scalability stems out of multi-resolution framework !


# Framework


$$\phi(t) = \sqrt{2} \sum_k h_k \phi(2t - k)$$

---

# Framework

$$\phi(t) = \sqrt{2} \sum_k h_k \phi(2t - k)$$


$$\phi(1t) = \phi(2^0 t) \in V_0$$


$$\phi(2t - k) = \phi(2^1 t - k) \in V_1$$

# Framework

$$V_0 \subset V_1$$

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# Framework

$$V_1 = V_0 \oplus W_0$$

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# Framework

$$\phi\left(\frac{t}{2}\right) = \sqrt{2} \sum_k h_k \phi(t - k)$$

$$\phi\left(\frac{t}{2}\right) = \phi(2^{-1}t) \in V_{-1}$$

$$\phi(t - k) = \phi(2^0 t - k) \in V_0$$

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$$V_{-1} \subset V_0$$

# Framework

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$$V_{-1} \subset V_0$$

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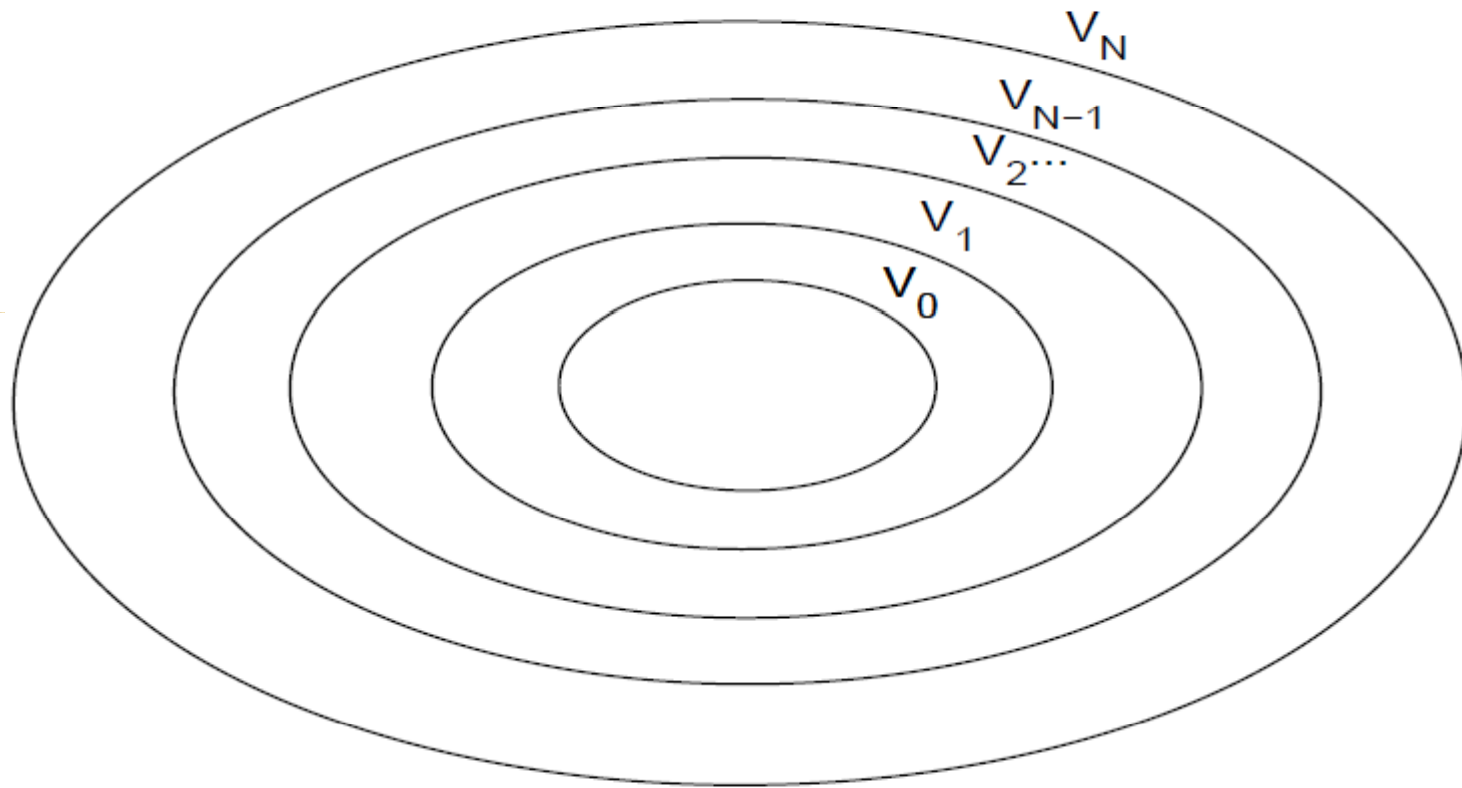
$$\phi(t - k) = \phi(2^0 t - k) \in V_0$$

$$\varphi\left(\frac{t}{2}\right) = \sqrt{2} \sum_k g_k \phi(t - k)$$

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$$\phi(t - k) = \phi(2^0 t - k) \in V_0$$

# Framework



$$\dots V_{-3} \subset V_{-2} \subset V_{-1} \subset V_0 \subset V_1 \subset V_2 \subset V_3 \subset V_4 \dots$$

# Framework

$$f_j(x) \in V_j, \text{ Scale } \frac{1}{2^j}$$

$$\{2^{j/2} \phi(2^j x - k)\}_k$$

$$f_j(x) = \sum_k \alpha_{j,k} 2^{j/2} \phi(2^j x - k)$$

$$\alpha_{j,k} = \int_{-\infty}^{\infty} f_j(x) 2^{j/2} \phi(2^j x - k) dx$$

# Framework

Normalization

$$f_j(x) \in V_j, \text{ Scale } \frac{1}{2^j}$$

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# Framework

Normalization

$$f_j(x) \in V_j, \text{ Scale } \frac{1}{2^j}$$

$$\{2^{j/2} \phi(2^j x - k)\}_k \rightarrow \text{Orthonormal basis}$$

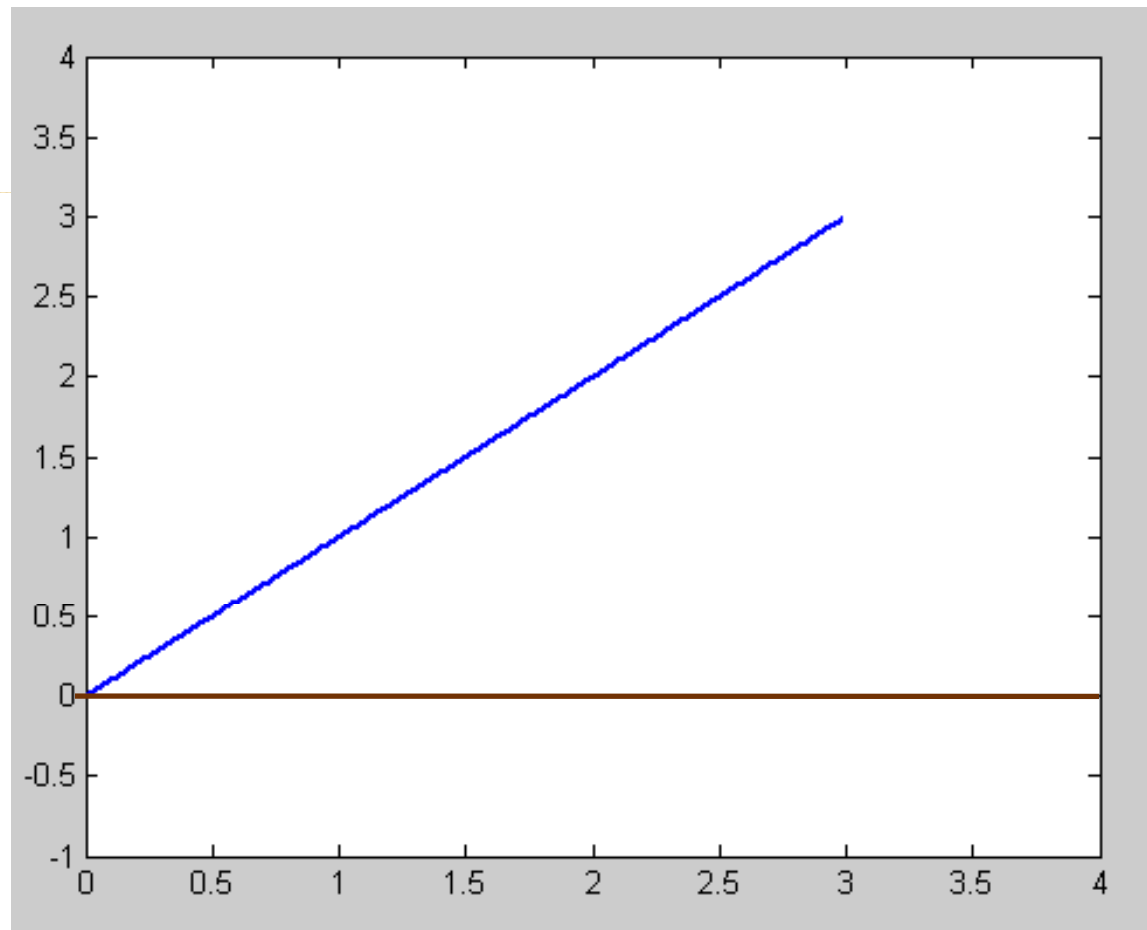
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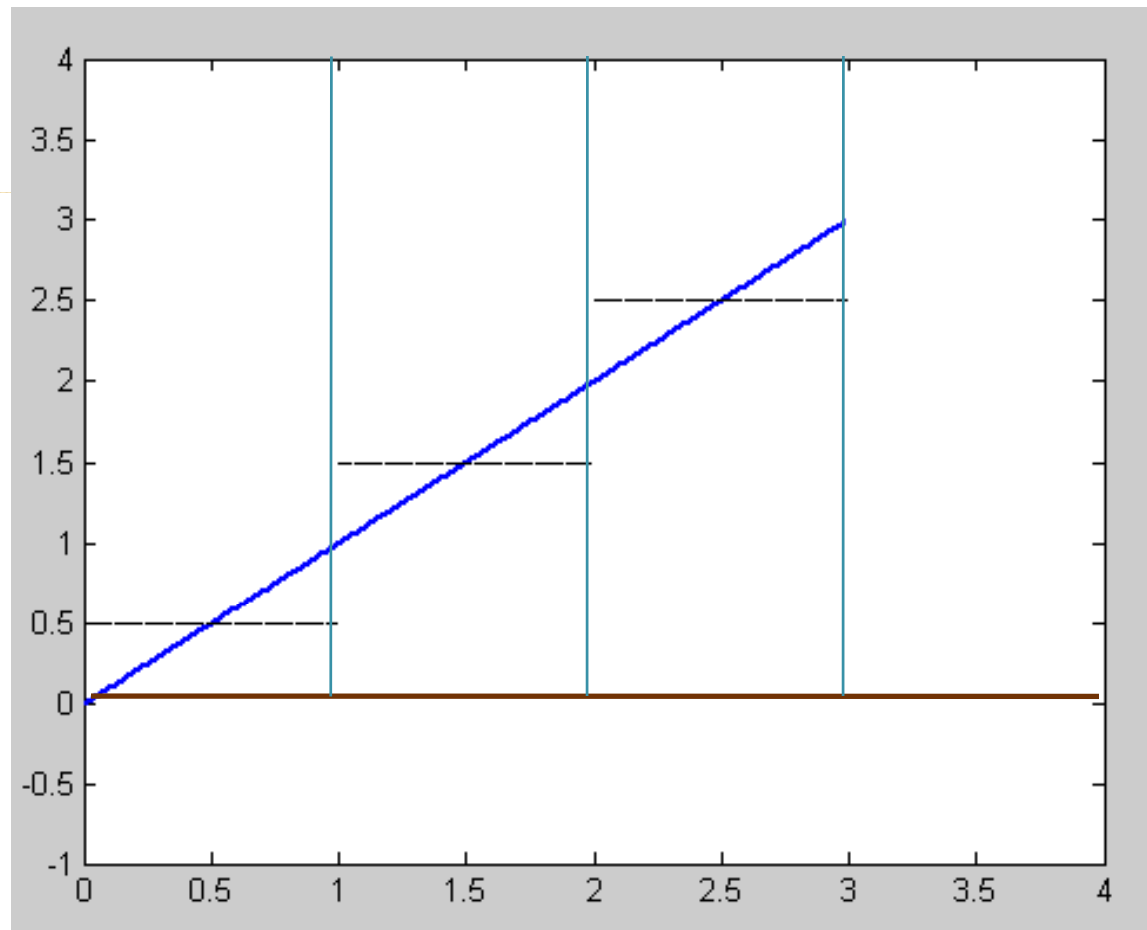
# Test Signal

$$f(x) = \begin{cases} x, & 0 \leq x \leq 3 \\ 0, & \text{otherwise} \end{cases}$$



# Test Signal

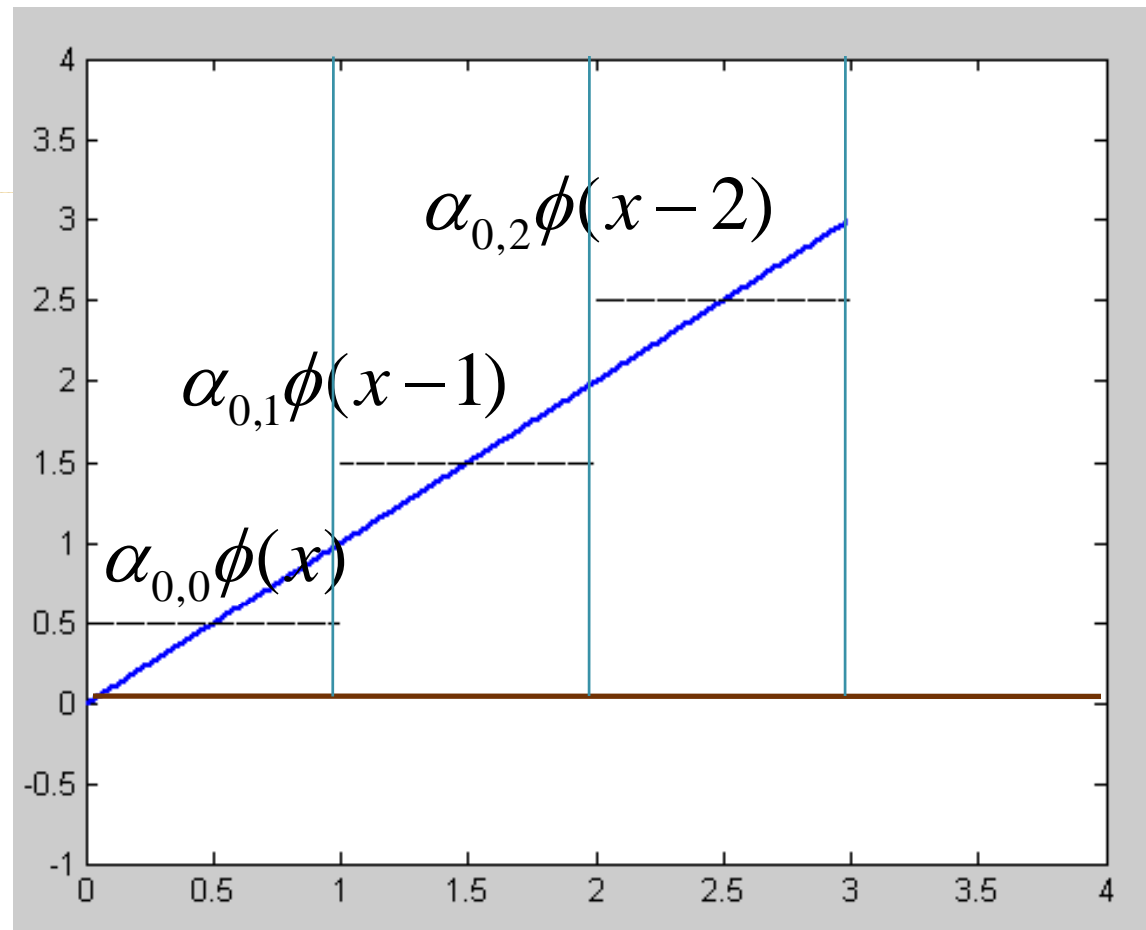
$$f(x) = \begin{cases} x, & 0 \leq x \leq 3 \\ 0, & \text{otherwise} \end{cases}$$



$$f_0(x) \in V_0$$

# Test Signal

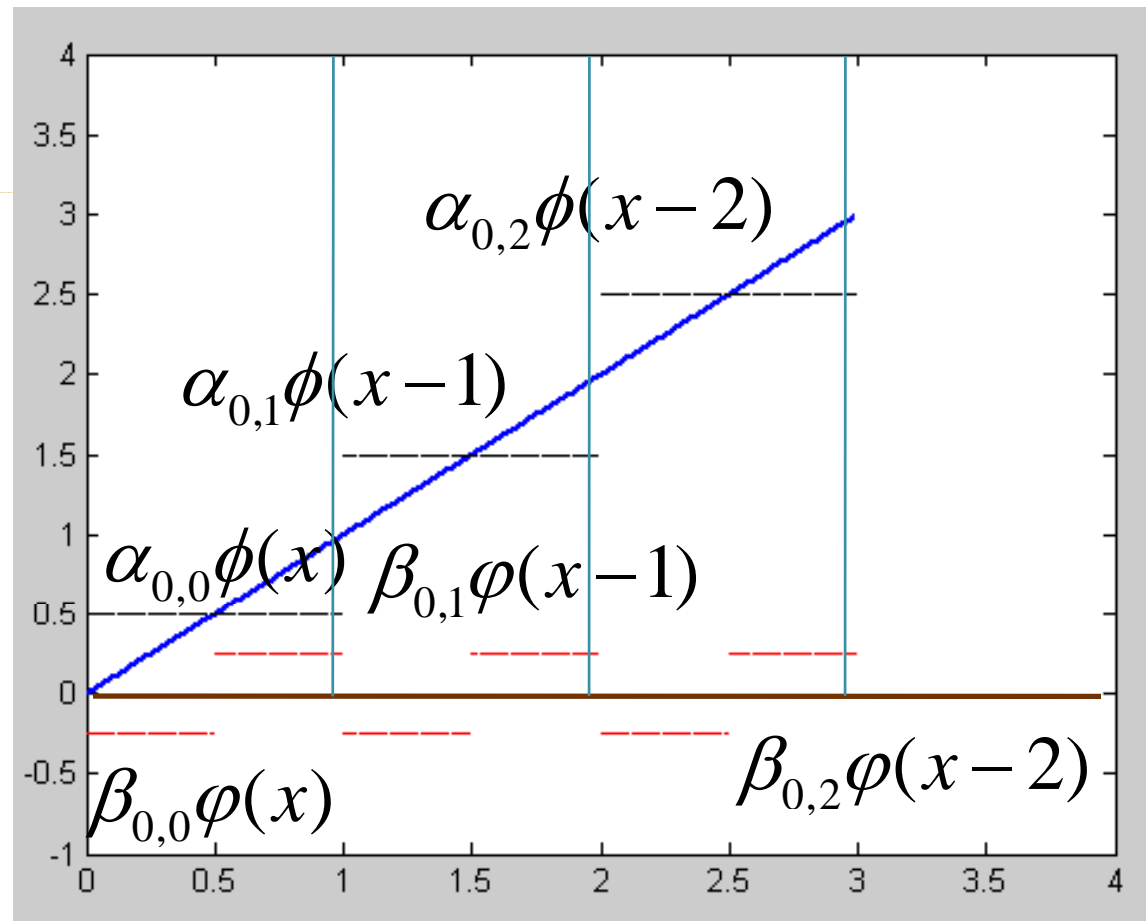
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$$f_0(x) \in V_0$$

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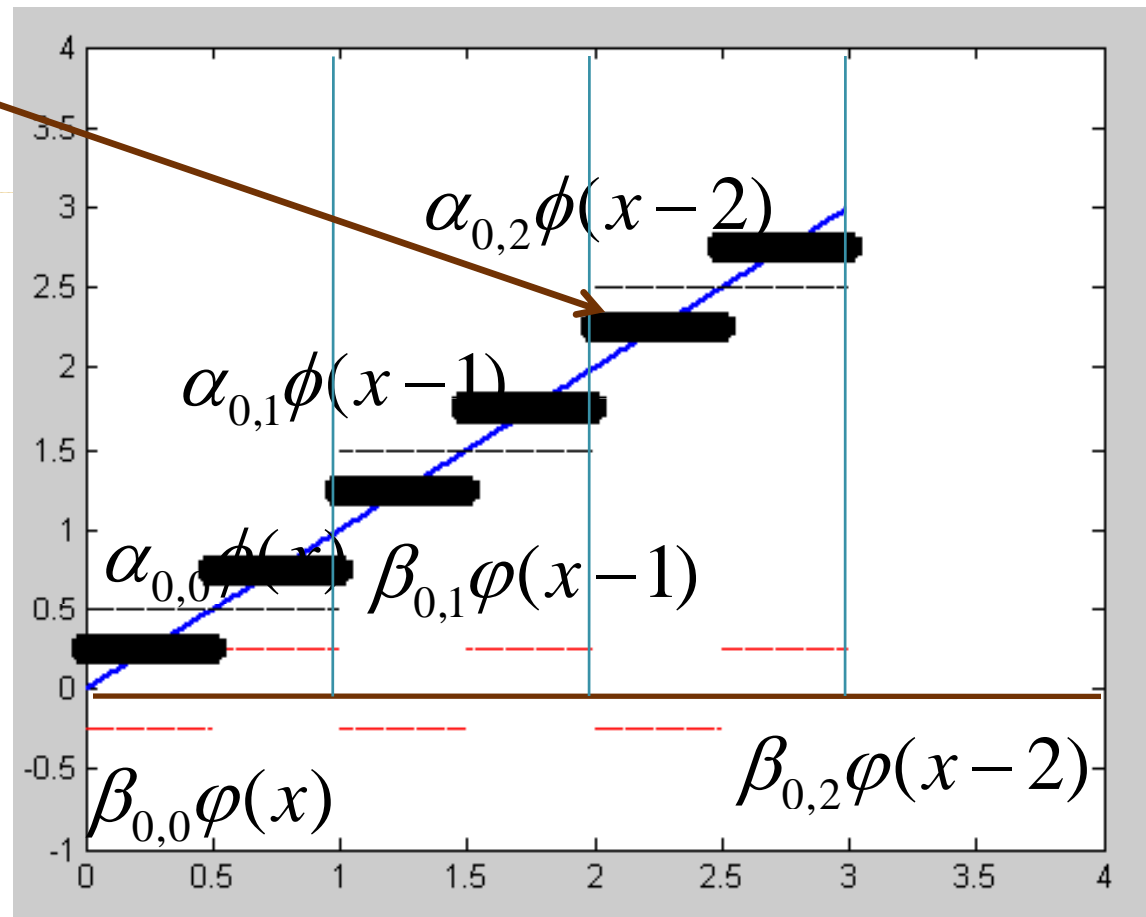
$$f_0(x) \in V_0$$

$$g_0(x) \in W_0$$

# Test Signal

$$f(x) = \begin{cases} x, & 0 \leq x \leq 3 \\ 0, & \text{otherwise} \end{cases}$$

$$f_1(x) \in V_1$$



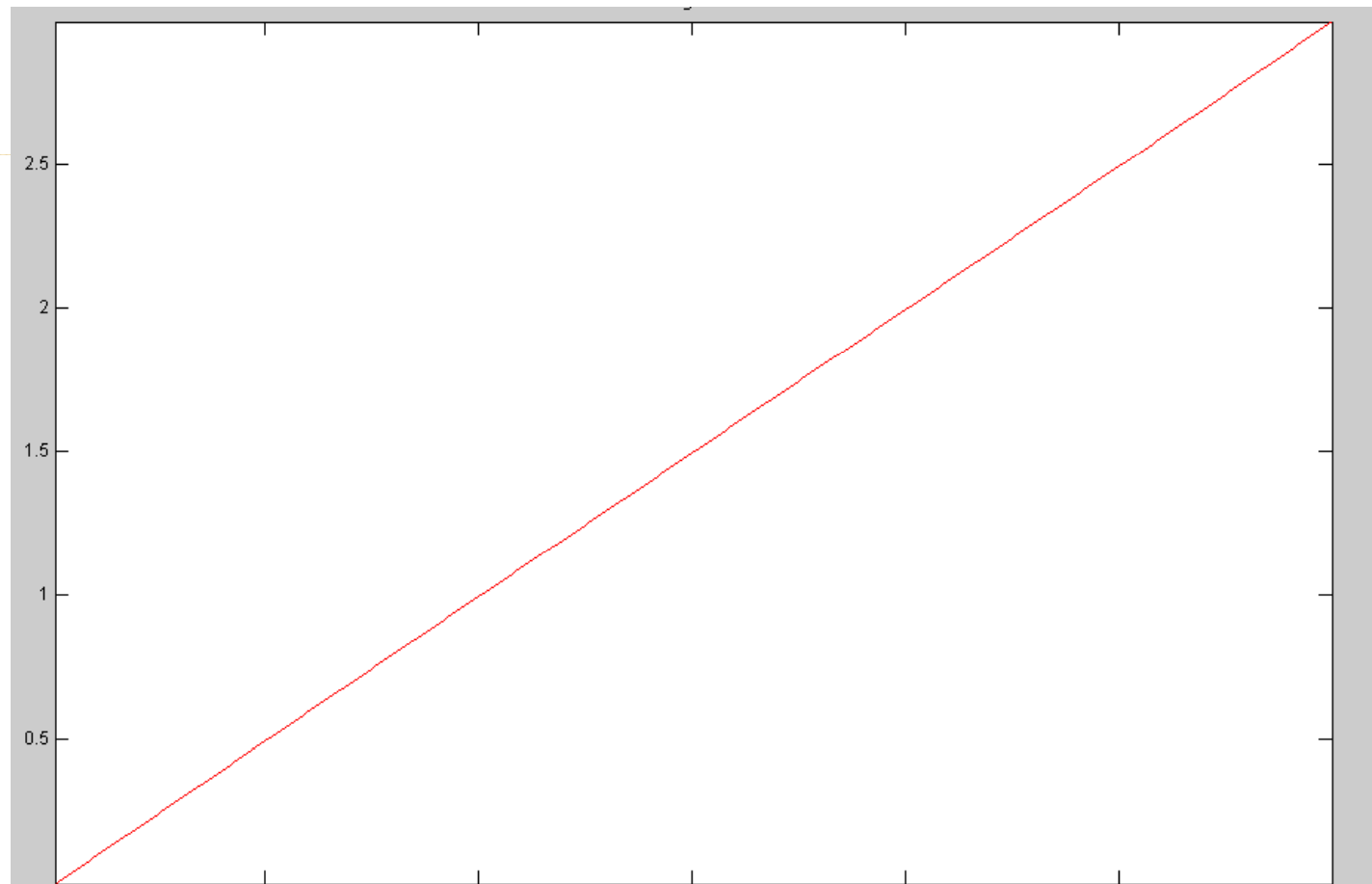
$$f_0(x) \in V_0$$

$$\oplus$$

$$g_0(x) \in W_0$$

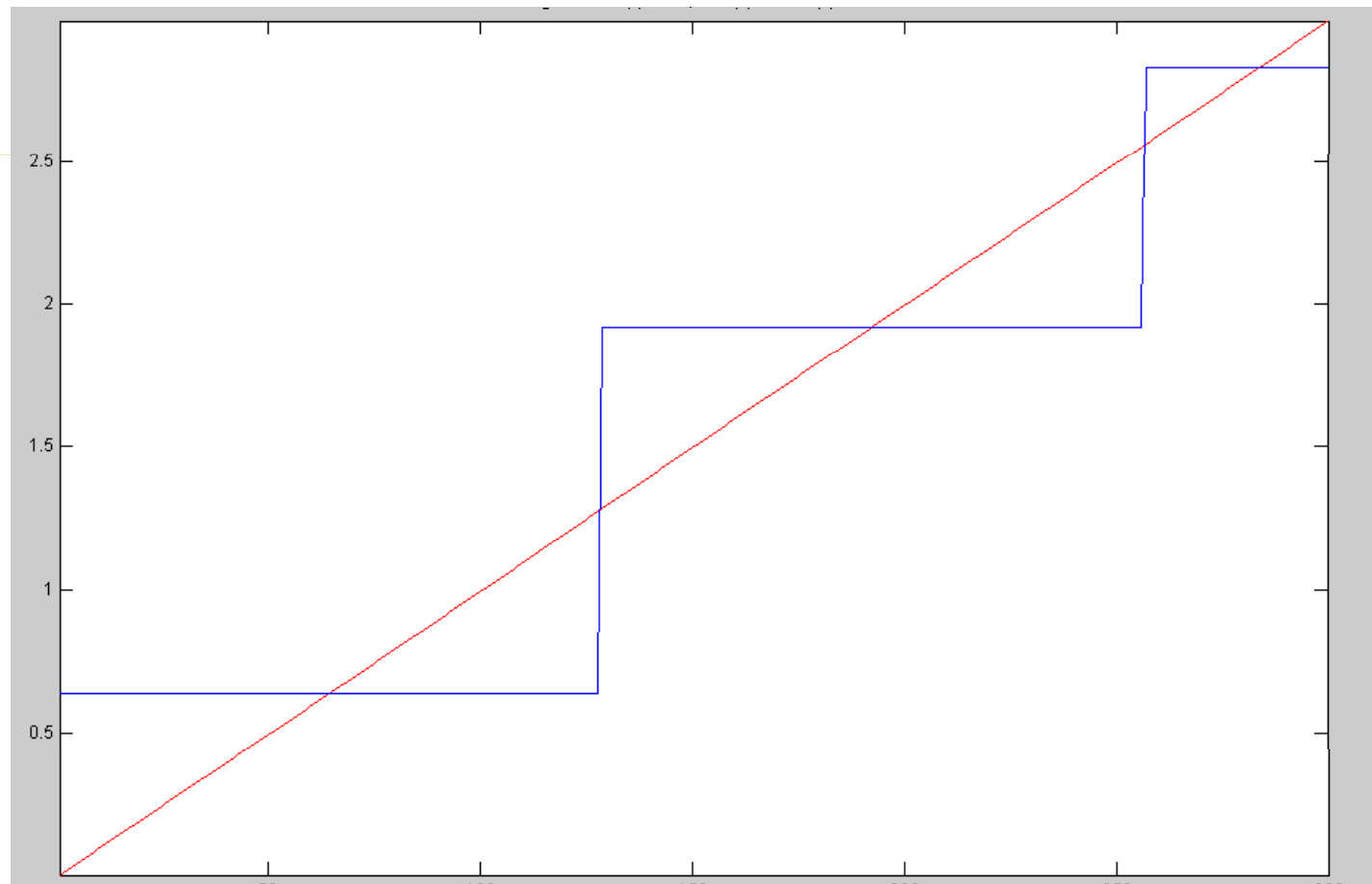
# Test Signal

$$f(x) = \begin{cases} x, & 0 \leq x \leq 3 \\ 0, & \text{otherwise} \end{cases}$$



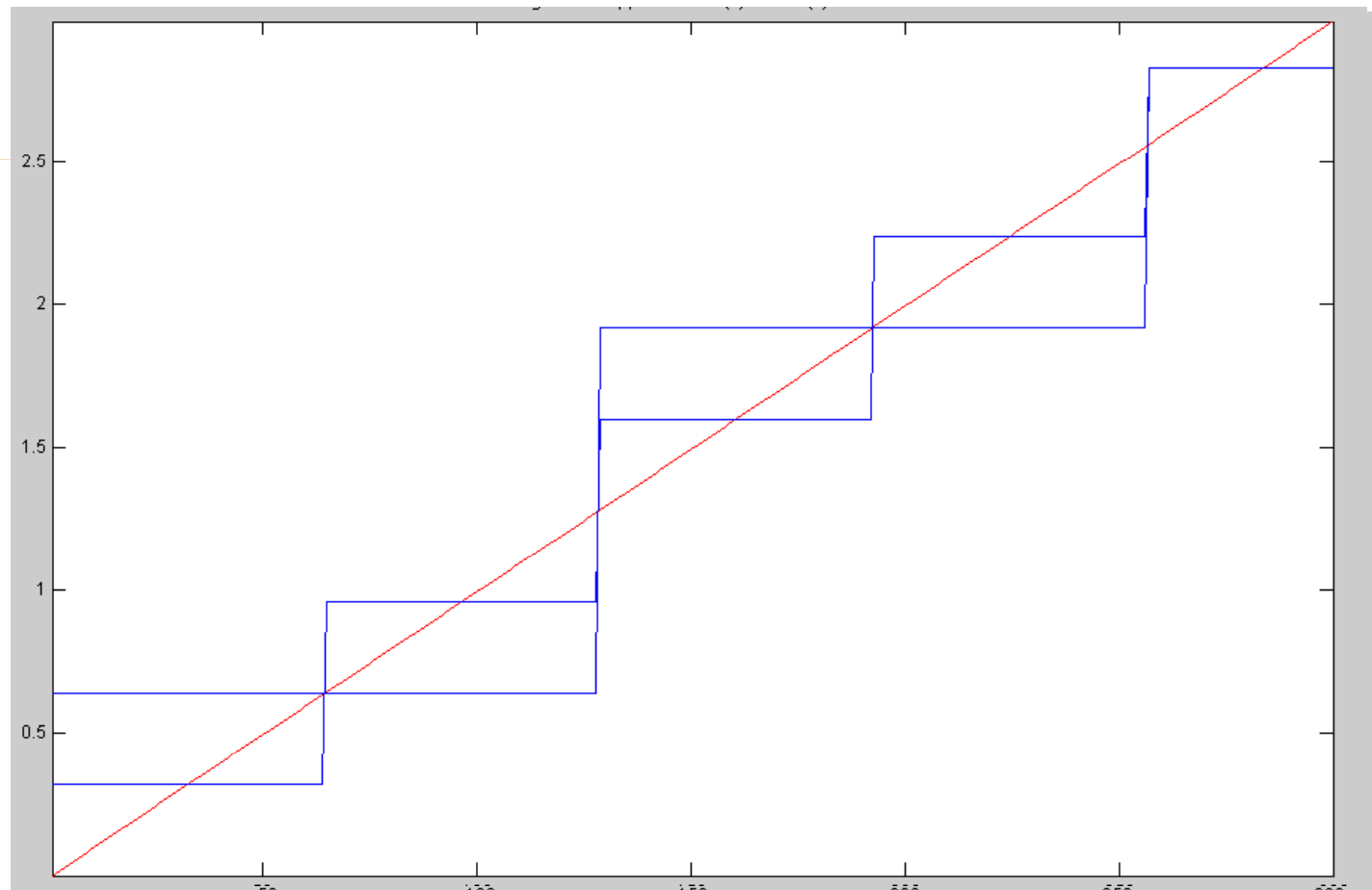
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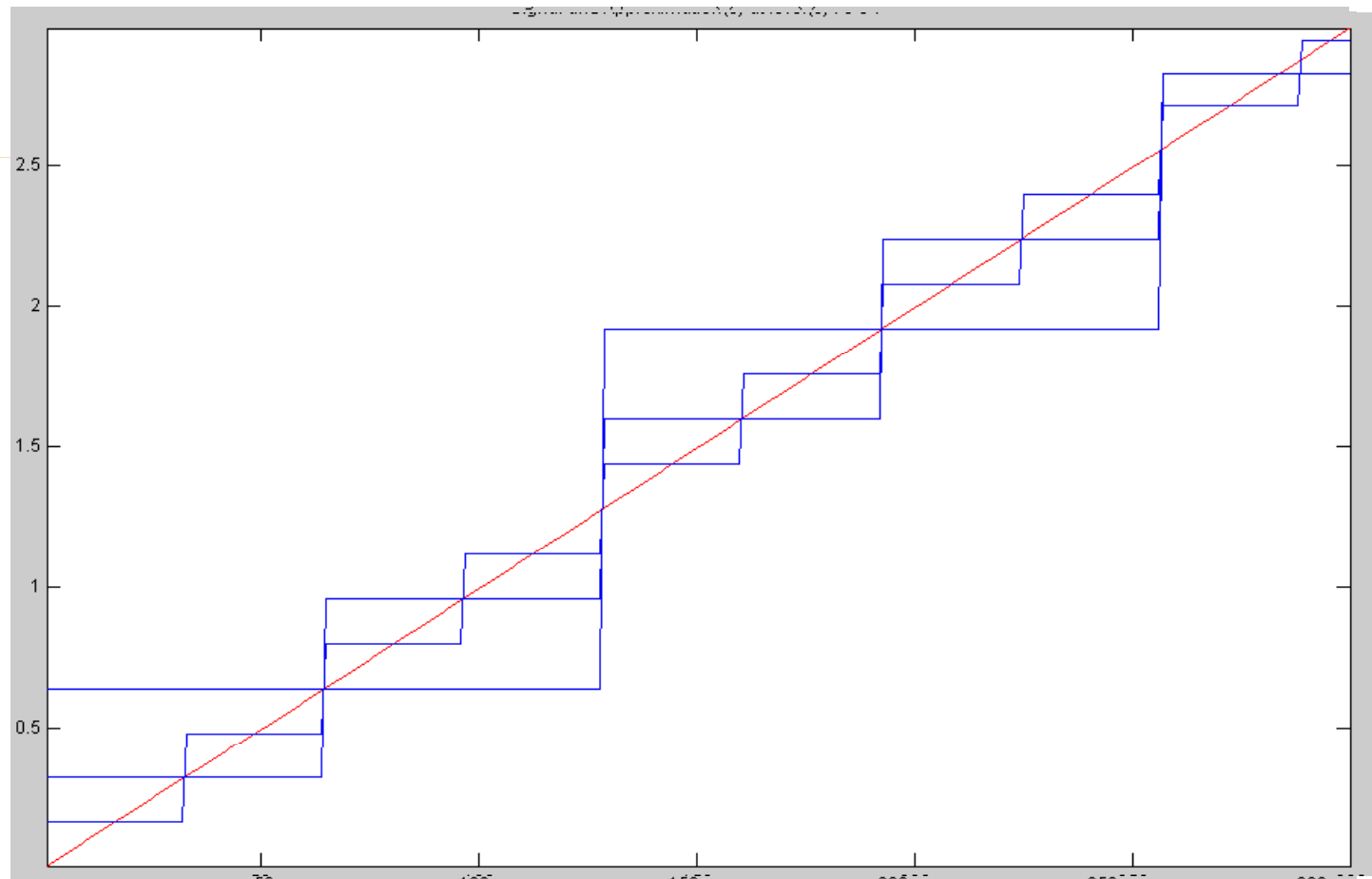
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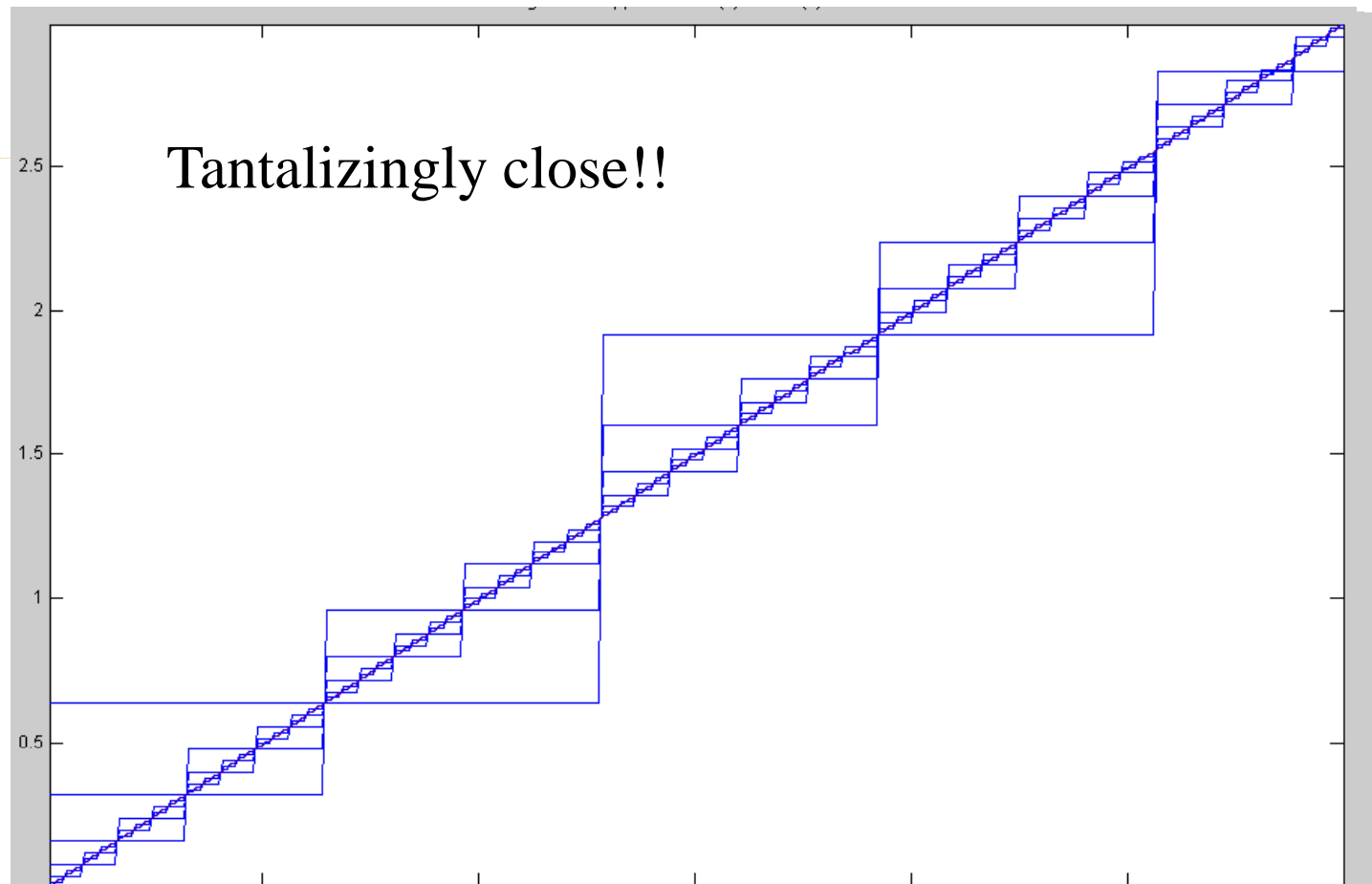
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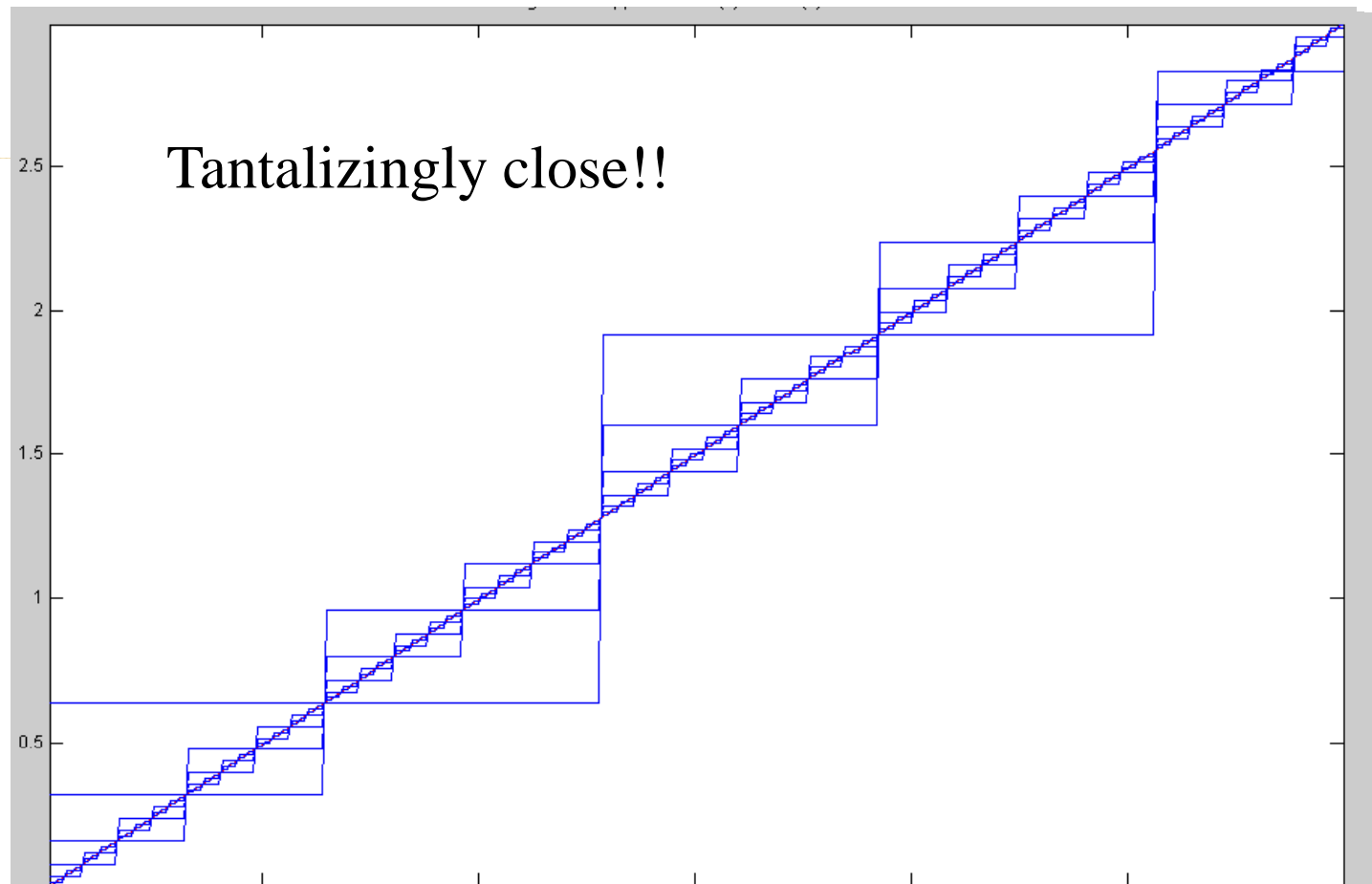
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# Framework

- Leads us to two questions
  - 1) How do we go about selecting the mother wavelet and scale of analysis?
  - 2) What is the procedure to calculate scaling and wavelet coefficients?



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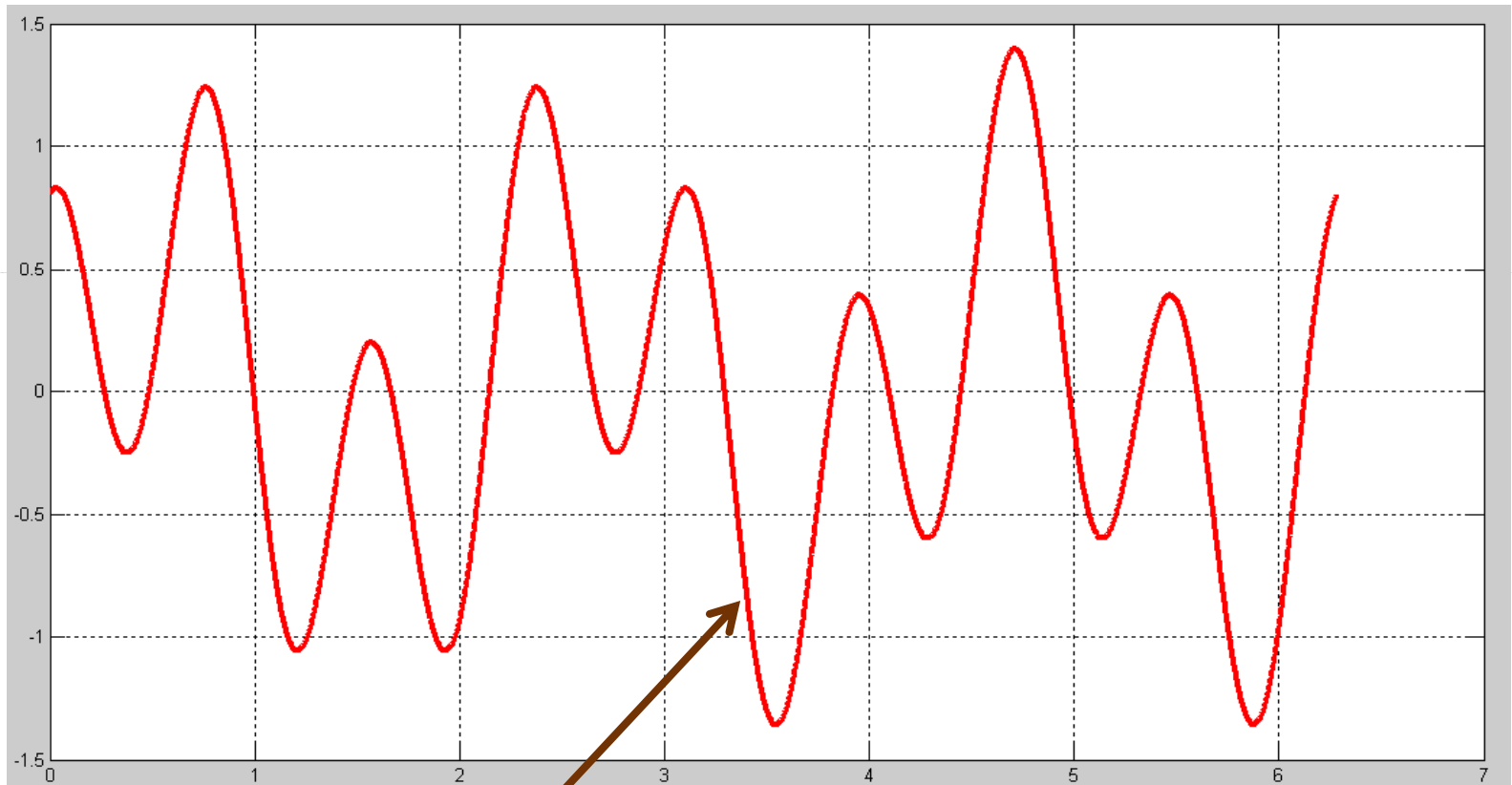
# Framework

- Leads us to two questions
  - 1) How do we go about selecting the **mother wavelet** and scale of analysis?
  - 2) What is the procedure to calculate scaling and wavelet coefficients?

**Vanishing moments**

**Correlation**

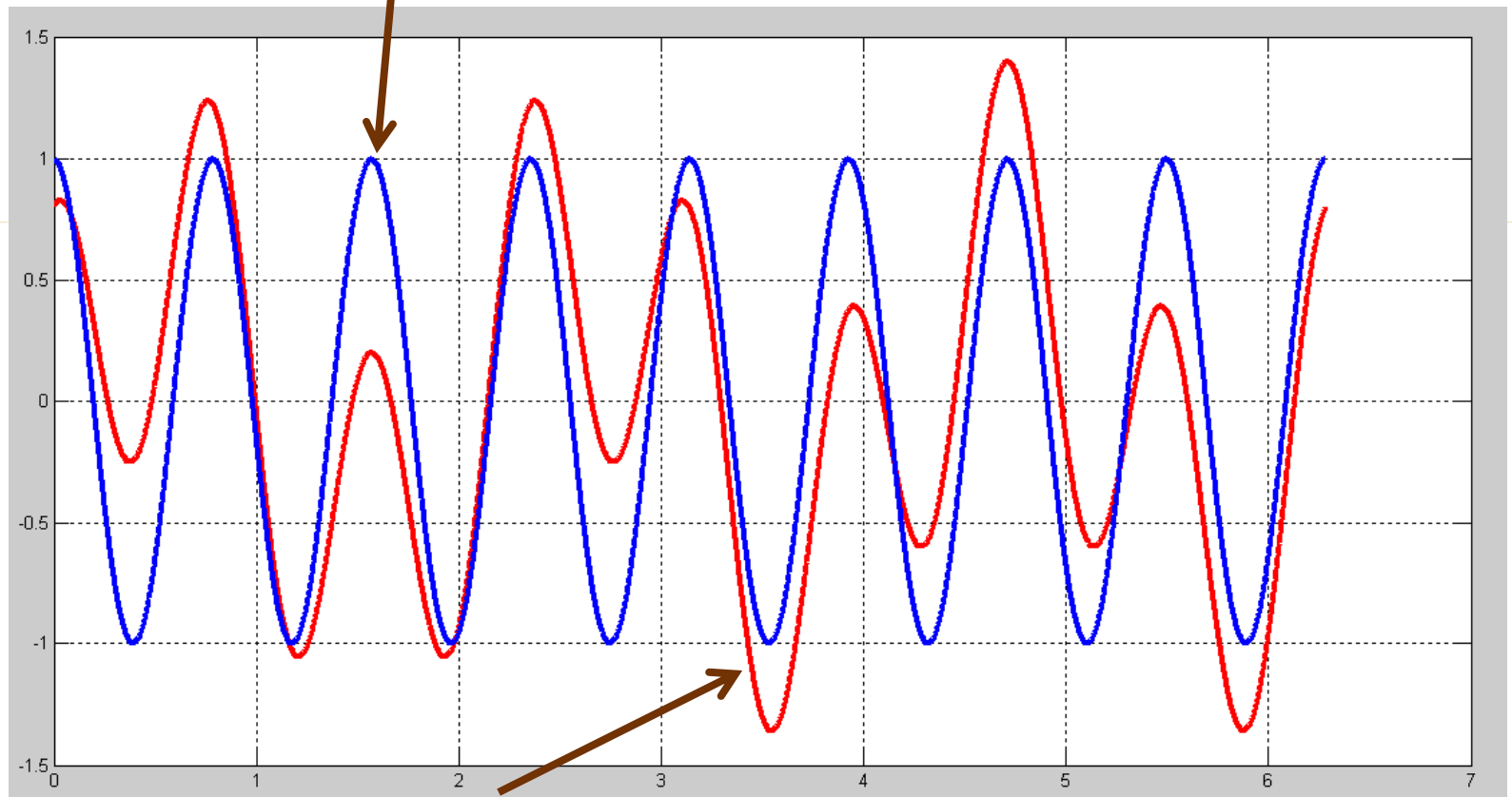
# How Fourier Works!!



$$y[n] = 0.6 \sin(2\pi 3n) + 0.8 \cos(2\pi 8n)$$

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$$\cos(2\pi 8n)$$

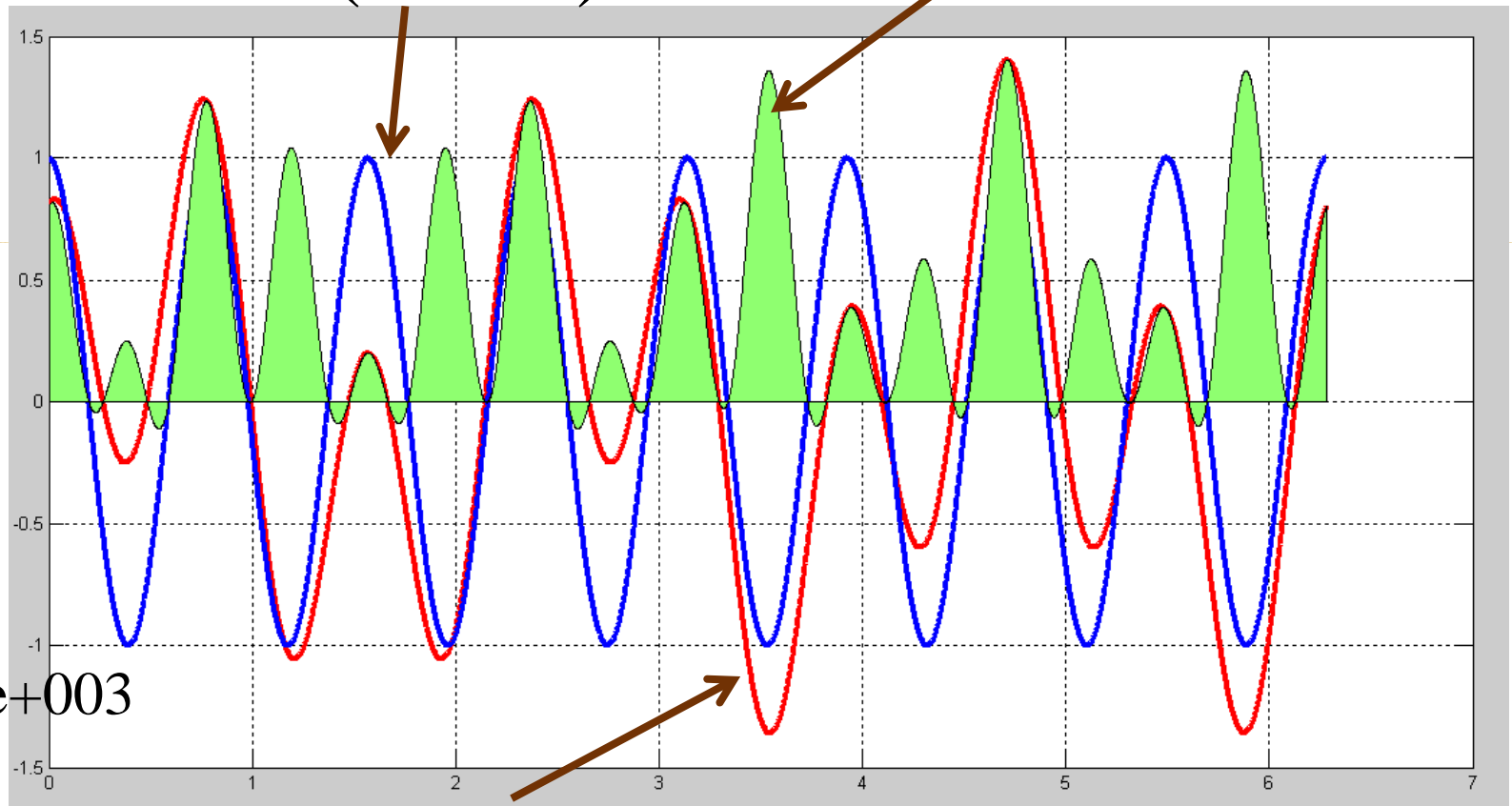


$$y[n] = 0.6 \sin(2\pi 3n) + 0.8 \cos(2\pi 8n)$$



# How Fourier Works!!

$$\cos(2\pi 8n) \quad \langle y[n], \cos(2\pi 8n) \rangle$$



2.0004e+003

$$y[n] = 0.6 \sin(2\pi 3n) + 0.8 \cos(2\pi 8n)$$

# Moments

- The moment of order  $m$ , of function  $f(x)$  on  $(a,b)$  can be given as

$$M_m = \int_a^b x^m f(x) dx$$

# Application

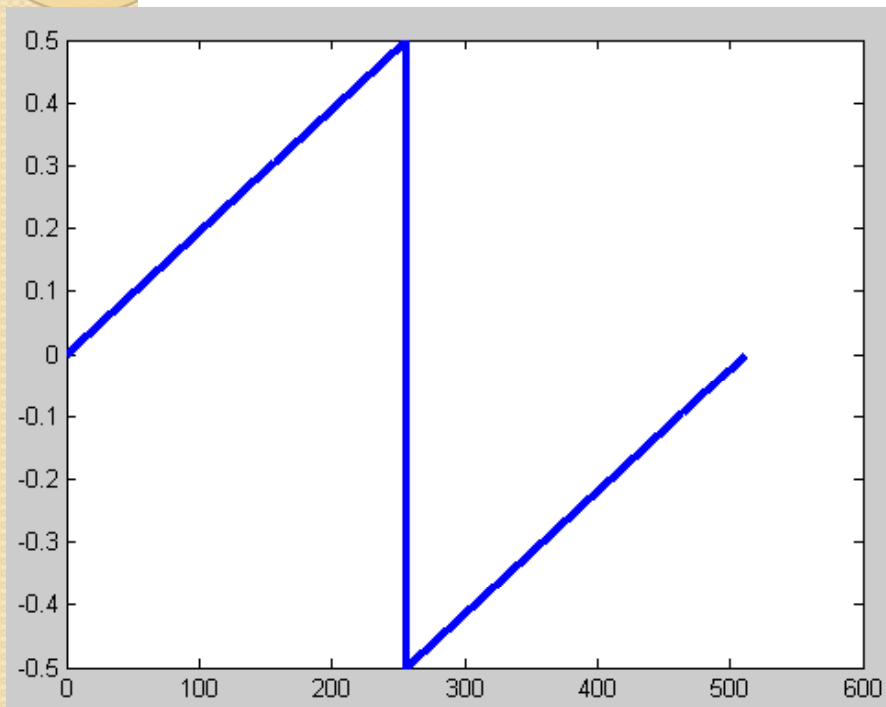
- Detecting hidden jump discontinuity
- Consider function

$$g(t) = \begin{cases} t, 0 \leq t < \frac{1}{2} \\ t - 1, \frac{1}{2} \leq t < 1 \end{cases}$$

- Clear jump at  $t=0.5$

# Application

- Detecting hidden jump discontinuity



$$g(t) = \begin{cases} t, 0 \leq t < \frac{1}{2} \\ t - 1, \frac{1}{2} \leq t < 1 \end{cases}$$

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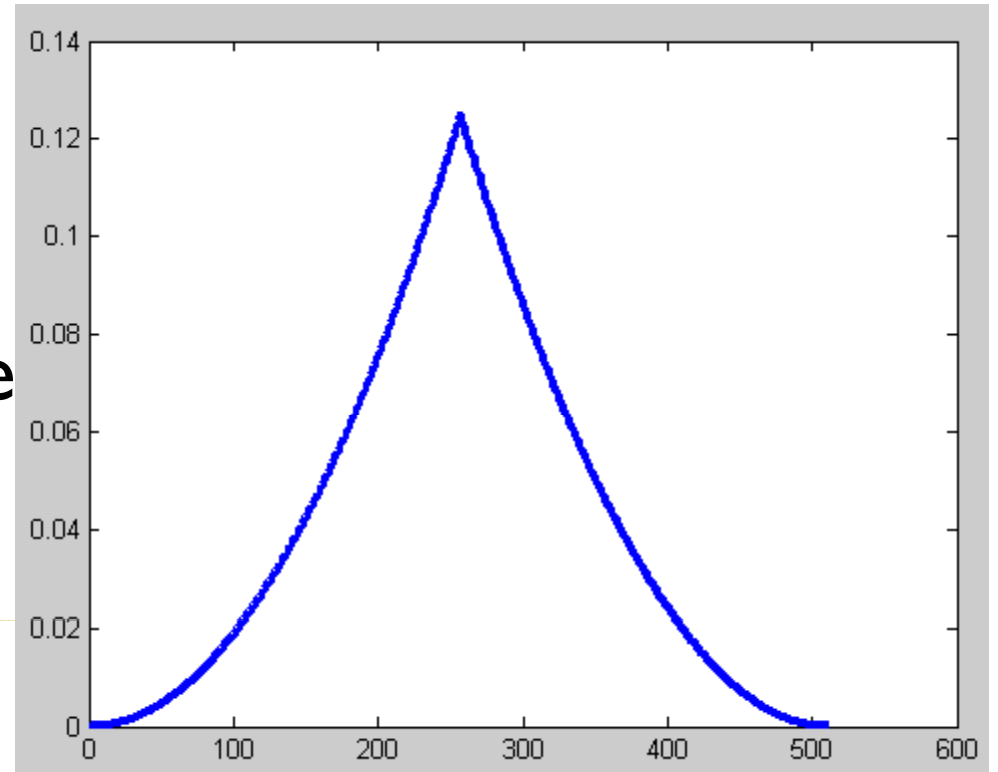
- Detecting hidden jump discontinuity
- Let's integrate

$$h(t) = \int g(t)dt = \begin{cases} \frac{t^2}{2}, 0 \leq t < \frac{1}{2} \\ \frac{t^2}{2} - t + \frac{1}{2}, \frac{1}{2} \leq t < 1 \end{cases}$$

- Cusp jump at  $t=0.5$

# Application

- Detecting hidden
- Let's integrate



$$h(t) = \int g(t)dt = \begin{cases} \frac{t^2}{2}, 0 \leq t < \frac{1}{2} \\ \frac{t^2}{2} - t + \frac{1}{2}, \frac{1}{2} \leq t < 1 \end{cases}$$

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# Application

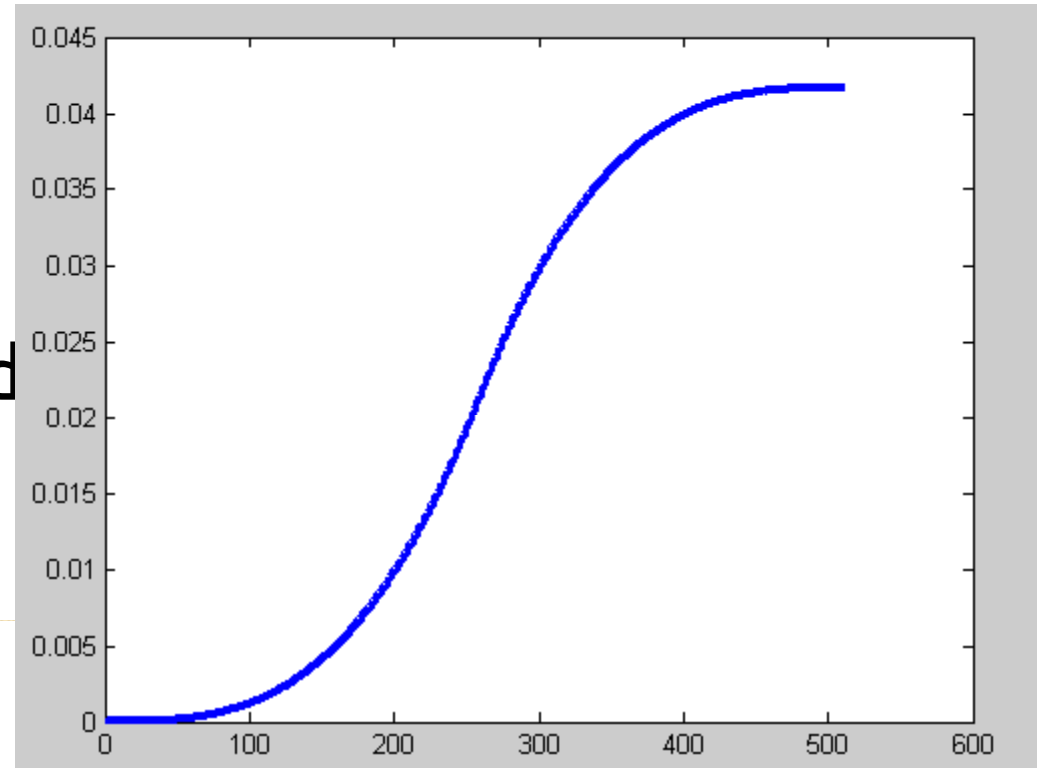
- Detecting hidden jump discontinuity
- Let's integrate again

$$f(t) = \int h(t)dt = \begin{cases} \frac{t^3}{6}, 0 \leq t < \frac{1}{2} \\ \frac{t^3}{6} - \frac{t^2}{2} + \frac{t}{2} - \frac{1}{8}, \frac{1}{2} \leq t < 1 \end{cases}$$

- Appears smooth to eye

# Application

- Detecting hidden
- Let's integrate



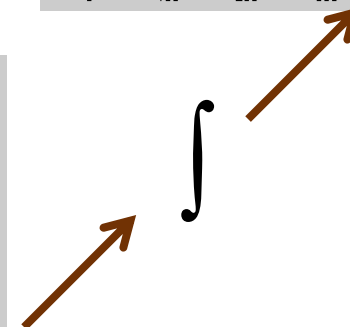
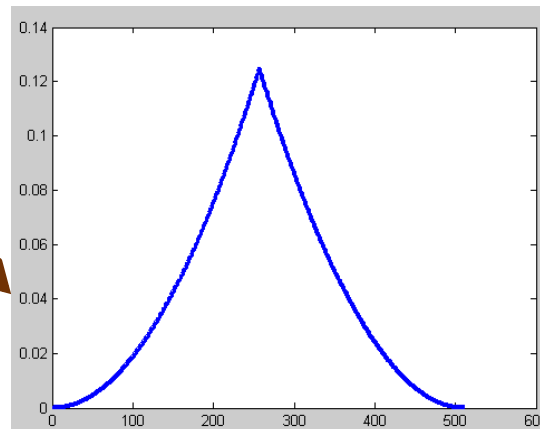
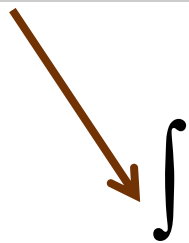
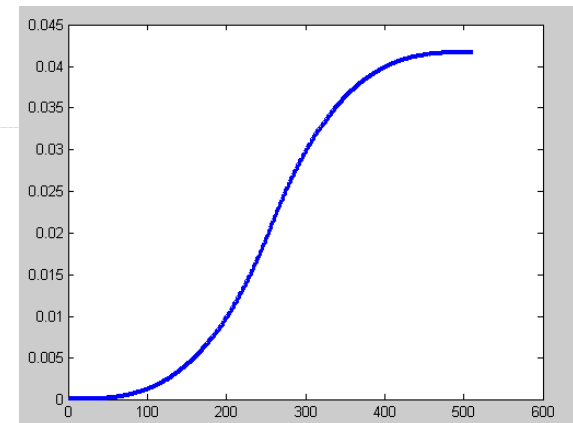
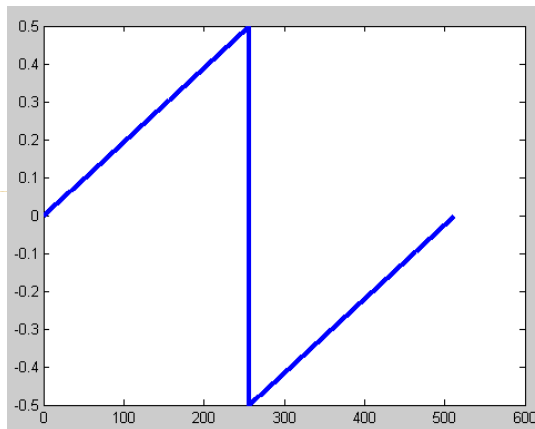
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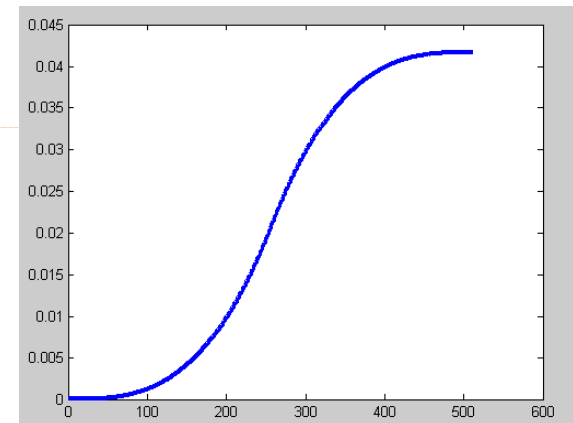
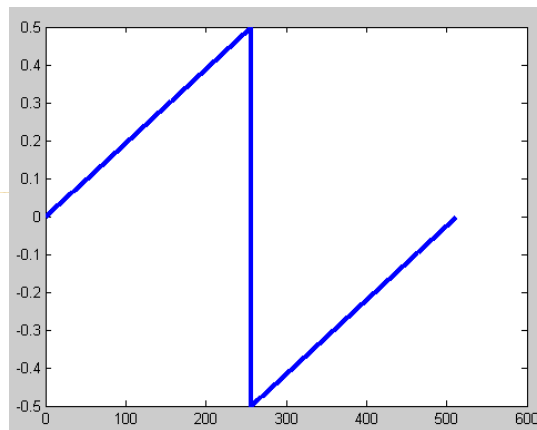
# Application

- Detecting hidden jump discontinuity



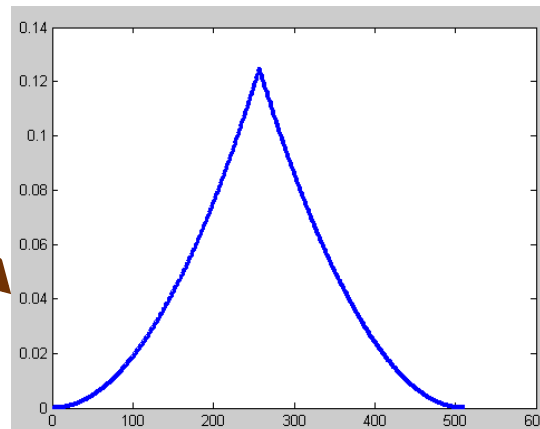
# Application

- Detecting hidden jump discontinuity



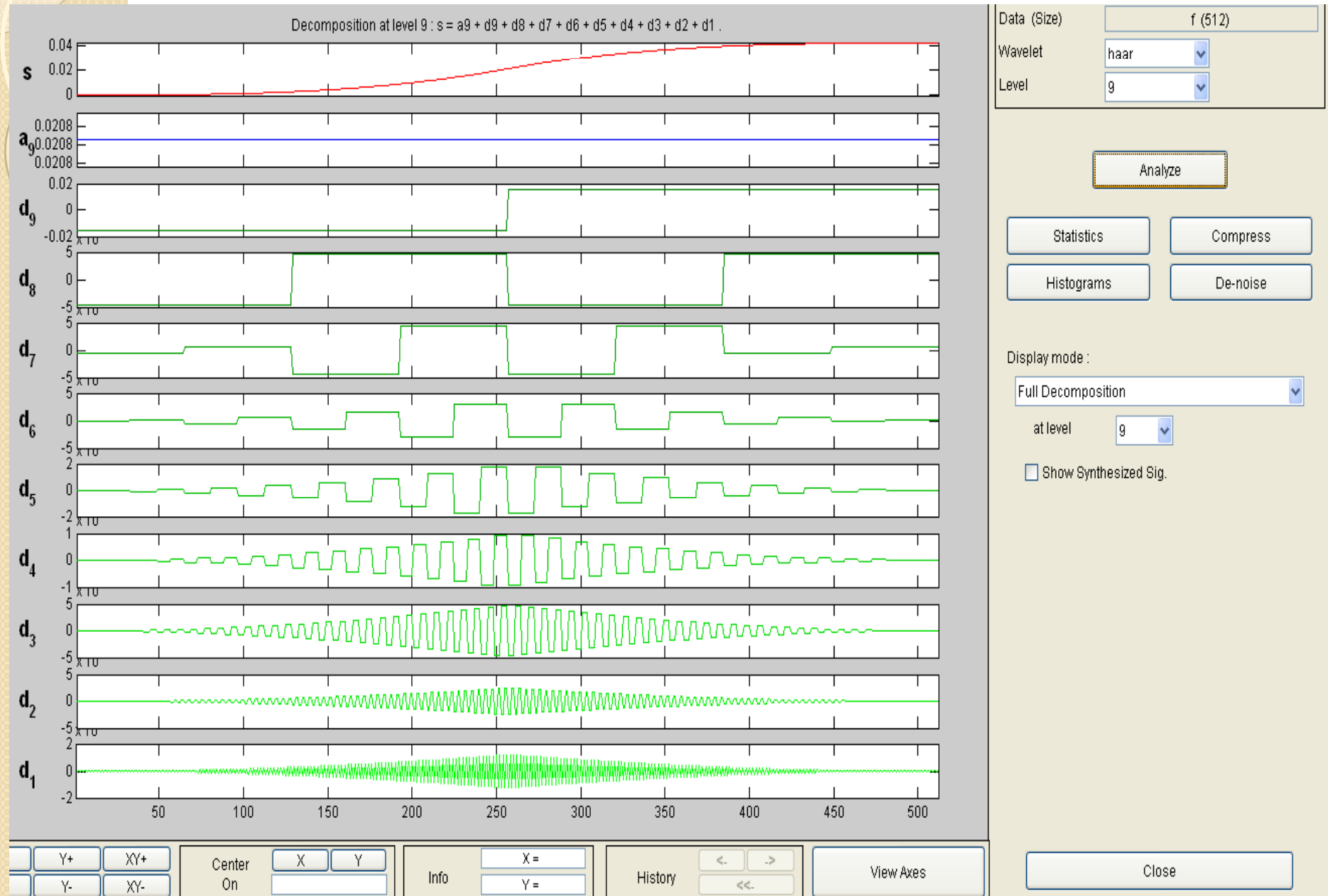
While selecting  
mother wavelet  
At least  $M0=M1=0$

$\int$

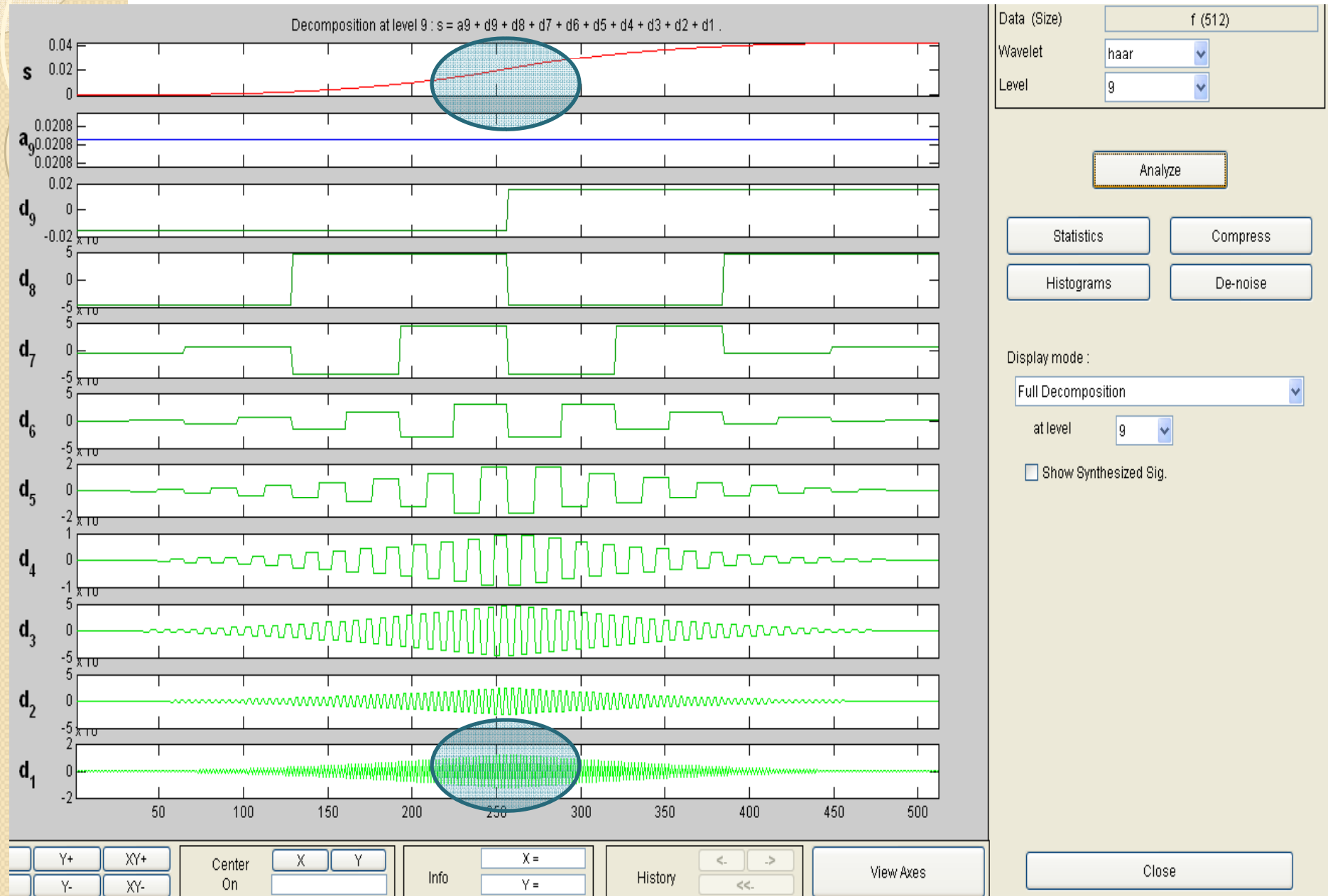


$\int$

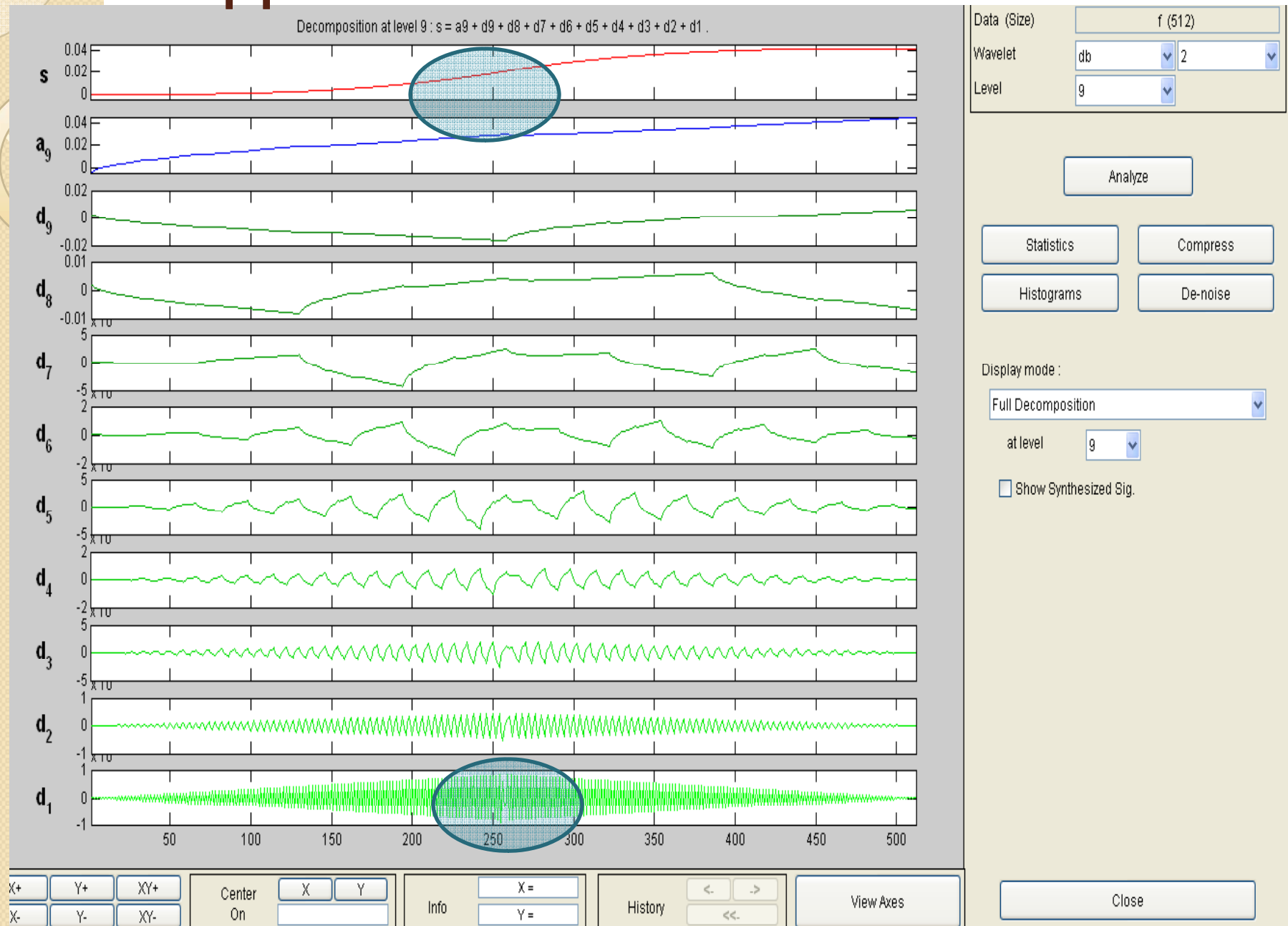
# Application



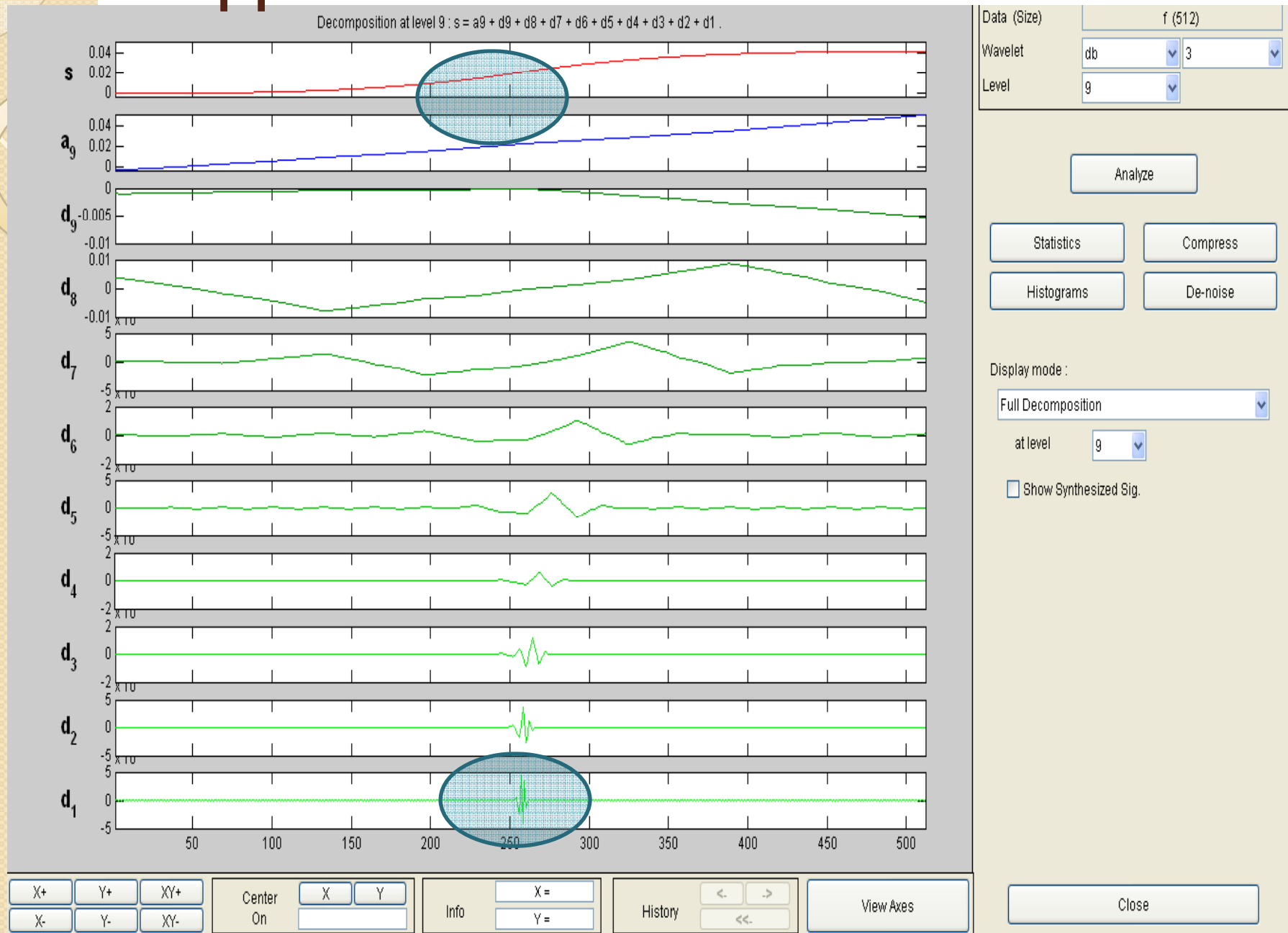
# Application



# Application



# Application



# Framework

- Leads us to two questions
  - 1) How do we go about selecting the **mother wavelet** and scale of analysis?
  - 2) What is the procedure to calculate scaling and wavelet coefficients?

**Vanishing moments**

**Correlation**



# Framework

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# Framework

- Leads us to two questions
  - 1) How do we go about selecting the mother wavelet and scale of analysis?
  - 2) What is the procedure to calculate scaling and wavelet coefficients?

# Coefficients

- Who gives us coefficients of scaling equation?
- Haar

$$\psi(t) = \begin{cases} 1, & 0 \leq x < \frac{1}{2} \\ -1, & \frac{1}{2} \leq x < 1 \\ 0, & \text{otherwise} \end{cases}$$

$$\phi(t) = \begin{cases} 1, & 0 \leq x < 1 \\ 0, & \text{otherwise} \end{cases}$$

# Framework

$$V_1 = V_0 \oplus W_0$$

$$V_0 \subset V_1$$

$$\phi(t) = \sqrt{2} \sum_k h_k \phi(2t - k)$$

$$\phi(1t) = \phi(2^0 t) \in V_0$$

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$$\phi(2t - k) = \phi(2^1 t - k) \in V_1$$

# In search of coefficients

- In search of scaling equation coefficients!!!

$$\phi(t) = \sqrt{2} \sum_k h_k \phi(2t - k)$$

$$\varphi(t) = \sqrt{2} \sum_k (-1)^k h_{1-k} \phi(2t - k)$$



# In search of coefficients

- In search of scaling equation coefficients!!!
- 
- We can think of using three guiding theorems !

# In search of coefficients

- We can think of using three guiding theorems !
  - Theorem I:
- 

For the scaling equation  $\phi(x) = \sum_k h_k \sqrt{2} \phi(2x - k)$ , with non-vanishing coefficients  $\{h_k\}_{k=N}^M$  only for  $N \leq k \leq M$ , its  $\phi(x)$  is with a compact support contained in interval  $[N, M]$

# In search of coefficients

- We can think of using three guiding theorems !
  - Theorem 2:
- 

If the scaling function  $\phi(x)$  has compact support on  $0 \leq x \leq N - 1$  and if,  $\{\phi(x - k)\}$  are linearly independent, then  $h_n = h(n) = 0$ , for  $n < 0$  and  $n > N - 1$ .

Hence  $N$  is the length of the sequence.

# In search of coefficients

- We can think of using three guiding theorems !
- Theorem 3:

If the scaling coefficients  $\{h_k\}$  satisfy the condition for existence and orthogonality of  $\phi(x)$ , then

$$\phi(x) = \sum_k g_k \sqrt{2} \phi(2x - k)$$

where,  $g_k = \pm(-1)^k h_{N-k}$

$$\text{and, } \int_{-\infty}^{\infty} \phi(x-l)\phi(x-k)dx = \delta_{l,k} = 0, l \neq k$$



# Properties of scaling coefficients

1.  $\sum h_k = \sqrt{2}$

2.  $\sum h_{2k} = \frac{1}{\sqrt{2}}$

---

3.  $\sum h_{2k+1} = \frac{1}{\sqrt{2}}$

# Properties of scaling coefficients

$$4. \quad \sum |h_k|^2 = 1$$

$$5. \quad \sum h_{k-2l} h_k = \delta_{l,0}$$

$$6. \quad \sum 2h_{k-2l} h_{k-2j} = \delta_{l,j}$$

Lets derive PI 1.  $\sum h_k = \sqrt{2}$

$$\phi(t) = \sum_k h_k \sqrt{2} \phi(2t - k)$$

$$\begin{aligned} \int_{-\infty}^{\infty} \phi(t) dt &= \int_{-\infty}^{\infty} \left[ \sum_k h_k \sqrt{2} \phi(2t - k) \right] dt \\ &= \sum_k h_k \sqrt{2} \int_{-\infty}^{\infty} \phi(2t - k) dt \end{aligned}$$

Lets derive PI 1.  $\sum h_k = \sqrt{2}$

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$$\int_{-\infty}^{\infty} \phi(t) dt = \int_{-\infty}^{\infty} \left[ \sum_k h_k \sqrt{2} \phi(2t - k) \right] dt$$

$$= \sum_k h_k \sqrt{2} \int_{-\infty}^{\infty} \phi(2t - k) dt$$

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$$\int_{-\infty}^{\infty} \phi(2t - k) dt = \frac{1}{2} \int_{-\infty}^{\infty} \phi(x) dx$$

Lets derive PI 1.  $\sum h_k = \sqrt{2}$

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$$= \sum_k h_k \sqrt{2} \int_{-\infty}^{\infty} \phi(2t - k) dt$$

$$\int_{-\infty}^{\infty} \phi(t) dt = \sum_k \frac{1}{2} h_k \sqrt{2} \int_{-\infty}^{\infty} \phi(x) dx$$

$$\int_{-\infty}^{\infty} \phi(t) dt = 1 \quad \text{Normalization !!}$$

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$$\int_{-\infty}^{\infty} \phi(t) dt = 1 \quad \text{Normalization !!}$$

$$1. \quad \sum h_k = \sqrt{2}$$

Lets derive P5 5.  $\sum h_{k-2l}h_k = \delta_{l,0}$

$$\phi(t) = \sum_k h_k \sqrt{2} \phi(2t - k)$$

change of the variable  $t$  in  $\phi(t)$  to  $x = 2^{j-1}t - l$ ,

$$\phi(2^{j-1}t - l) = \sum_{k=-\infty}^{\infty} h_k \sqrt{2} \phi(2(2^{j-1}t - l) - k)$$

change in the index  $k$  to  $m = k + 2l$ ,

$$\begin{aligned} \phi(2^{j-1}t - l) &= \sum_{m=-\infty}^{\infty} h_{m-2l} \sqrt{2} \phi(2^j t - 2l - m + 2l) \\ &= \sum_{m=-\infty}^{\infty} h_{m-2l} \sqrt{2} \phi(2^j t - m) \end{aligned}$$



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$$\phi(2^{j-1}t - l) = \sum_{m=-\infty}^{\infty} h_{k-2l} \sqrt{2} \phi(2^j t - k)$$

do this we multiply both sides of the above equation by  $\phi(t)$ , and integrate from  $-\infty$  to  $\infty$ , allowing the exchange of the integration

$$\int_{-\infty}^{\infty} \phi(t) \phi(2^{j-1}t - l) dt = \sum_{k=-\infty}^{\infty} h_{k-2l} \int_{-\infty}^{\infty} \sqrt{2} \phi(t) \phi(2^j t - k) dt$$

Lets derive P5 5.  $\sum h_{k-2l}h_k = \delta_{l,0}$

$$\int_{-\infty}^{\infty} \phi(t)\phi(t-l)dt = \delta_{0,l} = \sum_{k=-\infty}^{\infty} h_{k-2}h_k$$

$$\sum_{k=-\infty}^{\infty} h_{k-2l}h_k = \delta_{l,0}$$

Lets derive P4 4.  $\sum |h_k|^2 = 1$

special case of  $l = 0$  gives the property 4

$$\sum_{k=-\infty}^{\infty} h_{k-2l} h_k = \delta_{l,0}$$

$$\sum_{k=-\infty}^{\infty} h_k \overline{h_k} = \sum_{k=-\infty}^{\infty} |h_k|^2 = \delta_{0,0} 1$$

Lets derive P3 3.  $\sum h_{2k+1} = \frac{1}{\sqrt{2}}$

$$\delta_{l,0} = \sum_{k=-\infty}^{\infty} 2h_{k-2l}\overline{h_k} = 2 \sum_{k=-\infty}^{\infty} h_{k+2l}\overline{h_k}$$

$$\delta_{l,0} = 2 \sum_{k=-\infty}^{\infty} h_{2k+2l}\overline{h_{2k}} + 2 \sum_{k=-\infty}^{\infty} h_{2k+1+2l}\overline{h_{2k+1}}$$

$$\sum_{l=-\infty}^{\infty} \delta_{l,0} = 1 = 2 \sum_{l=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} [h_{2h+2l}h_{2k} + h_{2k+1+2l}h_{2k+1}]$$

Lets derive P3     3.  $\sum h_{2k+1} = \frac{1}{\sqrt{2}}$

$$1 = \sum_{k=-\infty}^{\infty} \overline{h_{2k}} \left[ \sum_{l=-\infty}^{\infty} h_{2k+2l} \right] + \sum_{k=-\infty}^{\infty} \overline{h_{2k+1}} \left[ \sum_{l=-\infty}^{\infty} h_{2k+1+2l} \right]$$

---

$$\sum_{l=-\infty}^{\infty} h_{2k+2l} = \sum_{l=-\infty}^{\infty} h_{2l} = \sum_{l=-\infty}^{\infty} h_{2k} \equiv A$$

$$\sum_{l=-\infty}^{\infty} h_{2k+2l+1} = \sum_{l=-\infty}^{\infty} h_{2l+1} = \sum_{l=-\infty}^{\infty} h_{2k+1} \equiv B$$

Lets derive P3    3.  $\sum h_{2k+1} = \frac{1}{\sqrt{2}}$

$$1 = \sum_{k=-\infty}^{\infty} \overline{h_{2k}} \sum_{k=-\infty}^{\infty} h_{2k} + \sum_{k=-\infty}^{\infty} \overline{h_{2k+1}} \sum_{k=-\infty}^{\infty} h_{2k+1}$$

$$1 = \overline{A}A + \overline{B}B = |A|^2 + |B|^2$$

Lets derive P3 3.  $\sum h_{2k+1} = \frac{1}{\sqrt{2}}$

$$\sum_k h_k = \sqrt{2}$$

$$\sqrt{2} = \sum_{k=-\infty}^{\infty} h_k = \sum_{k=-\infty}^{\infty} h_{2k} + \sum_{k=-\infty}^{\infty} h_{2k+1} = A + B$$

$$A + B = \sqrt{2}$$

Lets derive P3

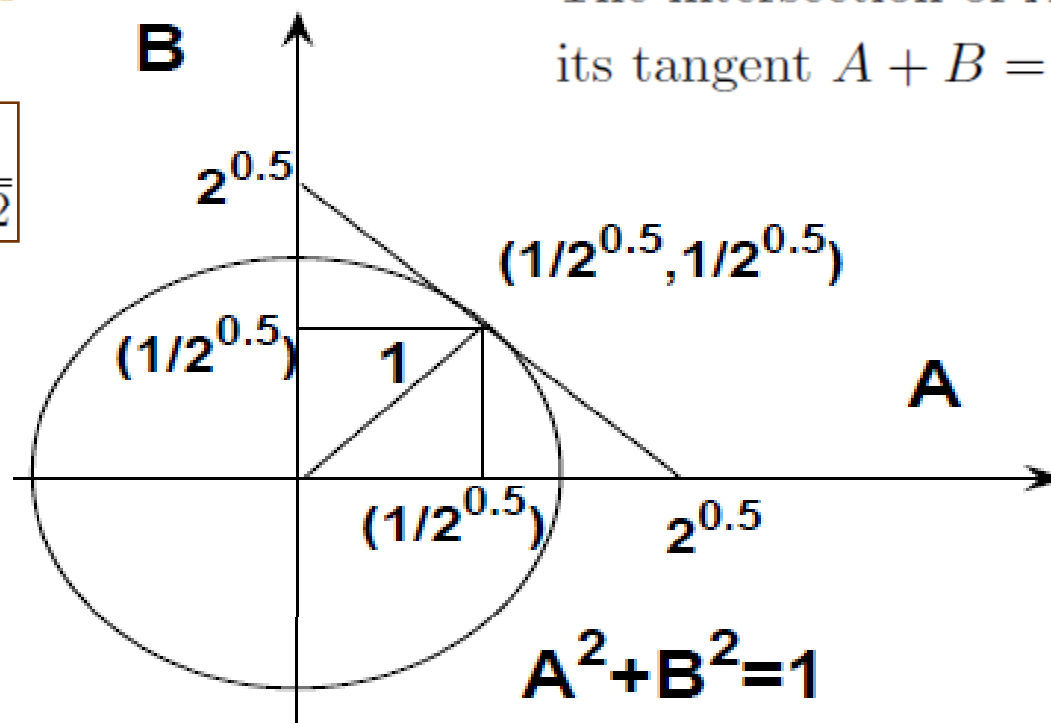
$$3. \quad \sum h_{2k+1} = \frac{1}{\sqrt{2}}$$

$$1 = \overline{A}A + \overline{B}B = |A|^2 + |B|^2$$

$$A + B = \sqrt{2}$$

The intersection of  $A^2 + B^2 = 1$  and its tangent  $A + B = \sqrt{2}$  at  $(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$ .

$$A = B = \frac{1}{\sqrt{2}}$$





Lets derive P3    3.  $\sum h_{2k+1} = \frac{1}{\sqrt{2}}$

$$A = B = \frac{1}{\sqrt{2}}$$

$$\sum_{l=-\infty}^{\infty} h_{2k+2l+1} = \sum_{l=-\infty}^{\infty} h_{2l+1} = \sum_{l=-\infty}^{\infty} h_{2k+1} \equiv B$$

$$3. \quad \sum h_{2k+1} = \frac{1}{\sqrt{2}}$$

Lets derive P2     2.  $\sum h_{2k} = \frac{1}{\sqrt{2}}$

$$A = B = \frac{1}{\sqrt{2}}$$

$$\sum_{l=-\infty}^{\infty} h_{2k+2l} = \sum_{l=-\infty}^{\infty} h_{2l} = \sum_{l=-\infty}^{\infty} h_{2k} \equiv A$$

$$2. \quad \sum h_{2k} = \frac{1}{\sqrt{2}}$$



# Verification for Haar

$$h_0 = h_1 = \frac{1}{\sqrt{2}}, h_k = 0 \text{ for } k \neq 0, 1,$$

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$$2. \quad \sum_{k=0} h_{2k} = h_0 = \frac{1}{\sqrt{2}}$$

$$3. \quad \sum_{k=1} h_{2k+1} = h_1 = \frac{1}{\sqrt{2}}$$

# Verification for Haar

$$h_0 = h_1 = \frac{1}{\sqrt{2}}, h_k = 0 \text{ for } k \neq 0, 1,$$

$$4. \quad \sum_{k=0,1} |h_k|^2 = |h_0|^2 + |h_1|^2 = \frac{1}{2} + \frac{1}{2} = 1$$

# Verification for Haar

$$h_0 = h_1 = \frac{1}{\sqrt{2}}, h_k = 0 \text{ for } k \neq 0, 1,$$

$$5. \sum_k 2h_{k-2l}h_k$$

$$\begin{aligned} \sum_{k=0,1;l=0} 2h_{k-2l}\overline{h_k} &= 2[h_0\overline{h_0} + h_1\overline{h_1}] \\ &= 2\left[\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\right] = 2\left[\frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}}\right] = 1 \end{aligned}$$

$$\sum_k 2h_{k-2l}\overline{h_k} = \delta_{l,0}$$

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**Discrete  
Orthonormality**

$$\sum_k 2h_{k-2l}\overline{h_k} = \delta_{l,0}$$



# Not all function obey!

- Roof scaling function

$$\phi(x) = \begin{cases} x, & 0 \leq x \leq 1 \\ 2 - x, & 1 \leq x \leq 2 \end{cases}$$

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- Roof scaling function

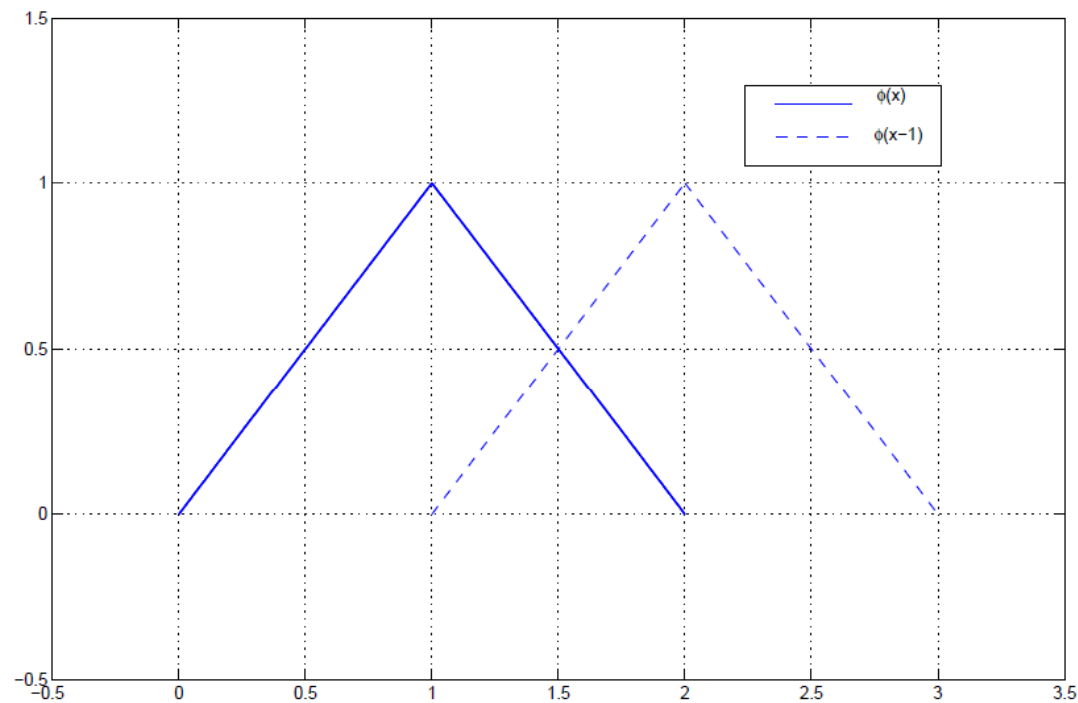
$$\phi(x) = \begin{cases} x, & 0 \leq x \leq 1 \\ 2 - x, & 1 \leq x \leq 2 \end{cases}$$

$$\int_{-\infty}^{\infty} \phi(x)\phi(x-1)dx = \int_1^2 \phi(x)\phi(x-1)dx \neq 0$$

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- Roof scaling function  $\phi(x) = \begin{cases} x, & 0 \leq x \leq 1 \\ 2 - x, & 1 \leq x \leq 2 \end{cases}$

$$\int_{-\infty}^{\infty} \phi(x)\phi(x-1)dx = \int_1^2 \phi(x)\phi(x-1)dx \neq 0$$





# Framework

- Leads us to two questions
  - 1) How do we go about selecting the mother wavelet and scale of analysis?
  - 2) What is the procedure to calculate scaling and wavelet coefficients?



# Applications

- Detecting (hidden) jumps/discontinuities
- Suppressing polynomials towards de-noising
- 2D applications: compression and pattern recognition



# Properties of Unitary Transform

- Energy **compaction**: only few transform coefficients have large magnitude
  - Such property is related to the decorrelating role of unitary transform
- Energy **conservation**: unitary transform preserves the 2-norm of input vectors
  - Such property essentially comes from the fact that rotating coordinates does not affect Euclidean distance

# Unitary Matrix and ID Unitary Transform

## Definition

A matrix  $A$  is called **unitary** if  $A^{-1} = A^{*T}$

conjugate  
transpose

When the transform matrix  $A$  is unitary, the defined transform  $\vec{y} = A\vec{x}$  is called **unitary transform**

## Example

$$\mathbf{A} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}, \mathbf{A}^{-1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} = \mathbf{A}^T$$

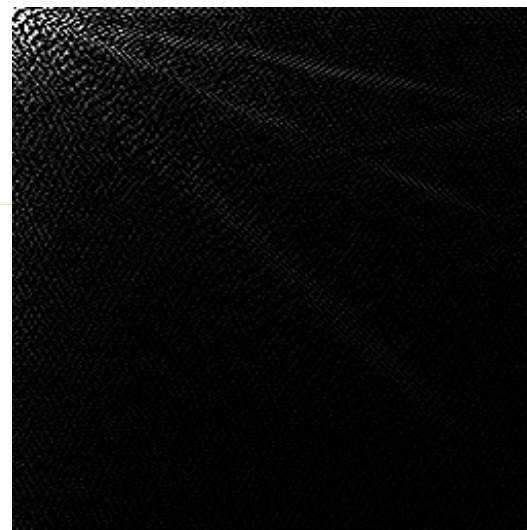
For a real matrix  $A$ , it is unitary if  $A^{-1} = A^T$

# 2D DCT

$$Y = CXC^T = CXC^{-1}$$



Original cameraman image X



Its DCT coefficients Y  
(2451 significant coefficients,  $th=64$ )

Notice the excellent energy compaction property of DCT

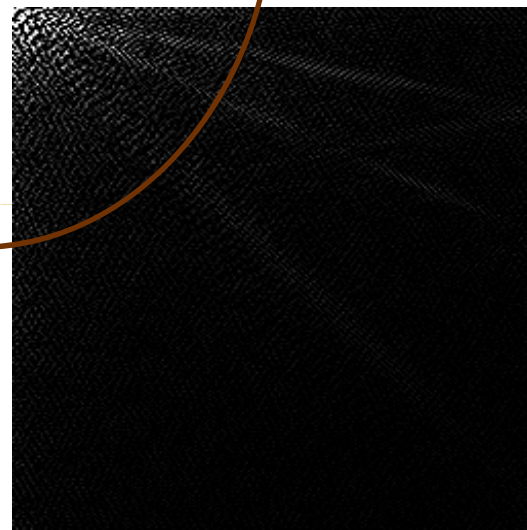


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**Thank You!**

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**Questions ??**