

SNS COLLEGE OF TECHNOLOGY



Coimbatore-35
An Autonomous Institution

Accredited by NBA – AICTE and Accredited by NAAC – UGC with 'A+' Grade Approved by AICTE, New Delhi & Affiliated to Anna University, Chennai

DEPARTMENT OF ELECTRONICS & COMMUNICATION ENGINEERING

19ECE351 – IMAGE PROCESSING AND COMPUTER VISION

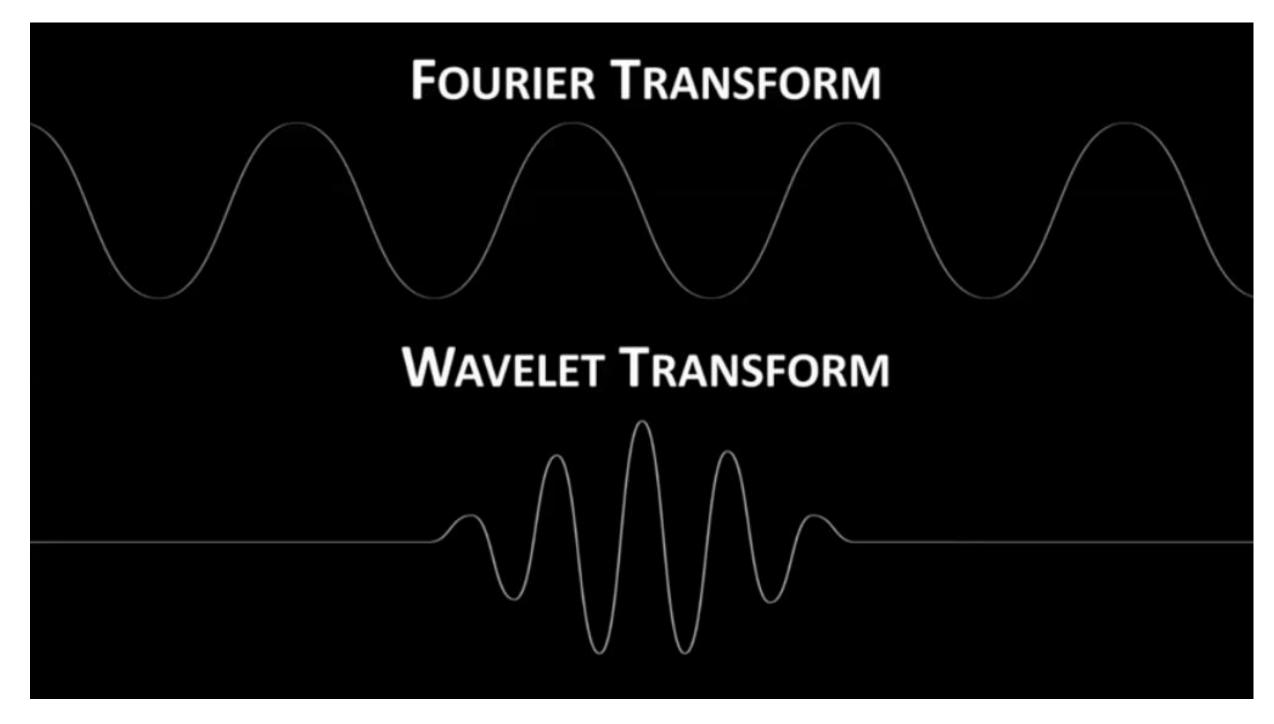
III B.E. ECE / V SEMESTER

UNIT 1 - DIGITAL IMAGE FUNDAMENTALS AND TRANSFORMS

TOPIC - WAVELET TRANSFORM

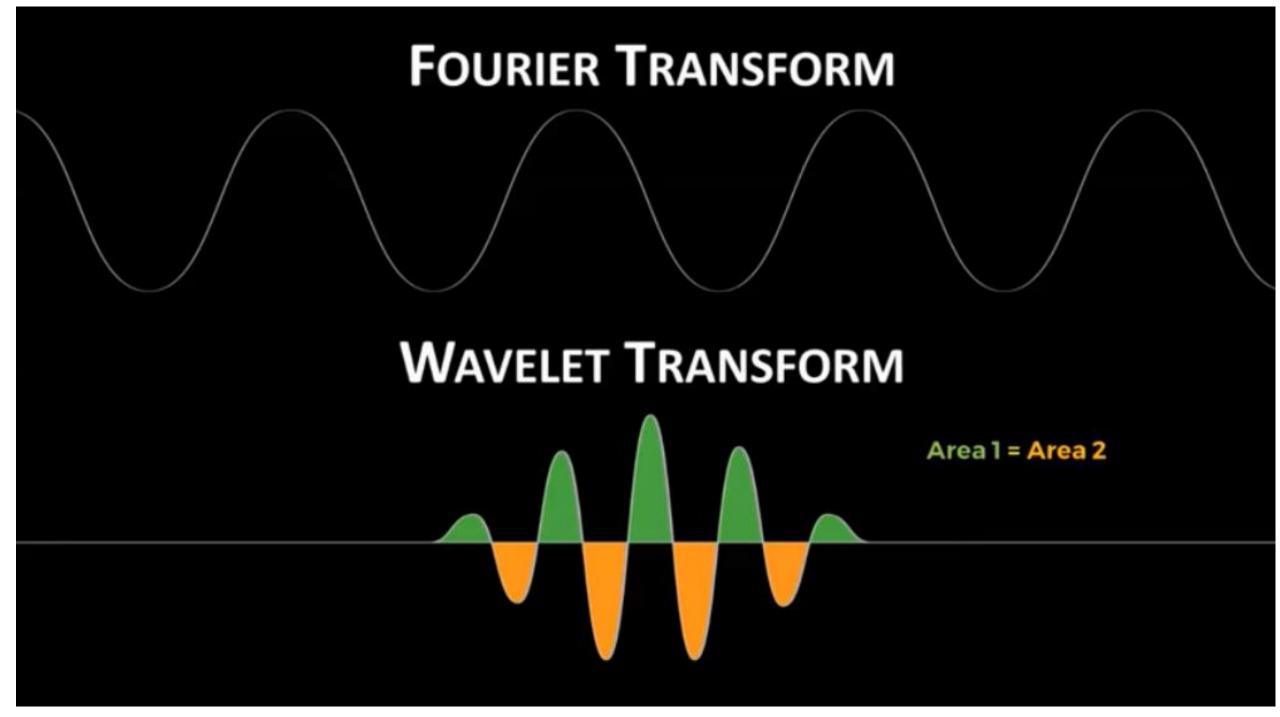






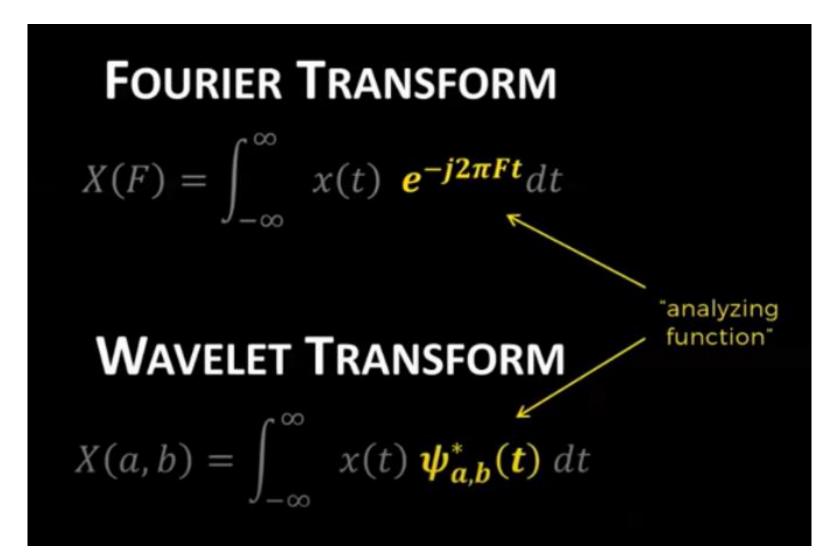


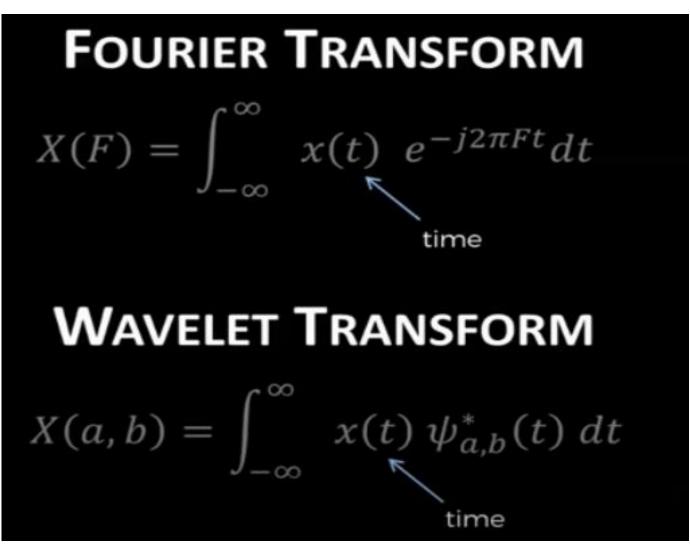








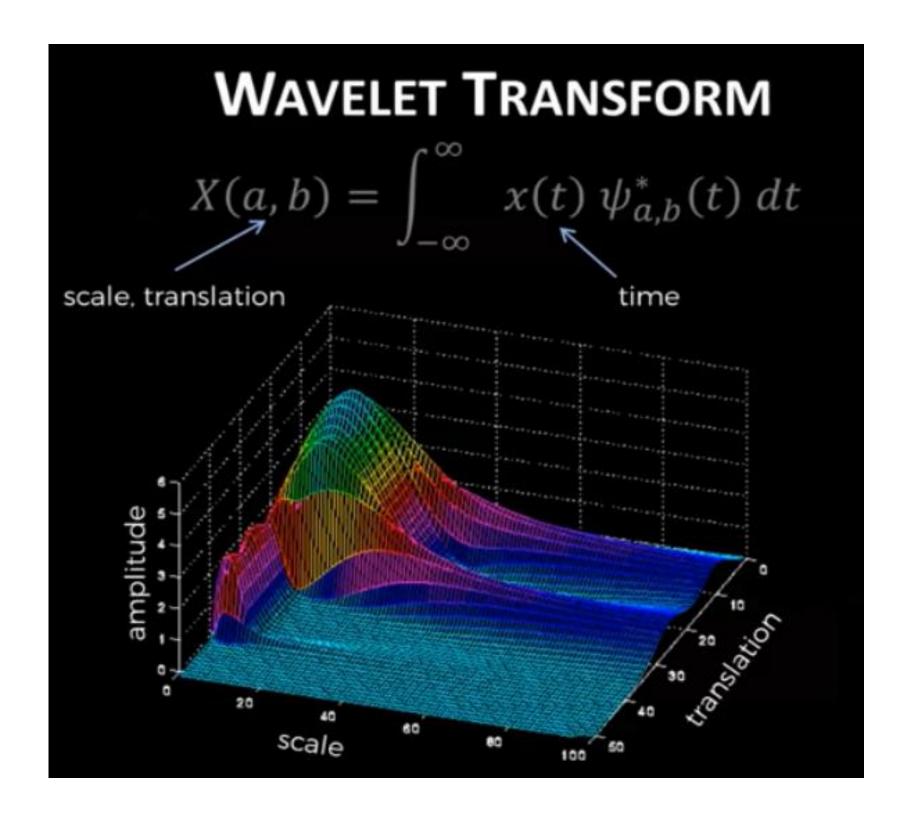






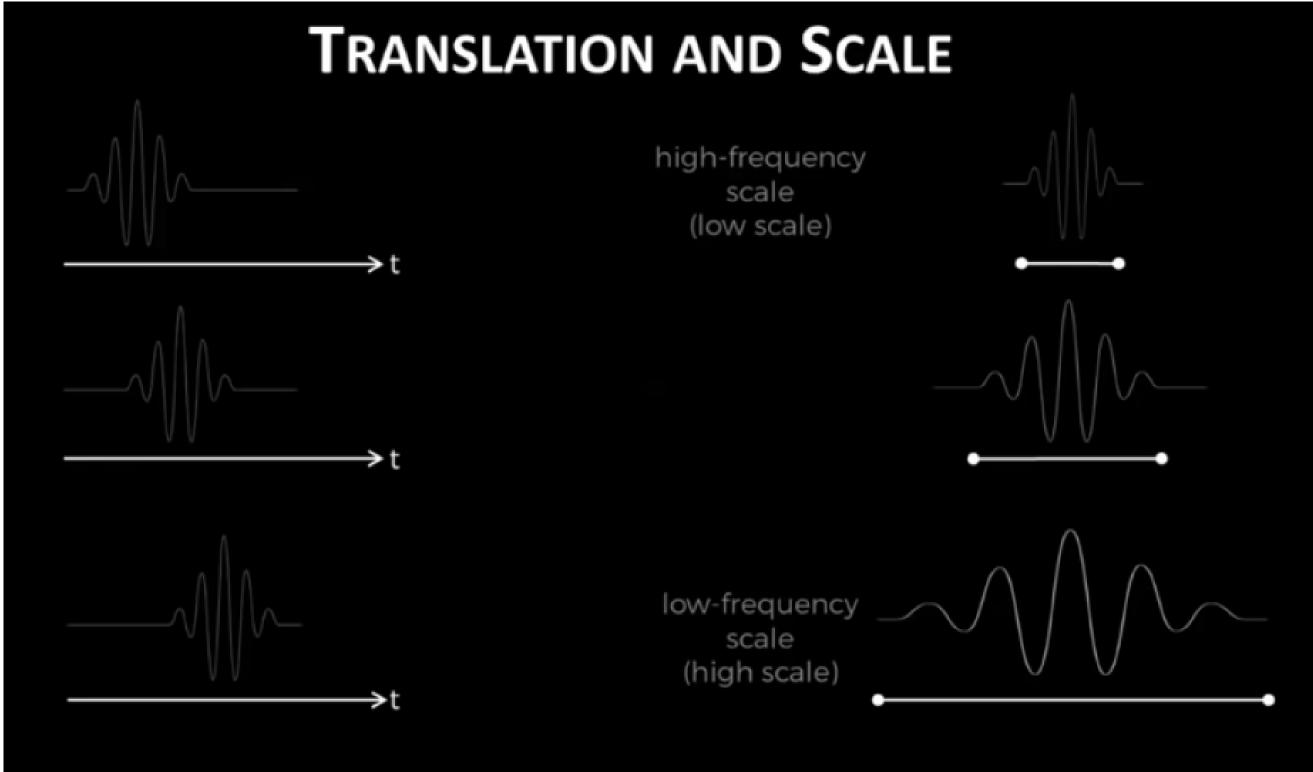






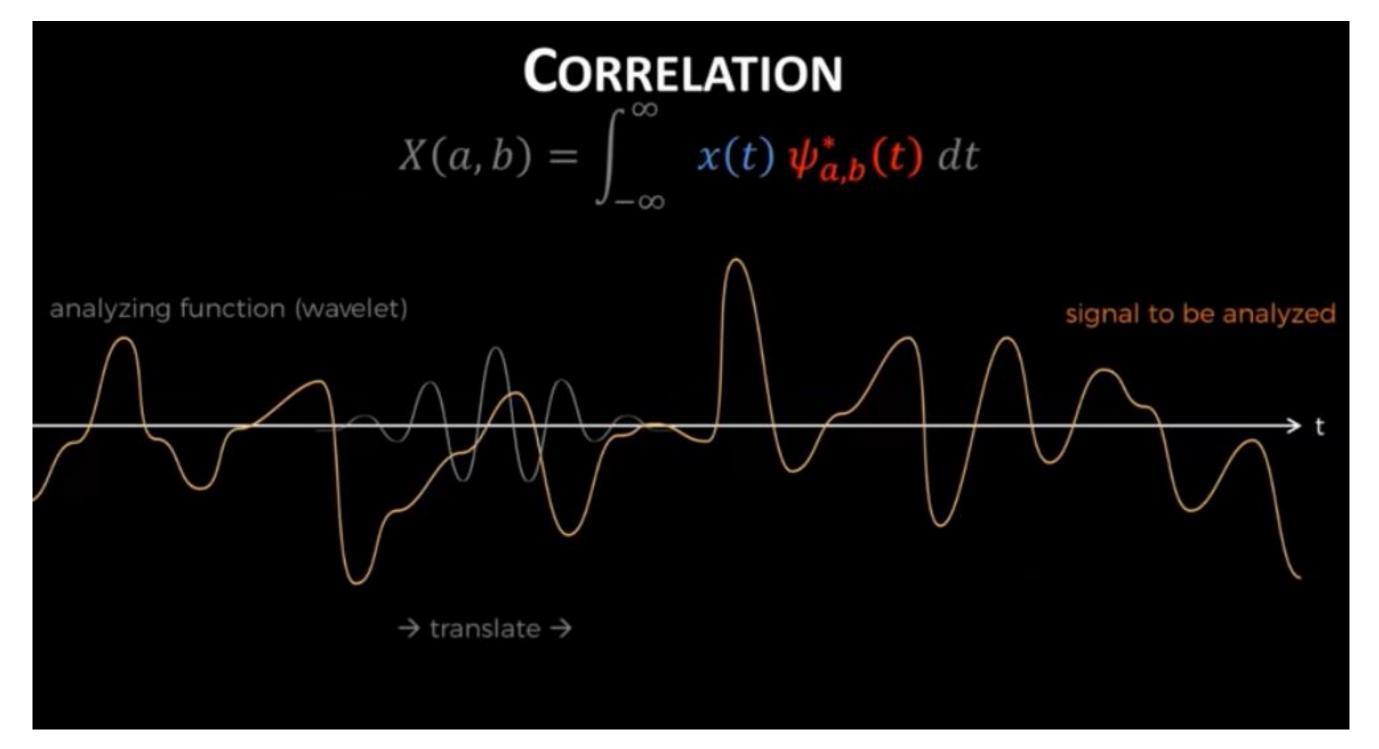










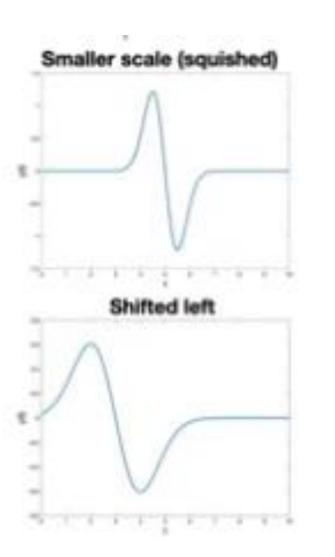


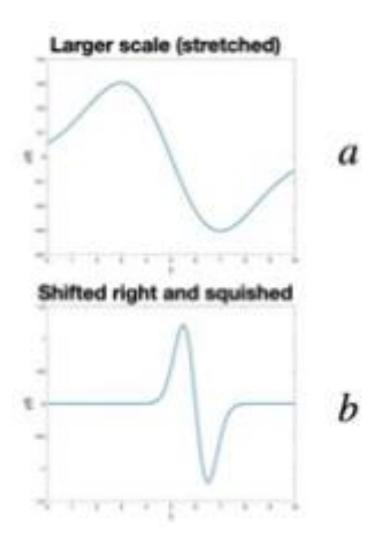




[1] Two basic properties:

- Scale (dilation) how "stretched" or "squished" wavelet is (related to frequency)
- Location where wavelet is positioned in time









Wavelet Transform

Two methods

Continuous Wavelet Transform (CWT)

"Mother" (basis) wavelets are defined everywhere

$$\psi = \psi(t)$$

Transform must be discretized for implementation

$$T(a,b) = \frac{1}{\sqrt{a}} \int_{-\infty}^{\infty} x(t) \, \psi^* \frac{(t-b)}{a} dt$$

T(a, b) may have redundant information since same feature may be captured by multiple scales

Discrete Wavelet Transform (DWT)

"Mother" (basis) wavelets are only defined on discrete grid

$$\psi = \psi_{m,n}(t)$$

Transform is already discrete!

$$T_{m,n} = \int_{-\infty}^{\infty} x(t) \, \psi_{m,n}(t) \, dt$$

Coefficients $(T_{m,n})$ do not have redundant information since discretized wavelets can be defined to be orthonormal

$$\int_{-\infty}^{\infty} \psi_{m,n}(t) \psi_{m',n'}(t) dt = \begin{cases} 1 & \text{if } m = m' \text{ and } n = n' \\ 0 & \text{otherwise.} \end{cases}$$





- Wavelets for 2-D Signals (Images)
- Wavelet Analysis and Synthesis of images using Filer Banks

Wavelets for 2-D signals

- Images are obviously two dimensional data [i.e. f(x, y)].
- To transform images, we need two dimensional wavelets [i.e. $\psi(x,y)$].





 We can apply the one dimensional transform to the rows and columns of the image successively as separable two dimensional transform.

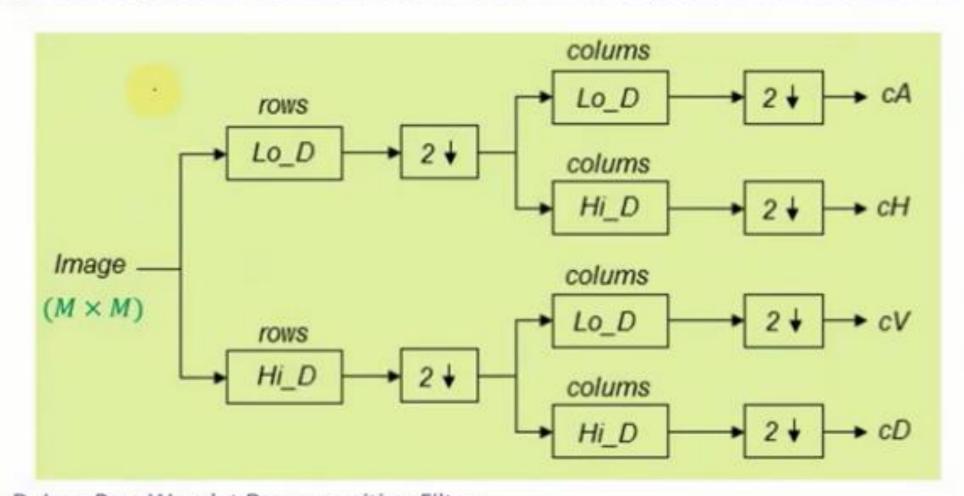
In most of the applications, where wavelets are used for image processing, this
approach is preferred because of the low computational complexity of separable
transforms.





Filter Bank Theory (Wavelet Decomposition)

Filter based approach based on row and column wise operation is shown below.



Lo_D: Low Pass Wavelet Decomposition Filter Hi_D: High Pass Wavelet Decomposition Filter 2 1: Down Sampled by 2.

Approximation Coefficients

$$\left(\frac{M}{2} \times \frac{M}{2}\right)$$

Horizontal Detailed Coefficients

$$\left(\frac{M}{2} \times \frac{M}{2}\right)$$

Vertical Detailed Coefficients

$$\left(\frac{M}{2} \times \frac{M}{2}\right)$$

Diagonal Detailed Coefficients

$$\left(\frac{M}{2} \times \frac{M}{2}\right)$$





Filter Bank Theory (Wavelet Reconstruction)

Approximation Coefficients

$$\left(\frac{M}{2} \times \frac{M}{2}\right)$$

Horizontal Detailed Coefficients

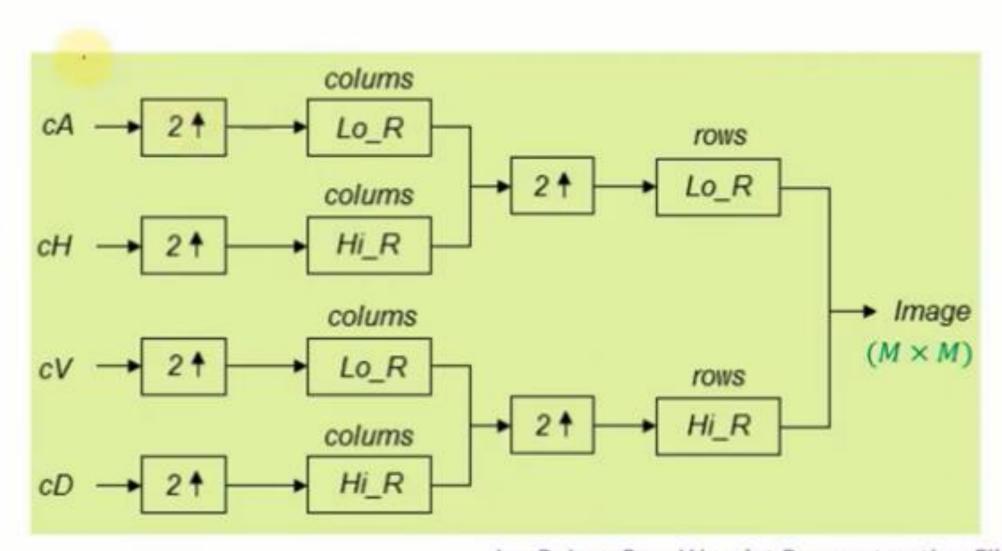
$$\left(\frac{M}{2} \times \frac{M}{2}\right)$$

Vertical Detailed Coefficients

$$\left(\frac{M}{2} \times \frac{M}{2}\right)$$

Diagonal Detailed Coefficients

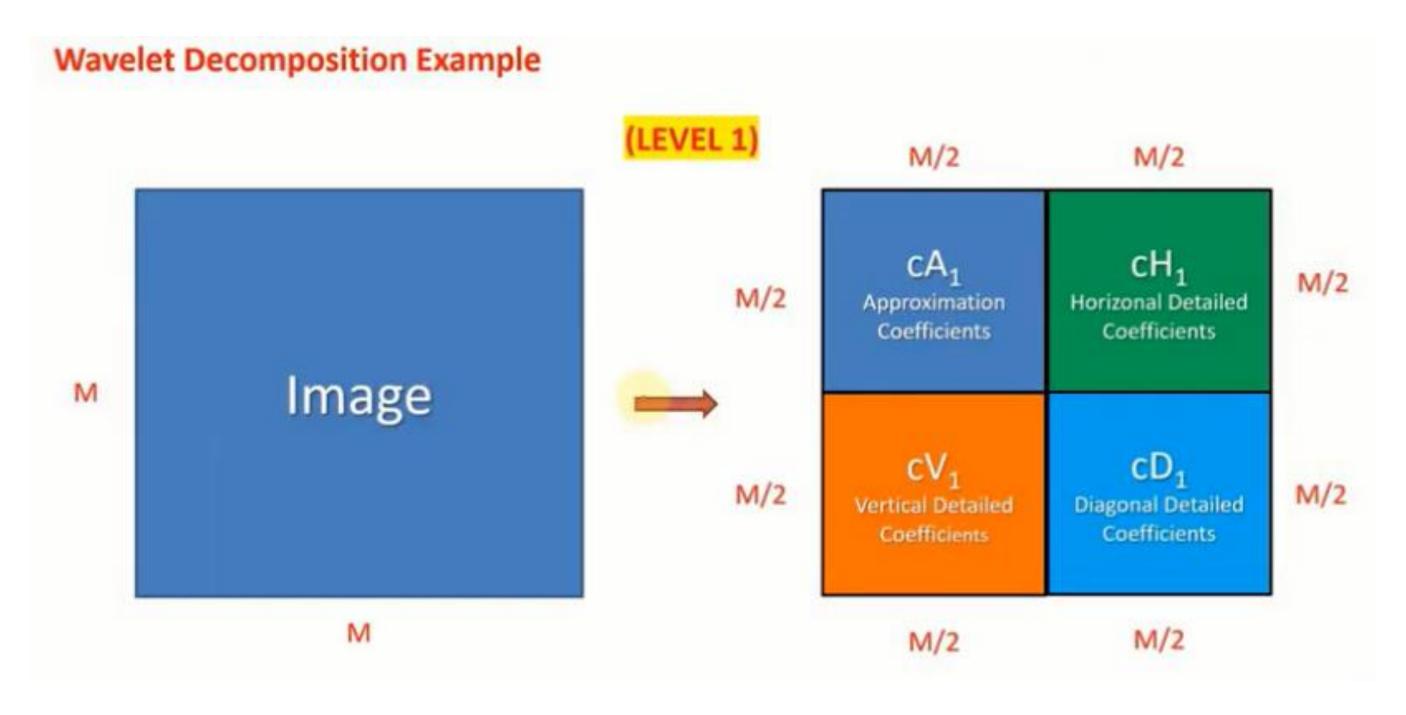
$$\left(\frac{M}{2} \times \frac{M}{2}\right)$$



Lo_R: Low Pass Wavelet Re-construction Filter Hi_R: High Pass Wavelet Re-construction Filter 2 1: Up Sampled by 2.

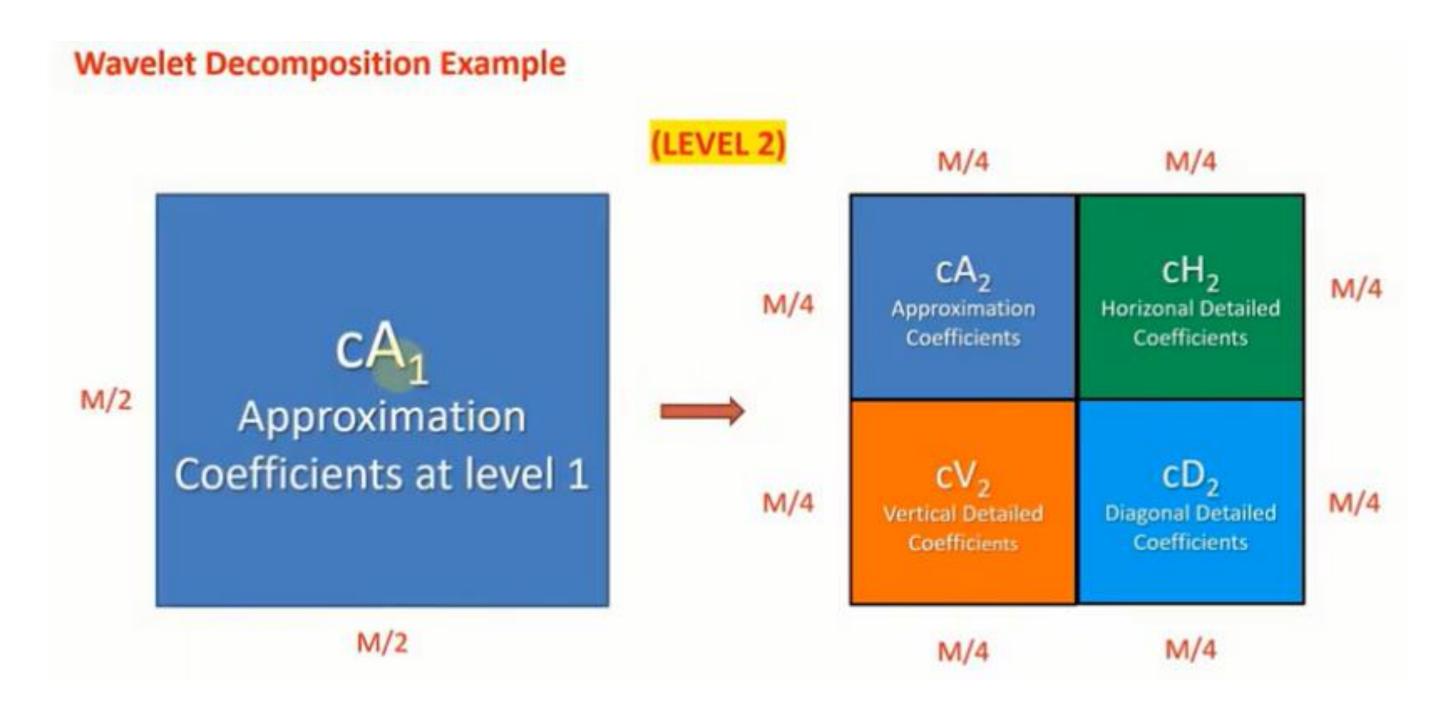






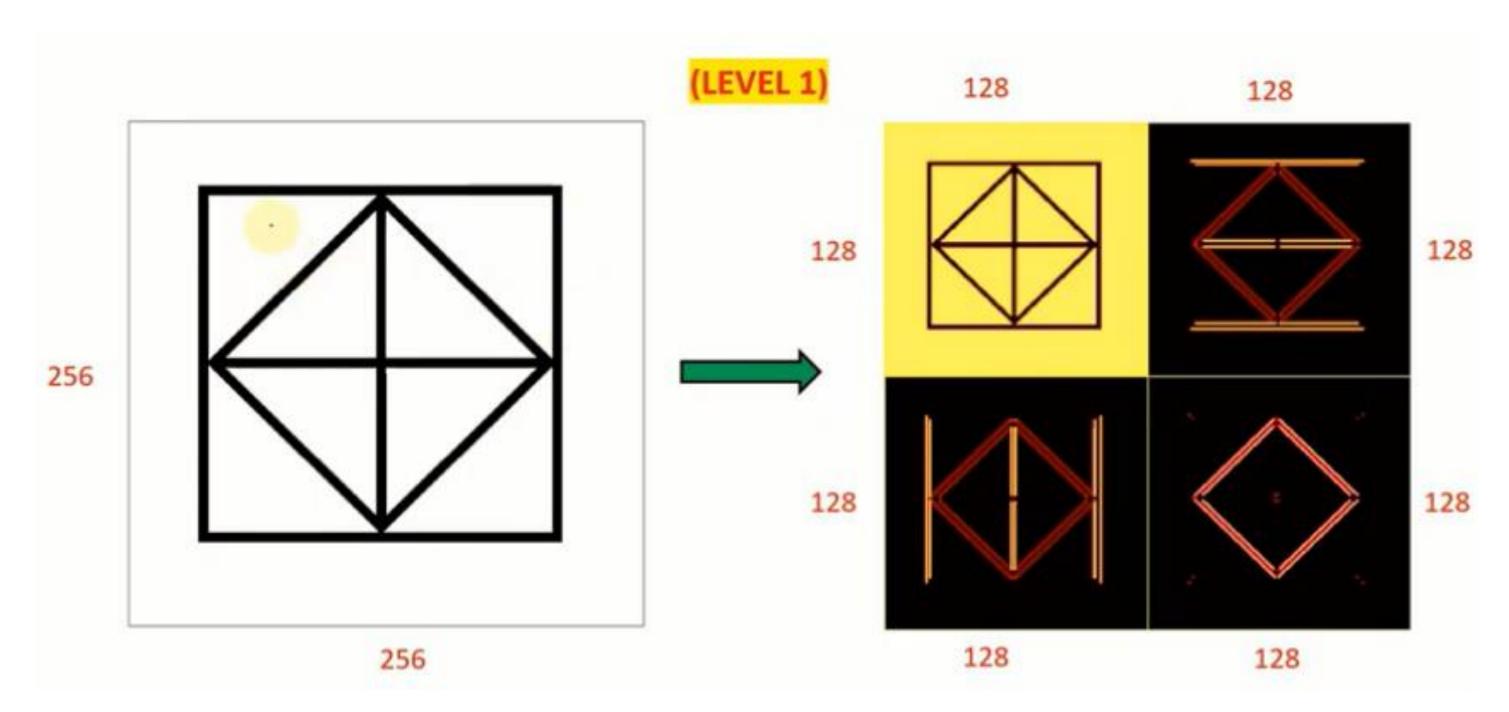






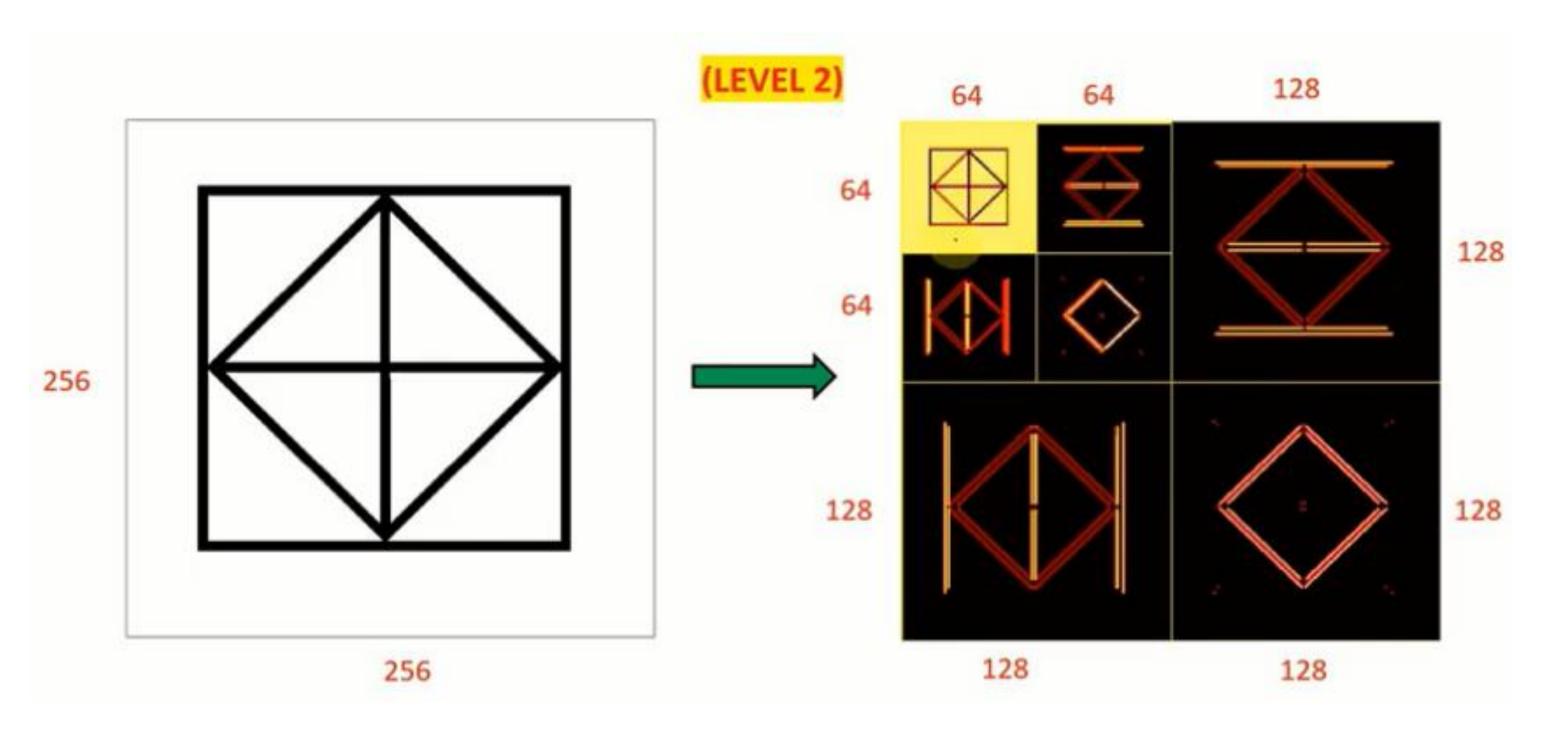
















Wavelet Decomposition Example

LEVEL - 1





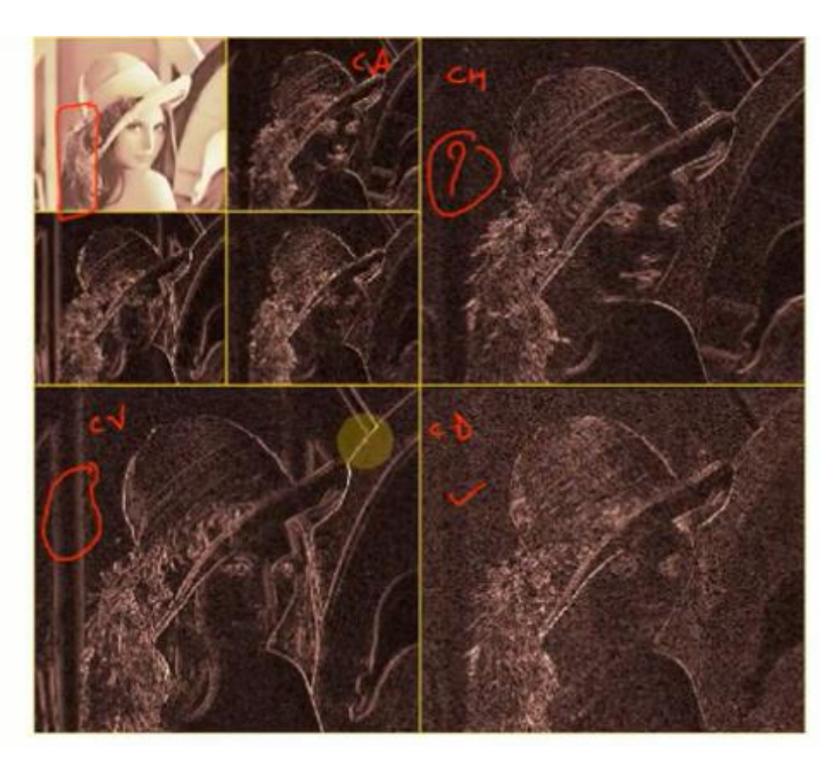




Wavelet Decomposition Example

LEVEL - 2









Wavelet Decomposition Example

LEVEL - 3



