

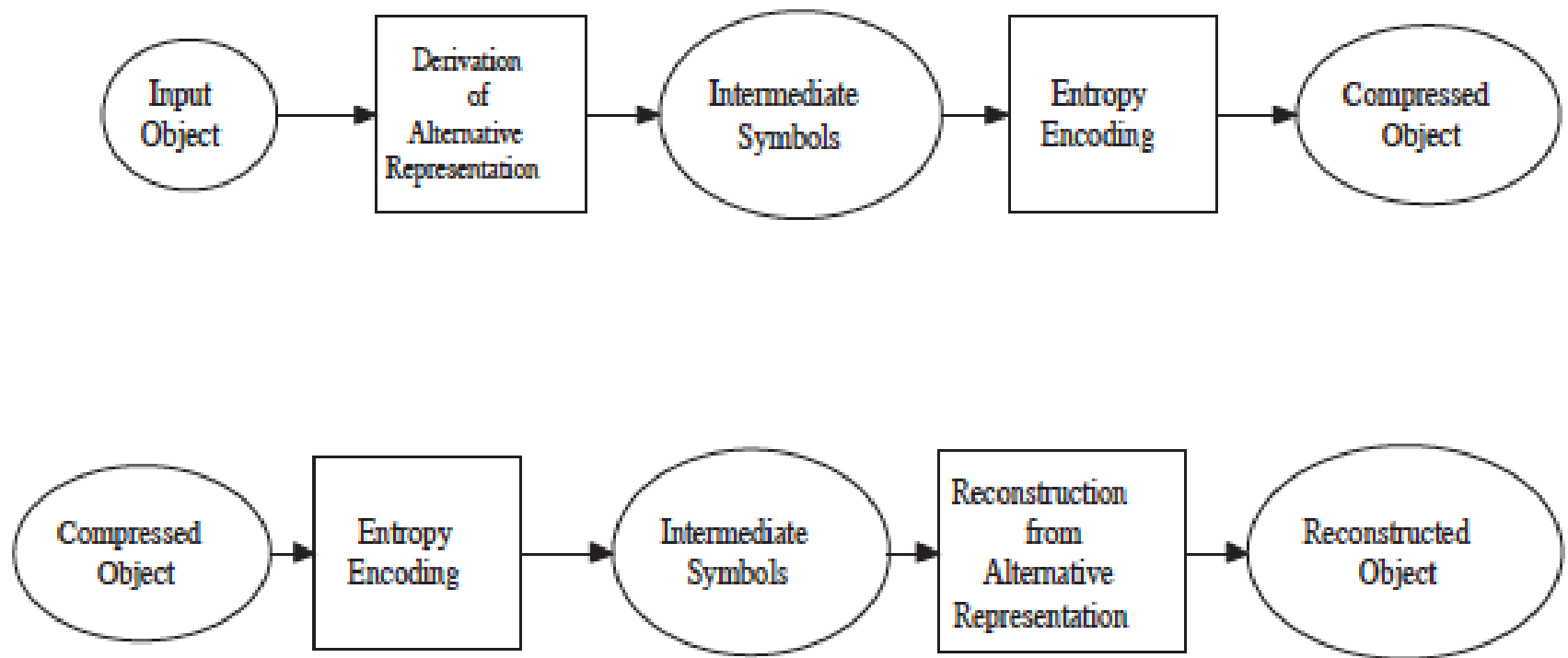


Processing in block DCT domain

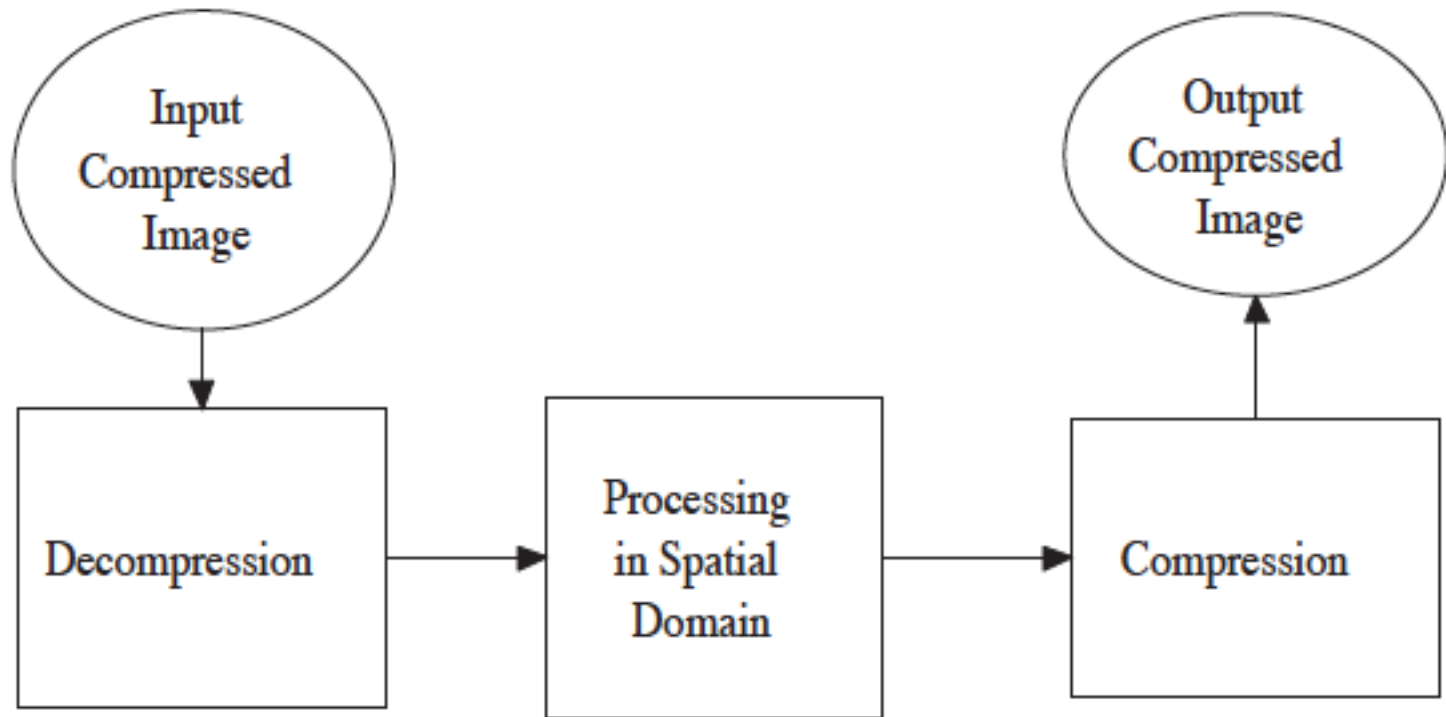
Jayanta Mukhopadhyay
Dept. of Computer Science and Engg.

“Image and video processing in the compressed domain”, CRC Press, 2011.

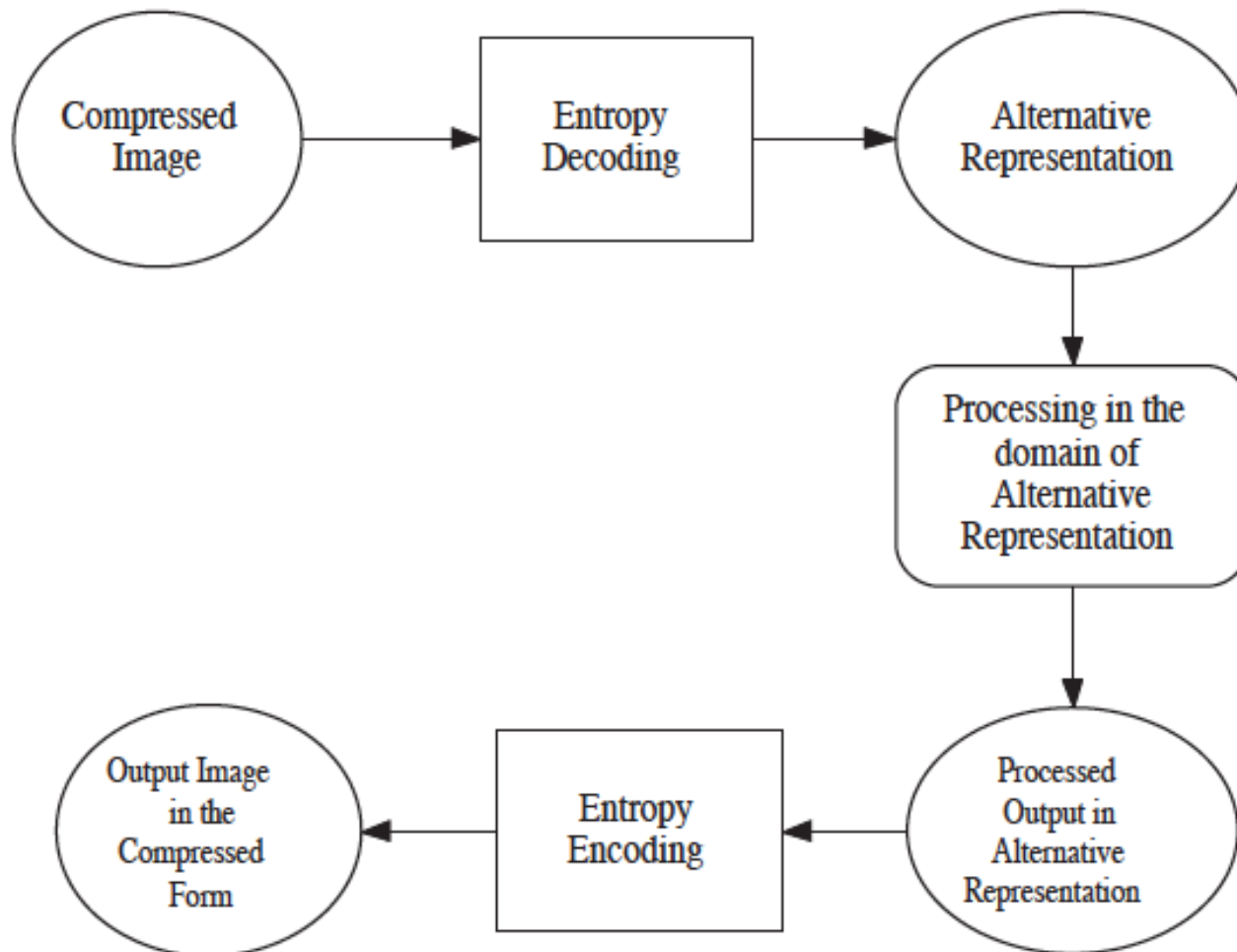
Image compression and decompression



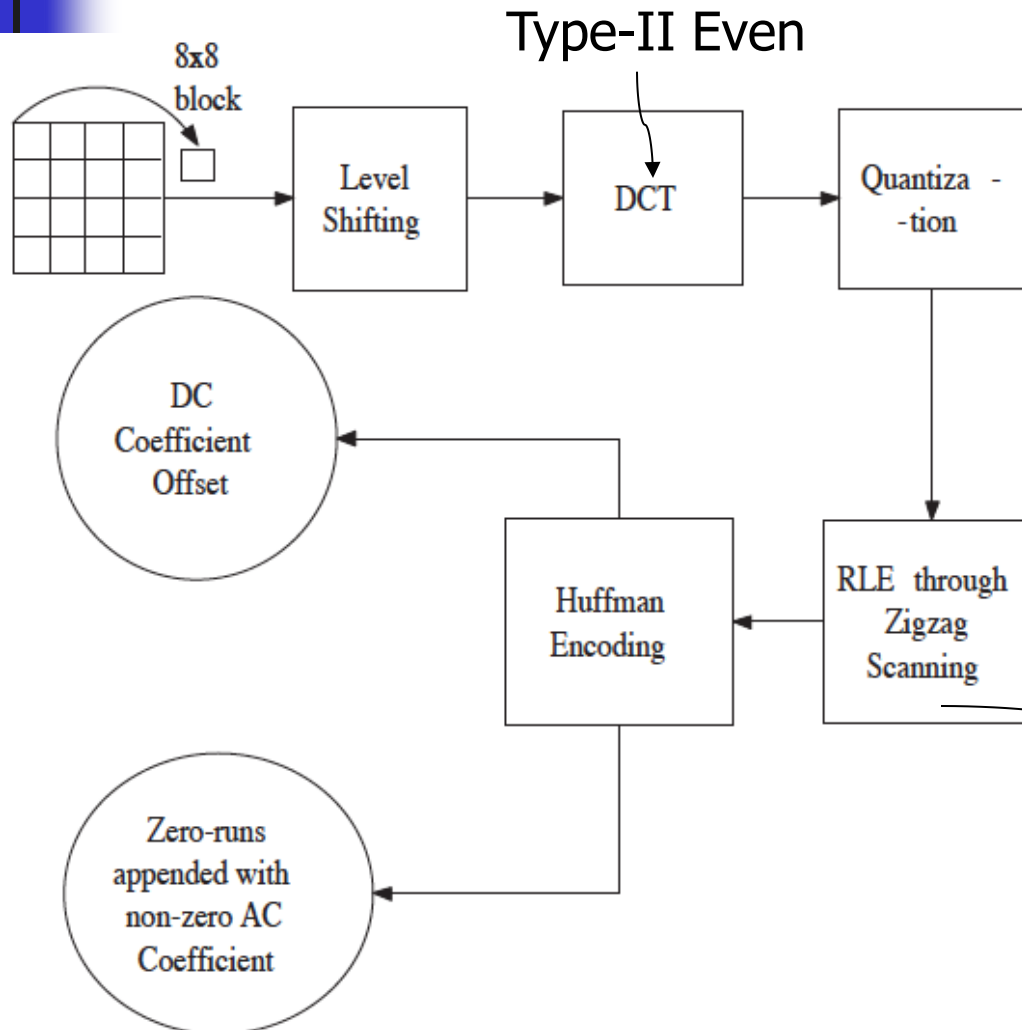
Processing with compressed image: Spatial domain approach



Processing with compressed image: Compressed domain approach

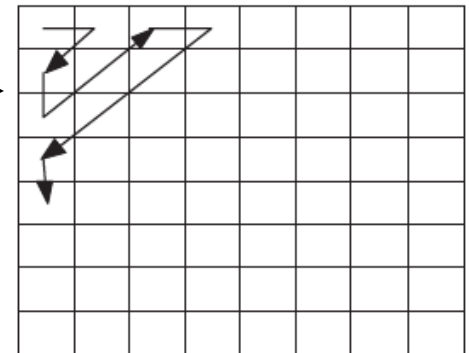


JPEG: Baseline scheme



16	11	10	16	24	40	51	61
12	12	14	19	26	58	60	55
14	13	16	24	40	57	69	56
14	17	22	29	51	87	80	62
18	22	37	56	68	109	103	77
24	35	55	64	81	104	113	92
49	64	78	87	103	121	120	101
72	92	95	98	112	100	103	99

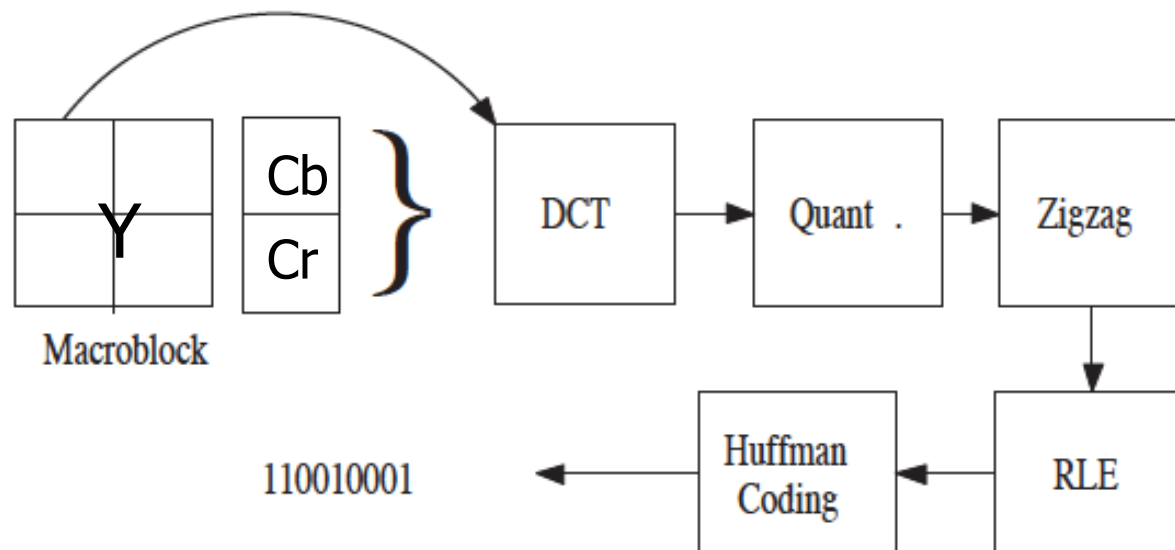
Typical quantization mask.



Color encoding in JPEG

- Y-Cb-Cr color space:

$$\begin{aligned}Y &= 0.502G + 0.098B + 0.256R, \\Cb &= -0.290G + 0.438B - 0.148R + 128, \\Cr &= -0.366G - 0.071B + 0.438R + 128.\end{aligned}$$





Motivations

- Computation with reduced storage.
- Avoid overhead of entropy decoding and encoding.
- Exploit spectral factorization for improving the quality of result and speed of computation.



Typical Applications

- Resizing.
- Filtering.
- Enhancement.

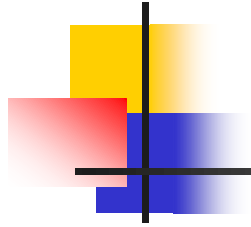


Image Resizing

Image Halving

- Use of linear and distributive properties.

$$D_{4 \times 8} = \begin{bmatrix} 0.5 & 0.5 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0.5 & 0.5 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0.5 & 0.5 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 0.5 \end{bmatrix} \mathbf{x}_d = \sum_{j=0}^1 \sum_{i=0}^1 p_i \mathbf{x}_{ij} p_j^T,$$

\mathbf{x}_{00}	\mathbf{x}_{01}
\mathbf{x}_{10}	\mathbf{x}_{11}

\Rightarrow

\mathbf{x}_d

$p_0 = \begin{bmatrix} D_{4 \times 8} \\ 0_{4 \times 8} \end{bmatrix} \quad p_1 = \begin{bmatrix} 0_{4 \times 8} \\ D_{4 \times 8} \end{bmatrix}$

$$DCT(\mathbf{x}_d) = \sum_{j=0}^1 \sum_{i=0}^1 DCT(p_i) DCT(\mathbf{x}_{ij}) DCT(p_j^T).$$

Not so sparse matrix multiplication!

DCT(p_0): Not so sparse.

No gain!

$$\begin{bmatrix} 0.5 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0.453 & 0.204 & -0.034 & 0.010 & 0 & -0.006 & 0.014 & -0.041 \\ 0 & 0.49 & 0 & 0 & 0 & 0 & 0 & -0.098 \\ -0.159 & 0.388 & 0.237 & -0.041 & 0 & 0.027 & -0.098 & -0.077 \\ 0 & 0 & 0.462 & 0 & 0 & 0 & -0.191 & 0 \\ 0.106 & -0.173 & 0.355 & 0.204 & 0 & -0.136 & -0.147 & 0.034 \\ 0 & 0 & 0 & 0.416 & 0 & -0.278 & 0 & 0 \\ -0.090 & 0.136 & -0.173 & 0.360 & 0 & -0.240 & 0.072 & -0.027 \end{bmatrix},$$

$$\begin{bmatrix} 0.500 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -0.453 & 0.204 & 0.034 & 0.010 & 0 & -0.006 & -0.014 & -0.041 \\ 0 & -0.49 & 0 & 0 & 0 & 0 & 0 & 0.098 \\ 0.159 & 0.388 & -0.237 & -0.041 & 0 & 0.027 & 0.098 & -0.077 \\ 0 & 0 & 0.462 & 0 & 0 & 0 & -0.191 & 0 \\ -0.106 & -0.173 & -0.355 & 0.204 & 0 & -0.136 & 0.147 & 0.034 \\ 0 & 0 & 0 & -0.416 & 0 & 0.278 & 0 & 0 \\ 0.090 & 0.136 & 0.173 & 0.360 & 0 & -0.240 & -0.072 & -0.027 \end{bmatrix}$$

DCT(p_1)



Typical result:



Original



Bi-linear



Linear and
distributive
method



2D DCT: Sub-band relation

$$x_{LL}(m, n) = \frac{1}{4} \{x(2m, 2n) + x(2m + 1, 2n) \\ + x(2m, 2n + 1) + x(2m + 1, 2n + 1)\}, \quad 0 \leq m, n \leq \frac{N}{2} - 1.$$

Low-pass truncated approximation:

$$X(k, l) = \begin{cases} 2\overline{X_{LL}}(k, l), & k, l = 0, 1, \dots, \frac{N}{2} - 1 \\ 0, & \text{otherwise.} \end{cases}$$

Block composition and decomposition

$$X^{(N)} = C_N x$$

- To convert M adjacent N-point DCT blocks to a single MxN-point DCT block.

$$X^{(MN)} = A_{(M,N)} [X_0^{(N)} X_1^{(N)} \dots X_{M-1}^{(N)}]^T,$$

$$A_{(M,N)} = C_{MN} \begin{bmatrix} C_N^{-1} & 0_N & 0_N & \dots & 0_N & 0_N \\ 0_N & C_N^{-1} & 0_N & \dots & 0_N & 0_N \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0_N & 0_N & 0_N & \dots & C_N^{-1} & 0_N \\ 0_N & 0_N & 0_N & \dots & 0_N & C_N^{-1} \end{bmatrix},$$

NxN zero matrix

2D DCT: Block composition and decomposition

$$X^{(LN \times MN)} = A_{(L,N)} \begin{bmatrix} X_{0,0}^{(N \times N)} & X_{0,1}^{(N \times N)} & \cdots & X_{0,M-1}^{(N \times N)} \\ X_{1,0}^{(N \times N)} & X_{1,1}^{(N \times N)} & \cdots & X_{1,M-1}^{(N \times N)} \\ \vdots & \vdots & \ddots & \vdots \\ X_{L-1,0}^{(N \times N)} & X_{L-1,1}^{(N \times N)} & \cdots & X_{L-1,M-1}^{(N \times N)} \end{bmatrix} A_{(M,N)}^T$$

$$\begin{bmatrix} X_{0,0}^{(N \times N)} & X_{0,1}^{(N \times N)} & \cdots & X_{0,M-1}^{(N \times N)} \\ X_{1,0}^{(N \times N)} & X_{1,1}^{(N \times N)} & \cdots & X_{1,M-1}^{(N \times N)} \\ \vdots & \vdots & \ddots & \vdots \\ X_{L-1,0}^{(N \times N)} & X_{L-1,1}^{(N \times N)} & \cdots & X_{L-1,M-1}^{(N \times N)} \end{bmatrix} = A_{(L,N)}^{-1} X^{(LN \times MN)} A_{(M,N)}^{-1T}$$

Useful conversion for halving or doubling 8-point DCT blocks.

$$A_{(2,4)} = C_8 \cdot \begin{bmatrix} C_4^{-1} & 0_4 \\ 0_4 & C_4^{-1} \end{bmatrix}$$

$$= \begin{bmatrix} 0.7071 & 0 & 0 & 0 & 0.7071 & 0 & 0 & 0 \\ 0.6407 & 0.294 & -0.0528 & 0.0162 & -0.6407 & 0.294 & 0.0528 & 0.0162 \\ 0 & 0.7071 & 0 & 0 & 0 & 0.7071 & 0 & 0 \\ -0.225 & 0.5594 & 0.3629 & -0.0690 & 0.225 & 0.5594 & -0.3629 & -0.069 \\ 0 & 0 & 0.7071 & 0 & 0 & 0 & 0.7071 & 0 \\ 0.1503 & -0.2492 & 0.5432 & 0.3468 & -0.1503 & -0.2492 & -0.5432 & 0.3468 \\ 0 & 0 & 0 & 0.7071 & 0 & 0 & 0 & -0.7071 \\ -0.1274 & 0.1964 & -0.2654 & 0.6122 & 0.1274 & 0.1964 & 0.2654 & 0.6122 \end{bmatrix}$$

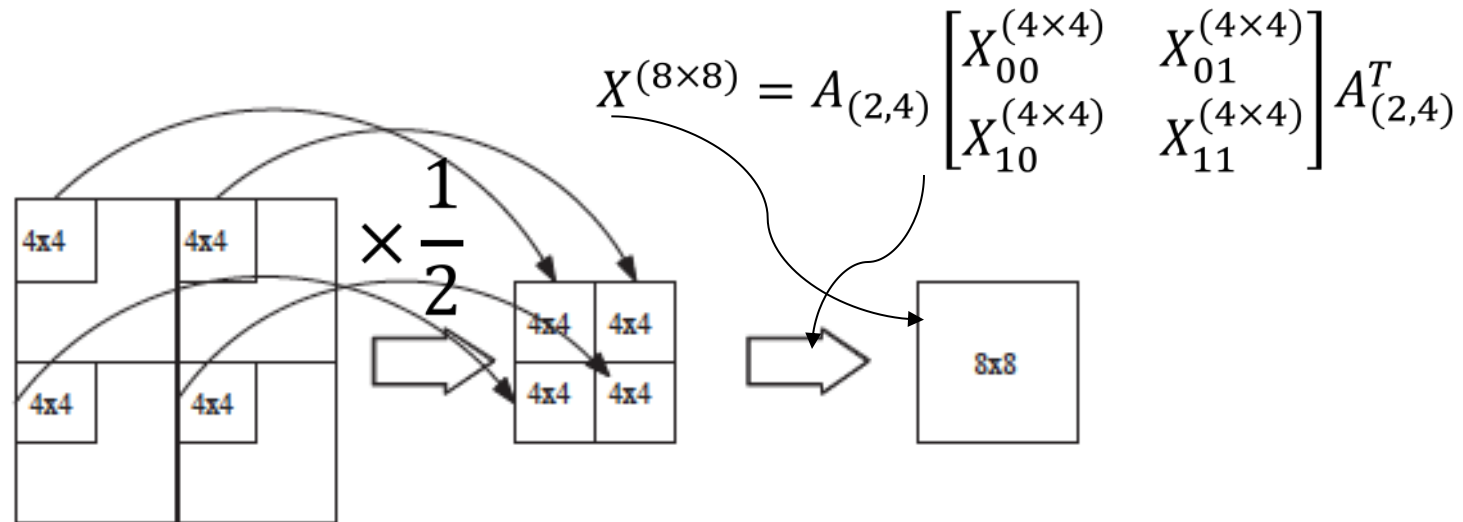
Composition

$$X^{(8 \times 8)} = A_{(2,4)} \begin{bmatrix} X_{00}^{(4 \times 4)} & X_{01}^{(4 \times 4)} \\ X_{10}^{(4 \times 4)} & X_{11}^{(4 \times 4)} \end{bmatrix} A_{(2,4)}^T$$

Decomposition

$$\begin{bmatrix} X_{00}^{(4 \times 4)} & X_{01}^{(4 \times 4)} \\ X_{10}^{(4 \times 4)} & X_{11}^{(4 \times 4)} \end{bmatrix} = A_{(2,4)}^{-1} X^{(8 \times 8)} A_{(2,4)}^{-T}$$

Image Halving: Approximation followed by Composition (IHAC)

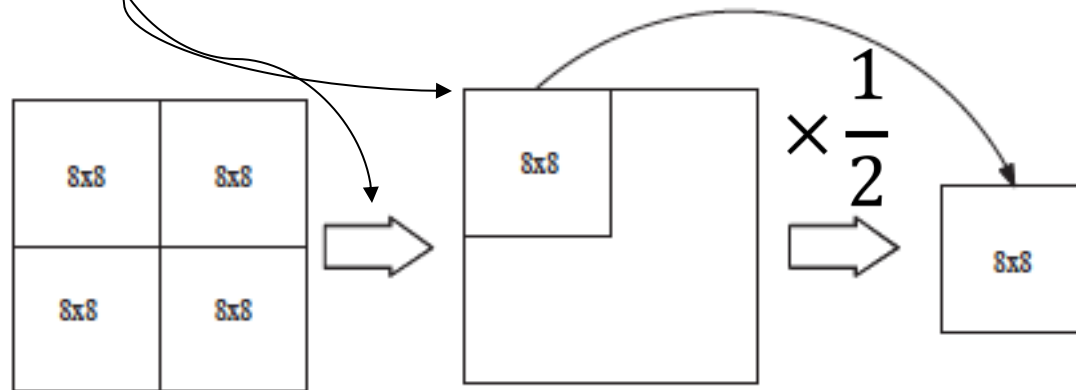


$$X_{ij}^d = \frac{1}{2} [X_{ij}(k, l)]_{0 \leq k, l \leq 3}.$$

$$Y = A_{(2,4)} \begin{bmatrix} X_{00}^d & X_{01}^d \\ X_{10}^d & X_{11}^d \end{bmatrix} A_{(2,4)}^T$$

Image Halving: Composition followed by Approximation (IHAC)

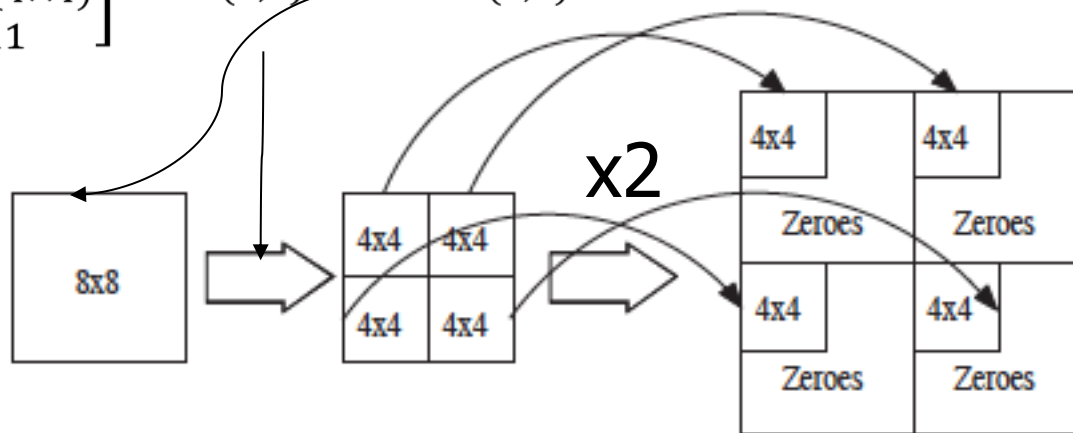
$$X^{(16 \times 16)} = A_{(2,4)} \begin{bmatrix} X_{00}^{(8 \times 8)} & X_{01}^{(8 \times 8)} \\ X_{10}^{(8 \times 8)} & X_{11}^{(8 \times 8)} \end{bmatrix} A_{(2,4)}^T$$



$$Z = A_{(2,8)} \begin{bmatrix} X_{00} & X_{01} \\ X_{10} & X_{11} \end{bmatrix} A_{(2,8)}^T, \quad Y = \frac{1}{2} [Z(k,l)] \quad 0 \leq k, l \leq 7.$$

Image Doubling: Decomposition followed by Approximation (IDDA)

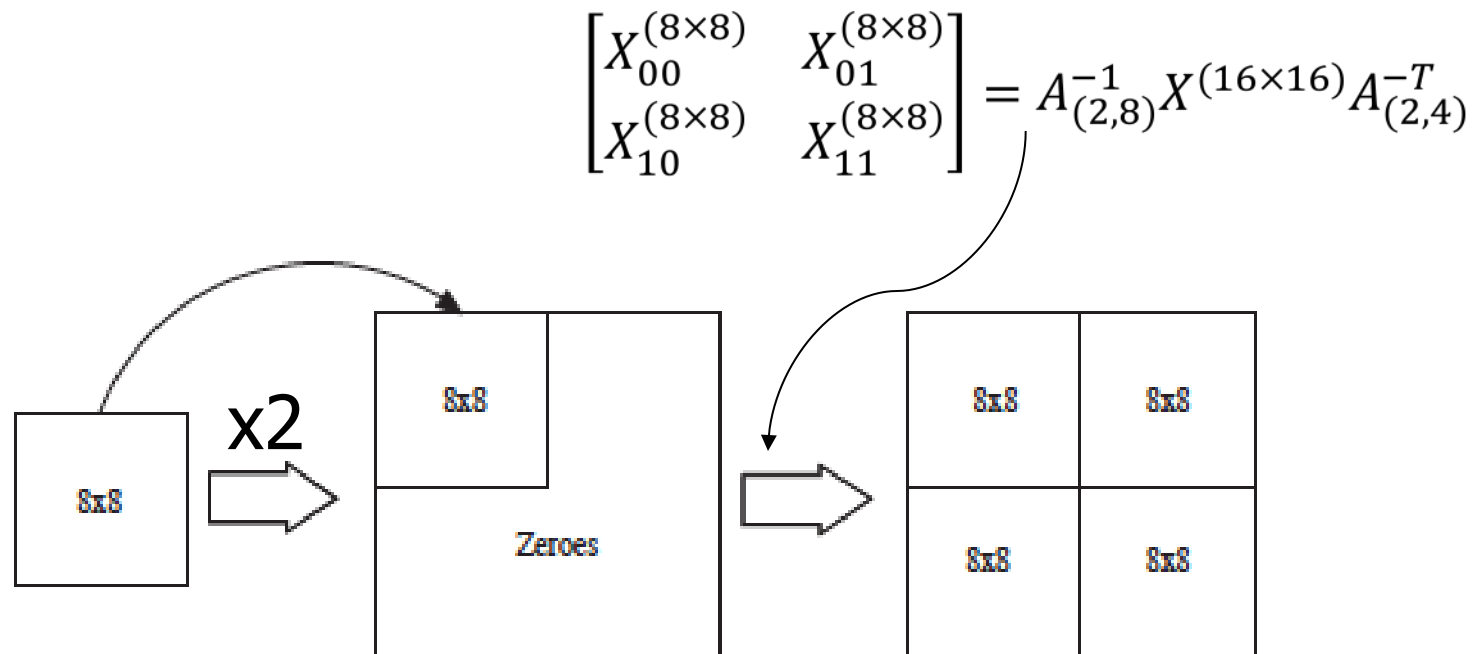
$$\begin{bmatrix} X_{00}^{(4 \times 4)} & X_{01}^{(4 \times 4)} \\ X_{10}^{(4 \times 4)} & X_{11}^{(4 \times 4)} \end{bmatrix} = A_{(2,4)}^{-1} X^{(8 \times 8)} A_{(2,4)}^{-T}$$

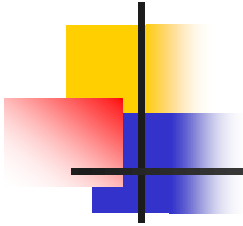


$$\begin{bmatrix} Y_{00}^d & Y_{01}^d \\ Y_{10}^d & Y_{11}^d \end{bmatrix} = A_{(2,4)}^T X A_{(2,4)}$$

$$Y_{00} = 2 \begin{bmatrix} Y_{00}^d & 0_4 \\ 0_4 & 0_4 \end{bmatrix}.$$

Image Doubling: Approximation followed by Decomposition (IDAD)







IDDA



IDAD





Resizing with integral factors

To convert $N \times N$ block to $LN \times MN$ block.

$LN \times MN$ block

$N \times N$ DCT block

$$X(k, l) = \begin{cases} \sqrt{LM} X_{LL}(k, l), & k, l = 0, 1, \dots, N-1, \\ 0, & \text{otherwise.} \end{cases}$$

$L \times M$ D/S (LMDS) 1. Merge $L \times M$ adjacent DCT blocks.

$$Z = A_{(L,N)} \begin{bmatrix} X_{0,0} & X_{0,1} & \cdots & X_{0,(M-1)} \\ X_{1,0} & X_{1,1} & \cdots & X_{1,(M-1)} \\ \vdots & \vdots & \cdots & \vdots \\ X_{(L-1),0} & X_{(L-1),1} & \cdots & X_{(L-1),(M-1)} \end{bmatrix} A_{(M,N)}^T$$

2. Sub-band approximation to a $N \times N$ DCT block.

$$Y = \sqrt{\frac{1}{LM}} [Z(k, l)]_{0 \leq k, l \leq N-1}$$



LxM U/S (LMUS)

1. Convert NxN to LNxMN block

$$\hat{B} = \begin{bmatrix} \sqrt{LM}B & 0_{(N,(M-1)N)} \\ 0_{((L-1)N,N)} & 0_{((L-1)N,(M-1)N)} \end{bmatrix}$$

Efficiently
compute
exploiting large
blocks of
zeroes.

2. Decompose into LxM NxN blocks.

An example:
3x2 D/S
and U/S





Arbitrary Resizing ($P/Q \times R/S$)

- U/S-D/S Resizing Algorithm (UDRA)
 - U/S by $P \times R$
 - D/S by $Q \times S$
- D/S-U/S Resizing Algorithm (DURA)
 - D/S by $Q \times S$
 - U/S $P \times R$

HDTV (1080x920) to NTSC (480x640)



UDRA



DURA





Hybrid Resizing (HRA)



More general sub-band relation

X: DCT block of $QN \times SN$

Y: DCT block of $PN \times RN$

Truncated DCT block of X or padded with zeroes, if required.

$$\begin{aligned}
 Y &= \sqrt{\frac{PR}{QS}} [X]_{0 \leq i < PN, 0 \leq j < RN}, & P \leq Q, R \leq S, \\
 &= \sqrt{\frac{PR}{QS}} \begin{bmatrix} [X]_{0 \leq i < QN, 0 \leq j < RN} \\ 0_{(P-Q)N \times RN} \end{bmatrix}, & P > Q, R \leq S, \\
 &= \sqrt{\frac{PR}{QS}} \begin{bmatrix} [X]_{0 \leq i < PN, 0 \leq j < SN} & 0_{PN \times (R-S)N} \end{bmatrix}, & P \leq Q, R > S, \\
 &= \sqrt{\frac{PR}{QS}} \begin{bmatrix} [X]_{0 \leq i < QN, 0 \leq j < SN} & 0_{QN \times (R-S)N} \\ 0_{(P-Q)N \times SN} & 0_{(P-Q)N \times (R-S)N} \end{bmatrix}, & P > Q, R > S.
 \end{aligned}$$

$$Y = \sqrt{\frac{PR}{QS}} \hat{X}$$

1. Merge $Q \times S$ input DCT blocks to form X.
2. Take $P \times R$ blocks to form Y with the sub-band approximation.
3. Decompose Y into $P \times R$ $N \times N$ blocks.

Hybrid Resizing: A few examples



$$\frac{2}{3} \times \frac{4}{5}$$



$$\frac{2}{3} \times \frac{3}{2}$$



$$\frac{3}{2} \times \frac{5}{4}$$



$$\frac{3}{2} \times \frac{2}{3}$$



DCT Domain Filtering



Filtering a finite sequence

Input: $x(n)$, $n=0, 1, \dots, N-1$ ↗ Pad with zeroes at
 Filter: $h(n)$: $n=0, 1, 2, \dots, L-1$ ↖ undefined points.

Linear convolution

$$y(n) = \sum_{m=0}^{N-1} x(m)h(n-m), \quad n = 0, 1, \dots, (N+L-1)$$

$$\begin{bmatrix} y(0) \\ y(1) \\ \vdots \\ y(L-1) \\ \vdots \\ y(N-1) \\ \vdots \\ y(N+L-1) \end{bmatrix} = \begin{bmatrix} h(0) & 0 & \cdots & 0 & \cdots & 0 \\ h(1) & h(0) & \cdots & 0 & \cdots & 0 \\ \vdots & \vdots & \cdots & \vdots & \cdots & \vdots \\ \vdots & \vdots & \cdots & \vdots & \cdots & \vdots \\ h(L-1) & h(L-2) & \cdots & h(0) & \cdots & 0 \\ \vdots & \vdots & \cdots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & \vdots & \cdots & h(0) \\ \vdots & \vdots & \cdots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & 0 & \cdots & h(L-1) \end{bmatrix} \begin{bmatrix} x(0) \\ x(1) \\ \vdots \\ x(L-1) \\ \vdots \\ x(N-1) \end{bmatrix}.$$

$DCT(\mathbf{y}) = DCT(H^{(N+L-1) \times N}_t) DCT(\mathbf{x})$

Filtering with periodic extension of input sequence and response

$$x(n+N)=x(n)$$

Input: $x(n)$, $n=0,1, \dots, N-1$ Periodically extended
with period N.

Filter: $h(n)$: $n=0,1,2, \dots, N-1$

Circular convolution

$$y(n) = x(n) \circledast h(n) = \sum_{m=0}^n x(m)h(n-m) + \sum_{m=n+1}^{N-1} x(m)h(n-m+N).$$

$\swarrow H_c^{(N \times N)}$

Can be made
equiv. to. linear
conv. by zero
padding of $x(n)$
and $h(n)$.

$$\begin{bmatrix} y(0) \\ y(1) \\ \vdots \\ y(N-1) \end{bmatrix} = \begin{bmatrix} h(0) & h(N-1) & \cdots & h(1) \\ h(1) & h(0) & \cdots & h(2) \\ \vdots & \vdots & \ddots & \vdots \\ h(N-1) & h(N-2) & \cdots & h(0) \end{bmatrix} \begin{bmatrix} x(0) \\ x(1) \\ \vdots \\ x(N-1) \end{bmatrix}.$$



Filtering with antiperiodic extension

Input: $x(n)$, $n=0, 1, \dots, N-1$

$$x(n+N) = -x(n)$$

Filter: $h(n)$: $n=0, 1, 2, \dots, N-1$

Antiperiodic extension extended with antiperiod N.

Skew Circular convolution

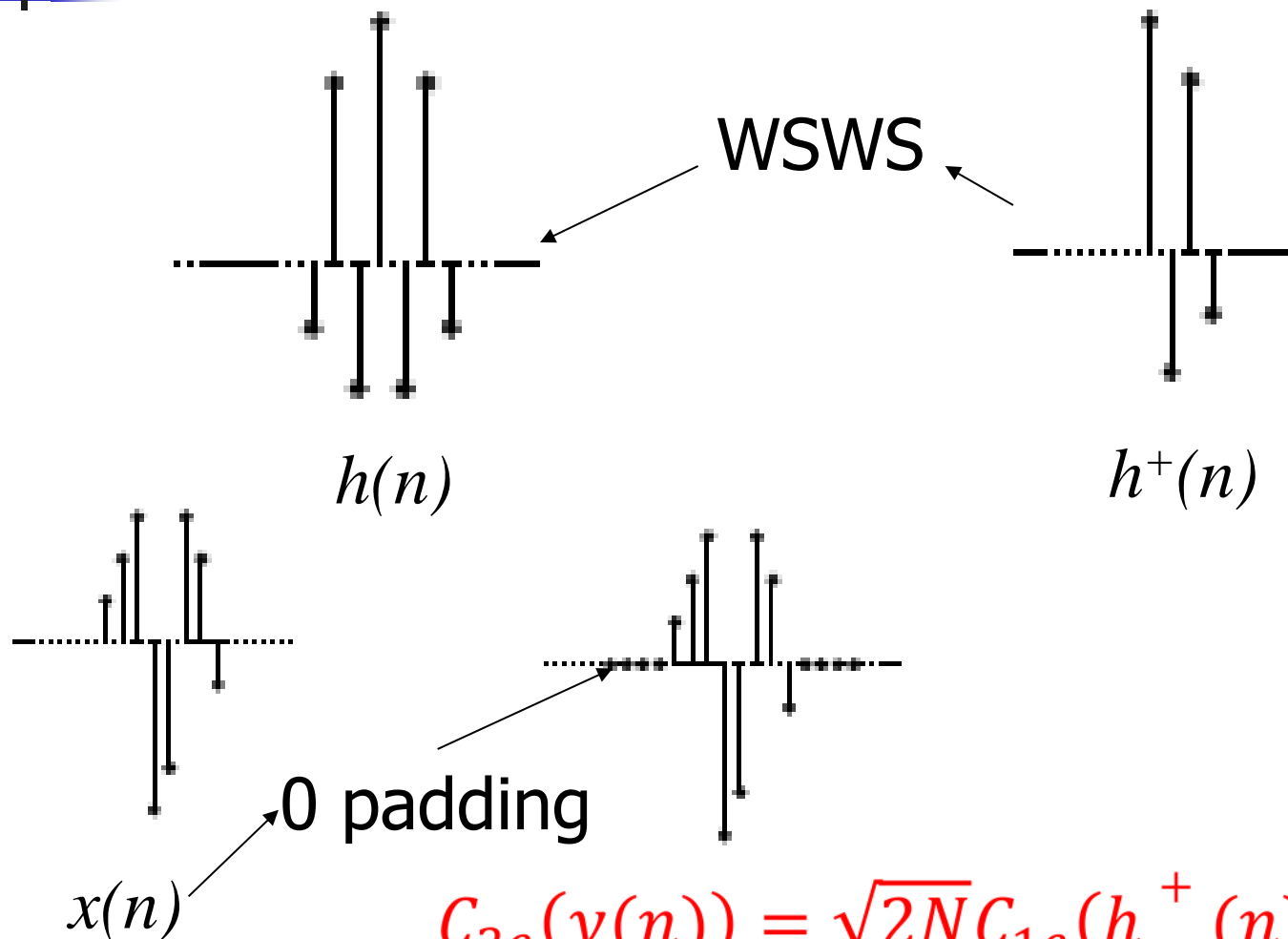
$$y(n) = x(n) \circledast h(n) = \sum_{m=0}^n x(m)h(n-m) - \sum_{m=n+1}^{N-1} x(m)h(n-m+N)$$

Can be made equiv. to. linear conv. by zero padding of $x(n)$ and $h(n)$.

$$H_s^{N \times N} = \begin{bmatrix} h(0) & -h(N-1) & \dots & -h(1) \\ h(1) & h(0) & \dots & -h(2) \\ \vdots & \vdots & \ddots & \vdots \\ h(N-1) & h(N-2) & \dots & h(0) \end{bmatrix}$$

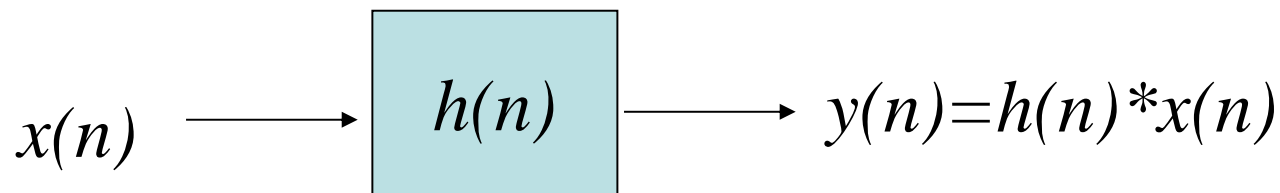
$H_s^{(N \times N)}$

Symmetric convolution and CMP for type II DCT

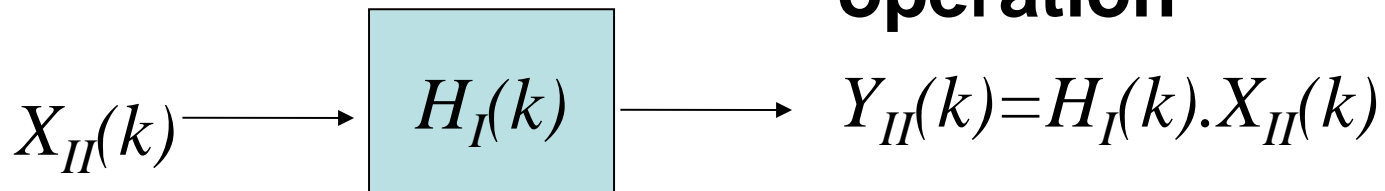


$$C_{2e}(y(n)) = \sqrt{2N} C_{1e}(h^+(n)) C_{2e}(x(n))$$

CONVOLUTION- MULTIPLICATION PROPERTY



↕ • **Symmetric convolution operation**



$$Y_{II}^{(N)} = \begin{bmatrix} H_{I,0} & 0 & 0 & 0 \\ 0 & H_{I,1} & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & H_{I,N-1} \end{bmatrix} X_{II}^{(N)}$$

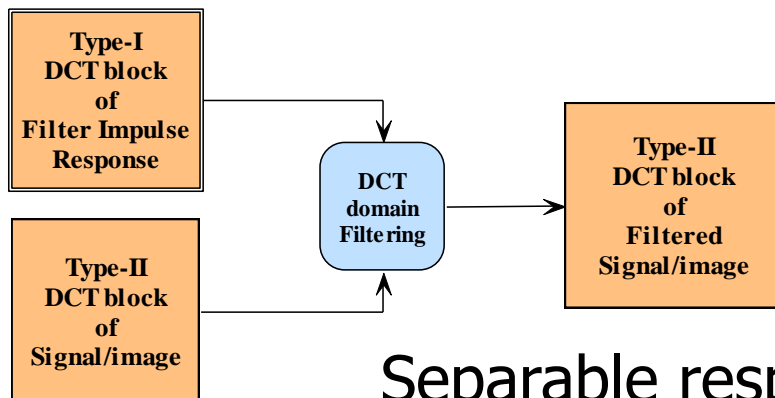
$\leftarrow \text{diag}(H_I)$

↖ Diagonal matrix formed by $H_I(k)$

Filtering in 2-D block DCT space

Given, **type-I DCT** of the impulse response of a filter and an input in **type-II DCT space**, filtered output can be transformed in the same space (i.e. type-II DCT space).

$$\begin{aligned} Y_{II}^{(N \times N)} &= (diag(H_I) (diag(H_I) X_{II}^{(N \times N)})^T)^T \\ &= diag(H_I) X_{II}^{(N \times N)} diag(H_I)^T \\ &= diag(H_I) X_{II}^{(N \times N)} diag(H_I) \end{aligned}$$



Separable response: $h(x,y)=h(x)h(y)$

BOUNDARY EFFECT IN BLOCK DCT DOMAIN

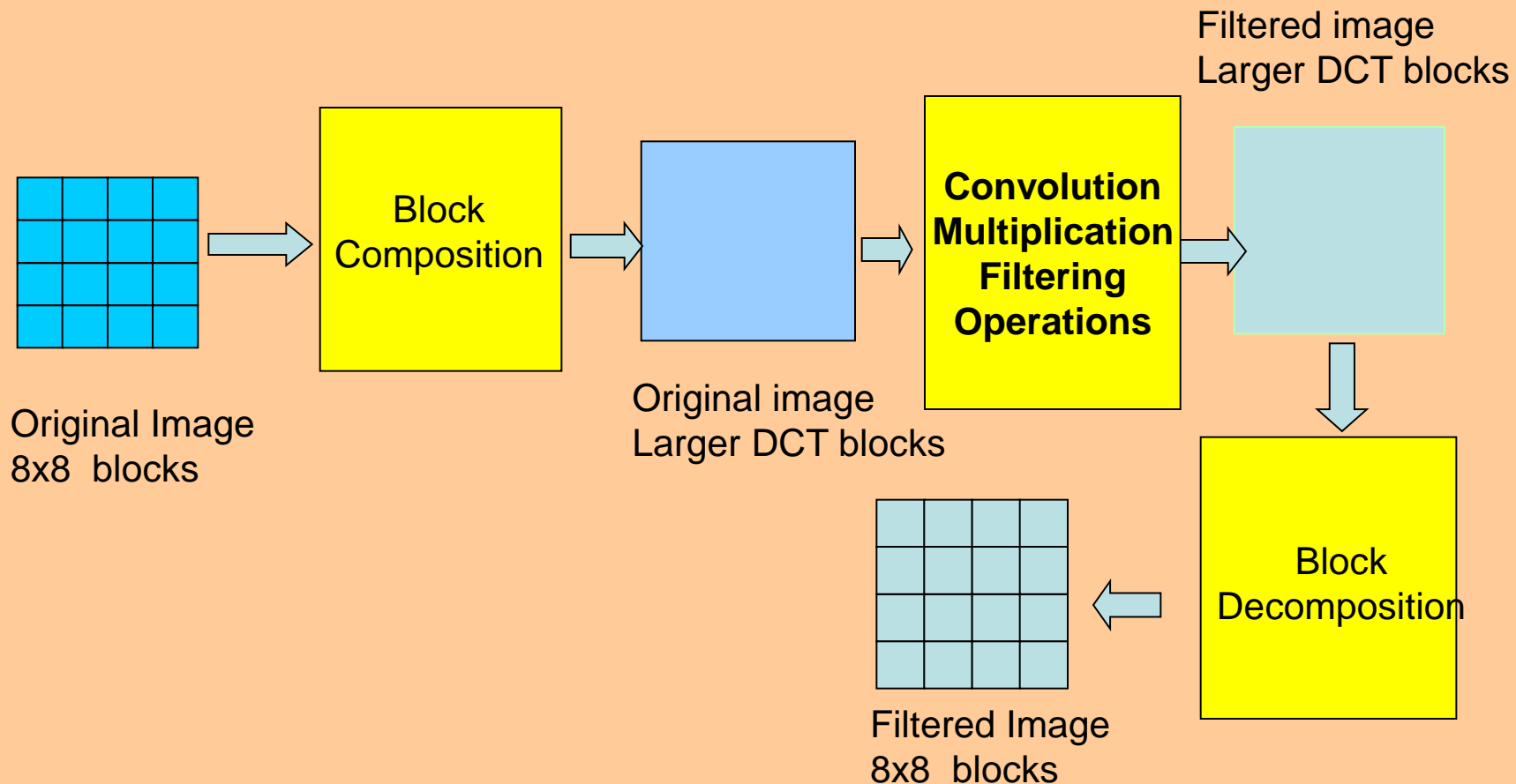


Linear Convolution



**Block Symmetric
Convolution**

Computation with larger blocks: The BFCD filtering Algorithm





Composite operation

- Block composition + Multiplication with filter coefficients + Block decomposition can be expressed by a composite matrix for the linear operation.

In 1-D: For filtering 3 NxN adjacent DCT blocks:

$$U^{(3N \times 3N)} = \underbrace{A_{(3,N)}^T}_{\text{Decomp.}} \underbrace{\mathbb{D}(\{\sqrt{6N} C_{3N}^T \mathbf{h}^+\}_{0}^{3N-1})}_{\text{Multiplication}} \underbrace{A_{(3,N)}}_{\text{Composition}}$$

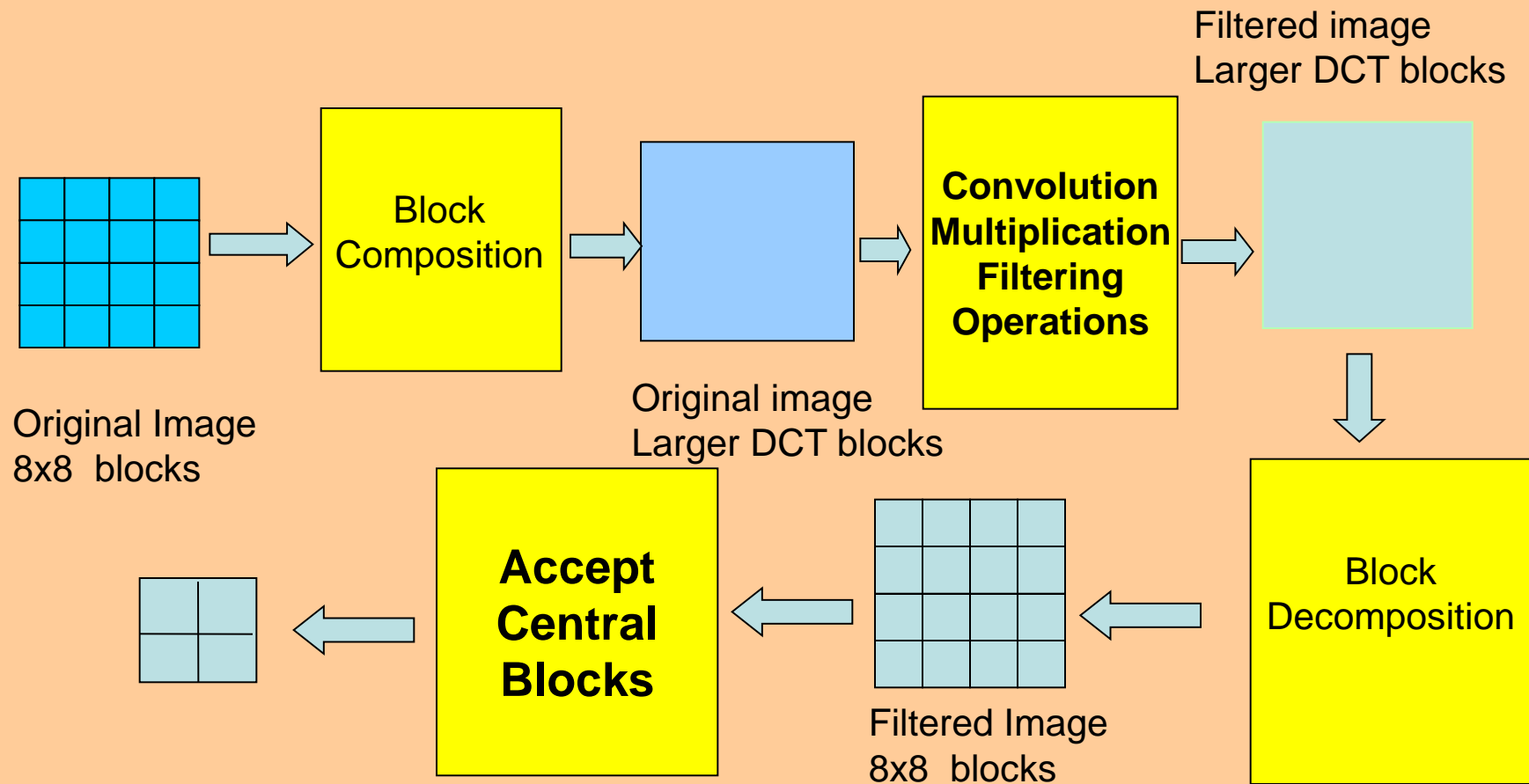
$$\begin{bmatrix} Y_1^{(N)} \\ Y_2^{(N)} \\ Y_3^{(N)} \end{bmatrix} = U \begin{bmatrix} X_1^{(N)} \\ X_2^{(N)} \\ X_3^{(N)} \end{bmatrix}$$



In 2-D

$$\begin{bmatrix} Y_{11}^{(N \times N)} & Y_{12}^{(N \times N)} & Y_{13}^{(N \times N)} \\ Y_{21}^{(N \times N)} & Y_{22}^{(N \times N)} & Y_{23}^{(N \times N)} \\ Y_{31}^{(N \times N)} & Y_{32}^{(N \times N)} & Y_{33}^{(N \times N)} \end{bmatrix} = U \begin{bmatrix} X_{11}^{(N \times N)} & X_{12}^{(N \times N)} & X_{13}^{(N \times N)} \\ X_{21}^{(N \times N)} & X_{22}^{(N \times N)} & X_{23}^{(N \times N)} \\ X_{31}^{(N \times N)} & X_{32}^{(N \times N)} & X_{33}^{(N \times N)} \end{bmatrix} U^T$$

Exact Computation: The Overlapping and Save (OBFCD) filtering Algorithm





Overlap and Save Strategy

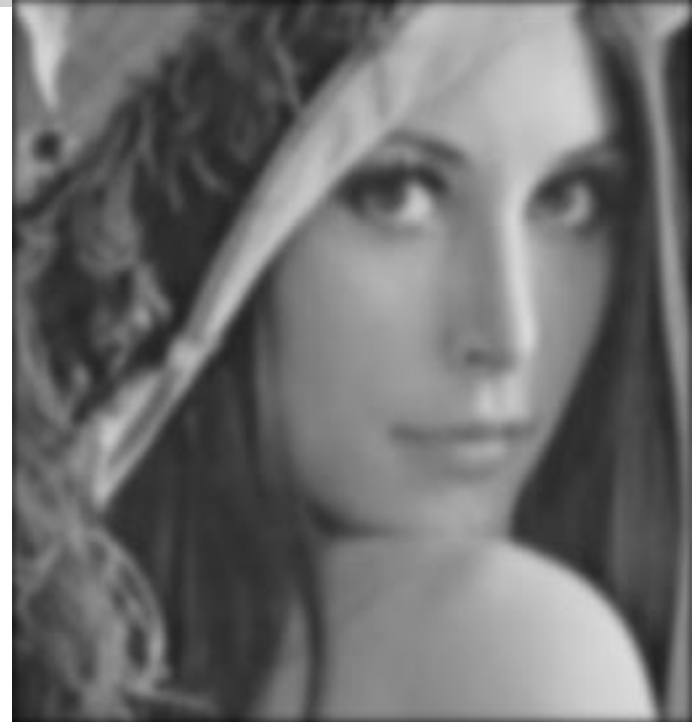
$$\begin{bmatrix} Y_{11}^{(N \times N)} & Y_{12}^{(N \times N)} & Y_{13}^{(N \times N)} \\ Y_{21}^{(N \times N)} & \color{red}{Y_{22}^{(N \times N)}} & Y_{23}^{(N \times N)} \\ Y_{31}^{(N \times N)} & Y_{32}^{(N \times N)} & Y_{33}^{(N \times N)} \end{bmatrix} = U \begin{bmatrix} X_{11}^{(N \times N)} & X_{12}^{(N \times N)} & X_{13}^{(N \times N)} \\ X_{21}^{(N \times N)} & X_{22}^{(N \times N)} & X_{23}^{(N \times N)} \\ X_{31}^{(N \times N)} & X_{32}^{(N \times N)} & X_{33}^{(N \times N)} \end{bmatrix} U^T$$

Save only the central block.

Filtered Images



BFCD Algorithm (5x5)



OBFCD Algorithm

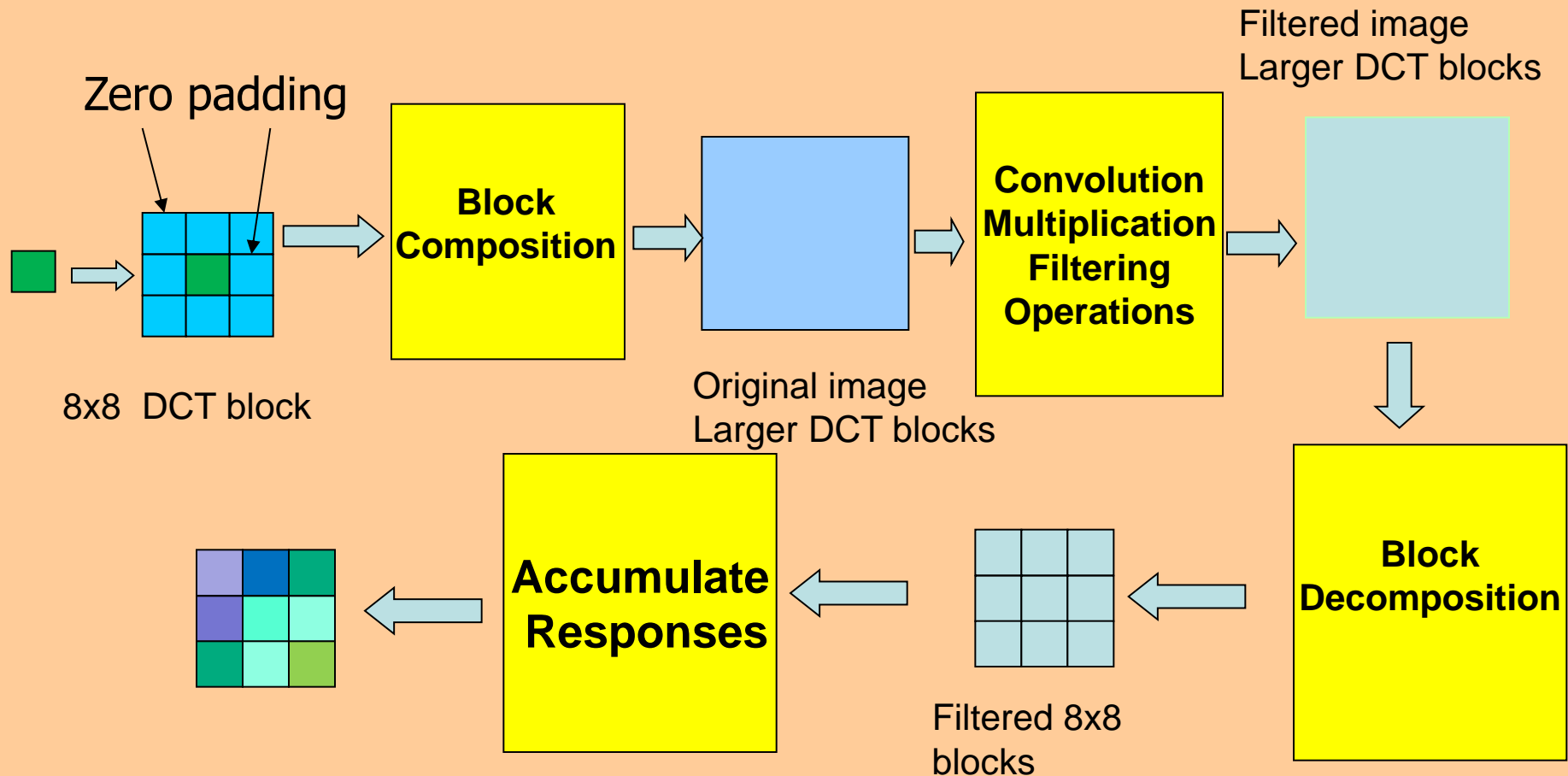


Overlap and add strategy

$$\begin{bmatrix} Y_{11}^{(N \times N)} & Y_{12}^{(N \times N)} & Y_{13}^{(N \times N)} \\ Y_{21}^{(N \times N)} & Y_{22}^{(N \times N)} & Y_{23}^{(N \times N)} \\ Y_{31}^{(N \times N)} & Y_{32}^{(N \times N)} & Y_{33}^{(N \times N)} \end{bmatrix} = U \begin{bmatrix} X_{11}^{(N \times N)} & X_{12}^{(N \times N)} & X_{13}^{(N \times N)} \\ X_{21}^{(N \times N)} & \color{red}{X_{22}^{(N \times N)}} & X_{23}^{(N \times N)} \\ X_{31}^{(N \times N)} & X_{32}^{(N \times N)} & X_{33}^{(N \times N)} \end{bmatrix} U^T$$

- For computing contribution of the central block $\color{red}{X_{22}}$ to neighboring blocks all other blocks in the input set to zeros.
- Add accumulated contribution at every block.

Exact Computation: The Overlapping and Add filtering Algorithm



Removal of Blocking Artifacts



Blocking artifacts
Compressed with quality
factor 10



Artifacts removed
By filtering in DCT
domain

Image Sharpening in DCT domain

$$X_s = X + k \cdot (X - X_{lpf})$$



Lena



Peppers

Color Sharpening

Original

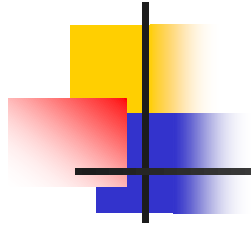


Filtered



Sharpened →





Color Enhancement

$$0 < \alpha < 1$$

$$\tilde{X}(i, j) = X(i, j) \left| \frac{X(i, j)}{Y_{max}} \right|^{\alpha-1}$$

Maximum Luminance

Alpha Rooting

- Smaller values get magnified and larger values get reduced.



Original



AR

Multi-contrast Enhancement (MCE)

- X : 8x8 DCT block: Only AC Coefficients modified.

- Spectral energy: $E_k = \frac{1}{N_k} \sum_{i+j=k} |X(i, j)|,$

- Contrast measure: $c_k = \frac{E_k}{\sum_{t < k} E_t}.$

- The coefficients of k -th band are scaled so that the resulting contrast becomes $w.c_k$, where w is a constant.

- By scaling coefficients in k -th band with $w.H_k$.

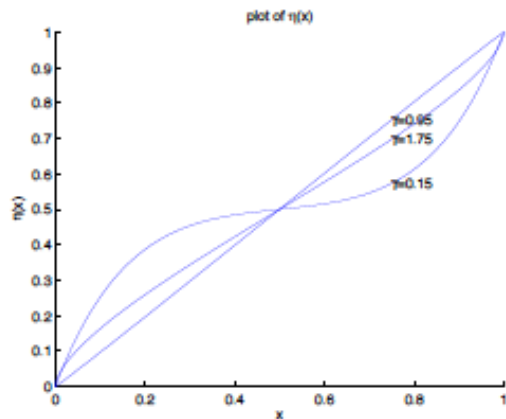
$$H_k = \frac{\sum_{t < k} \tilde{E}_t}{\sum_{t < k} E_t},$$

Spectral energy in the enhanced image

MCE with Dynamic Range Compression

- Also modifies DC coefficient, by increasing its dynamic range.
- Expression for spectral energy slightly modified.

$$E_k = \sqrt{\frac{1}{N_k} \sum_{i+j=k} |X(i, j)|^2}.$$



$$\eta(x) = \frac{(x^{\frac{1}{\gamma}} + (1 - (1 - x)^{\frac{1}{\gamma}}))}{2}, \quad 0 \leq x \leq 1.$$



Contrast : Definition

Let μ and σ denote the mean and standard deviation of an image. Contrast ζ of an image is defined here as:

$$\zeta = \frac{\mu}{\sigma}$$

Weber Law: $\zeta = \frac{\Delta L}{L}$

where ΔL is the difference in luminance between a stimulus and its surround, and L is the luminance of the surround

Theorem on Contrast Preservation in the DCT Domain

κ_d : the scale factor for the DC coefficient

κ_a : the scale factor for the AC coefficients

$$Y_e(i, j) = \begin{cases} \kappa_d Y(i, j), & i=j=0 \\ \kappa_a Y(i, j), & \text{otherwise} \end{cases}$$

The contrast of the processed image : κ_a / κ_d times of the contrast of the original image.

$\kappa_d = \kappa_a = \kappa$ preserves the contrast.

Preservation of Colors in the DCT Domain

U, V : Blocks of DCT coefficients of C_b and C_r
 κ :Scale factor for the luminance component Y

$$U_e(i, j) = \begin{cases} N(\kappa (\frac{U(i, j)}{N} - 128)) + 128, & i=j=0 \\ \kappa U(i, j), & \text{otherwise} \end{cases}$$

$$V_e(i, j) = \begin{cases} N(\kappa (\frac{V(i, j)}{N} - 128)) + 128, & i=j=0 \\ \kappa V(i, j), & \text{otherwise} \end{cases}$$

Color Enhancement by Scaling Coefficients (CES)

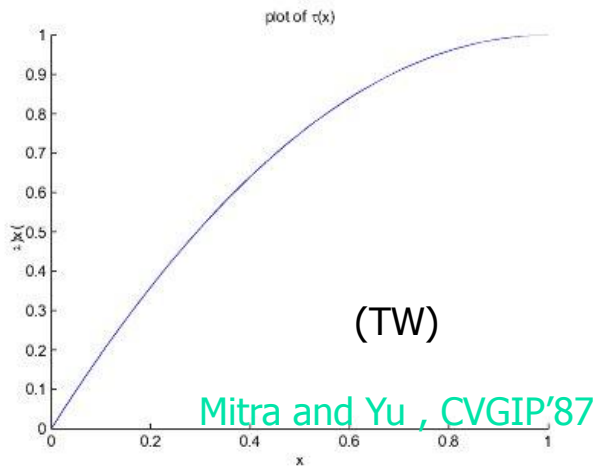
- Find the scale factor by mapping the DC coefficient with a monotonically increasing function.

$$1 \leq \kappa \leq \frac{B_{max}}{\mu + \lambda \cdot \sigma}$$

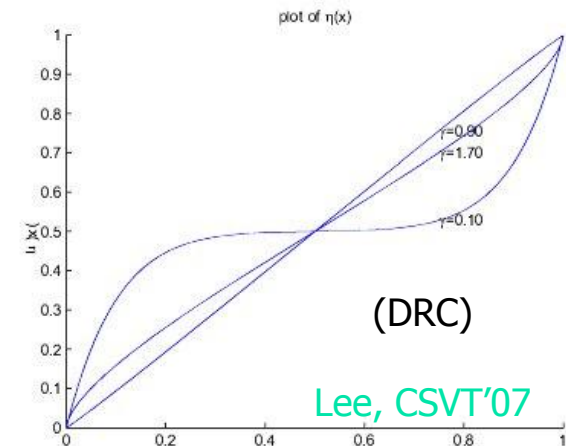
Scale factor κ ← Max. intensity

- Apply scaling to all other coefficients in all the components.
- For blocks having greater details judged by s.d., apply block decomposition and re-composition strategy.

Mapping functions for adjusting the local background illumination

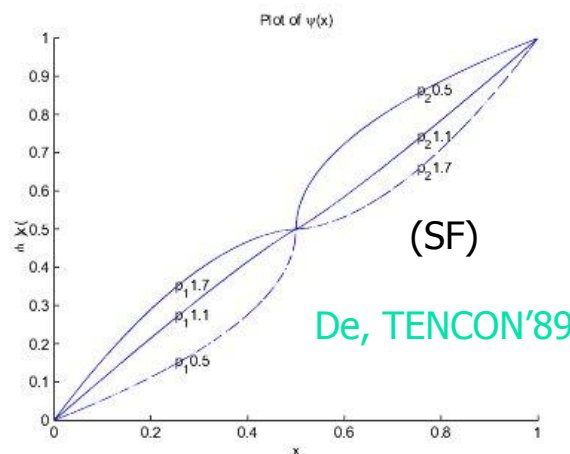


$$\kappa = \frac{f(Y(0,0))}{Y(0,0)}$$

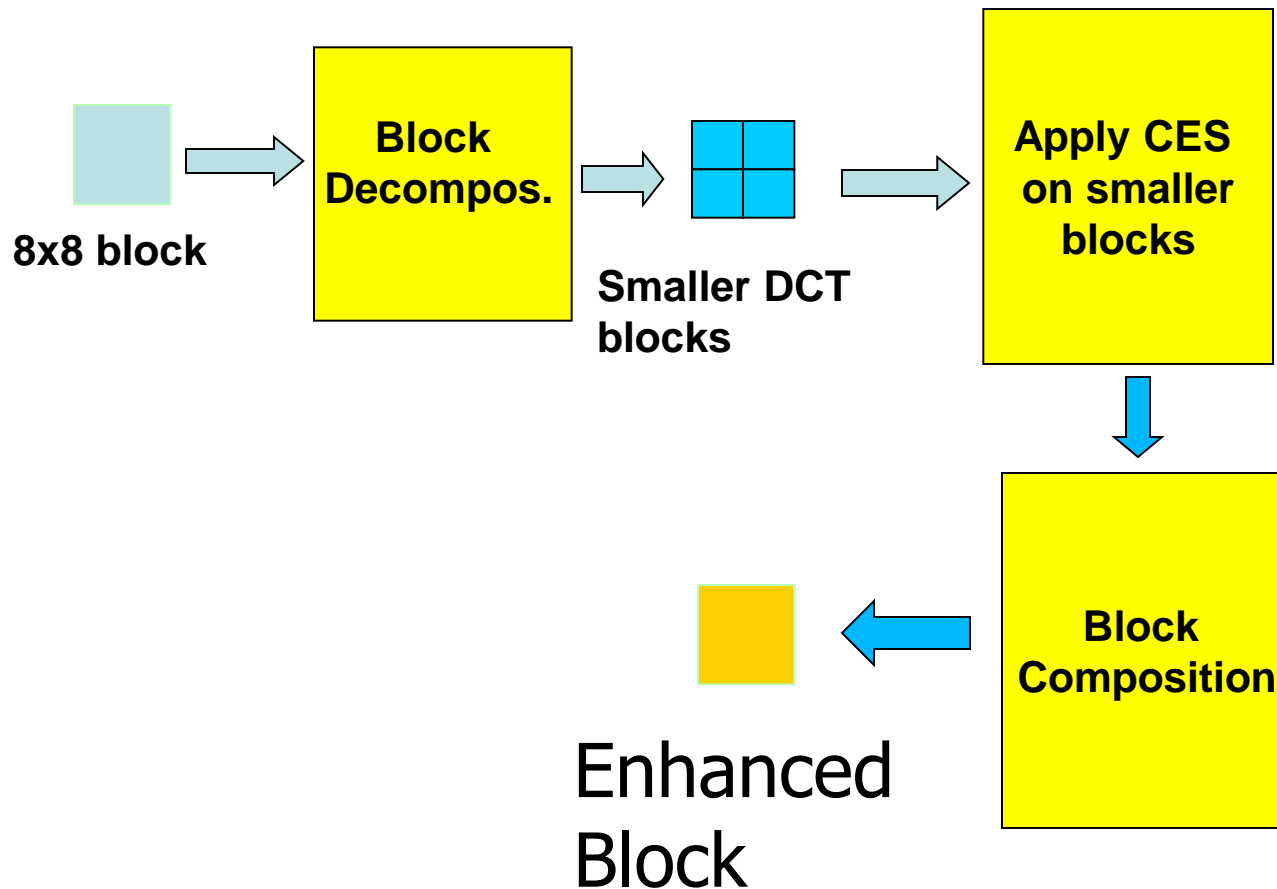


$$\tau(x) = x(2 - x)$$

$$\eta(x) = \frac{(x^{\frac{1}{\gamma}} + (1 - (1 - x)^{\frac{1}{\gamma}}))}{2}, \quad 0 \leq x \leq 1.$$



Enhancement of Blocks with more details





MCE



MCEDRC



CES

Some Results



original



AR



MCE



MCEDRC



TW-CES-BLK



MSR

Iterative Enhancement



Iteration no.=1

Run CES
iteratively.



Iteration no.=3



original



Iteration no.=2



Iteration no.=4



Color constancy

- Computation of color of illuminant
 - Avg. of colors (Gray world)
 - Maximum of each channel (White world)
 - Use of DC coefficient of DCT blocks in compressed domain.

- Diagonal Color correction: $(R_s, G_s, B_s) \rightarrow (R_d, G_d, B_d)$

$$\begin{aligned}
 k_r &= \frac{R_d}{R_s}, & k_g &= \frac{G_d}{G_s}, & k_b &= \frac{B_d}{B_s}, \\
 f &= \frac{R+G+B}{k_r R + k_g G + k_b B}, \\
 R_u &= f k_r R, & G_u &= f k_g G, & B_u &= f k_b B.
 \end{aligned}$$

- Chromatic Shift in Y-Cb-Cr:

$$\begin{aligned}
 Y_u &= Y, \\
 C_{bu} &= C_b + C_{bd} - C_{bs}, \\
 C_{ru} &= C_r + C_{rd} - C_{rs}
 \end{aligned}$$



Color correction: Example



Original



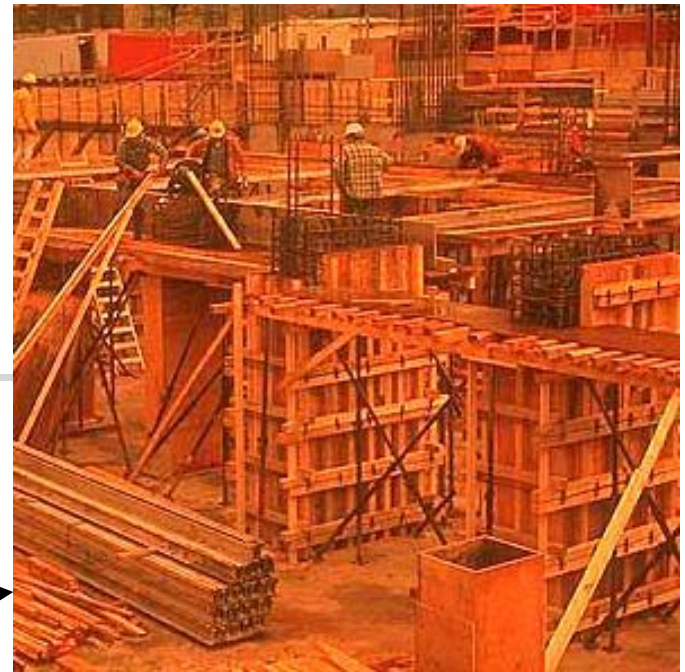
Diagonal Correction



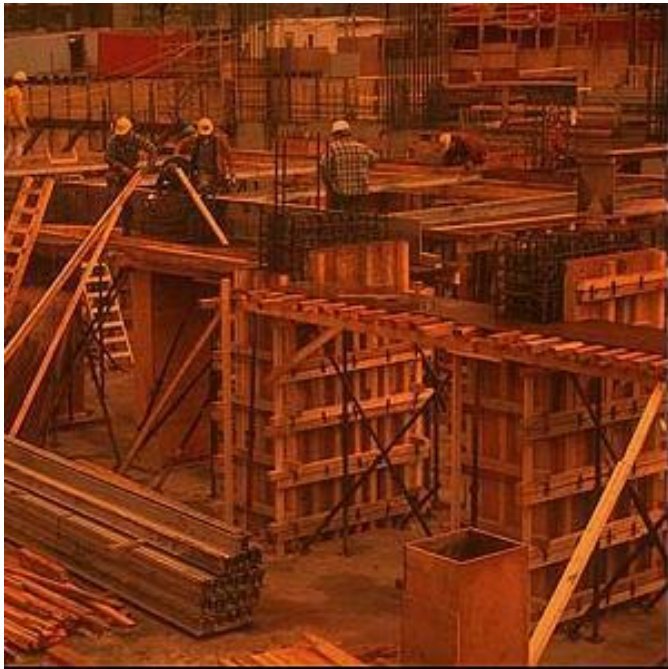
Chromatic Shift

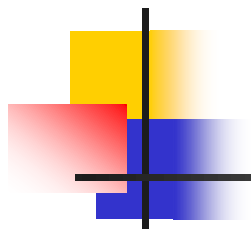
Color constancy coupled enhancement

- Perform color correction
- Perform color enhancement



Enhancement
without color
correction





Thank you!