

Image Compression

Image Compression

Fundamentals

\* Redundancy

Relative data redundancy RD.

$$RD = 1 - \frac{1}{CR}$$

$CR \rightarrow$  Compression ratio  $\Rightarrow CR = \frac{n_1}{n_2}$

$n_2 = n_1 \Rightarrow CR = 1 \Rightarrow RD = 0$   
 $n_2 \ll n_1, CR \rightarrow \infty, RD \rightarrow 1$   
 $n_2 \gg n_1, CR \rightarrow 0, RD \rightarrow 1$

\* Types of Redundancy

- 1) Coding redundancy.
- 2) Interpixel redundancy  $\rightarrow$  Spatial redundancy, geometric redundancy, Interscene " (mapping) / temporal "
- 3) Psychovisual "  $\rightarrow$

\* 3 can be identified and exploited.

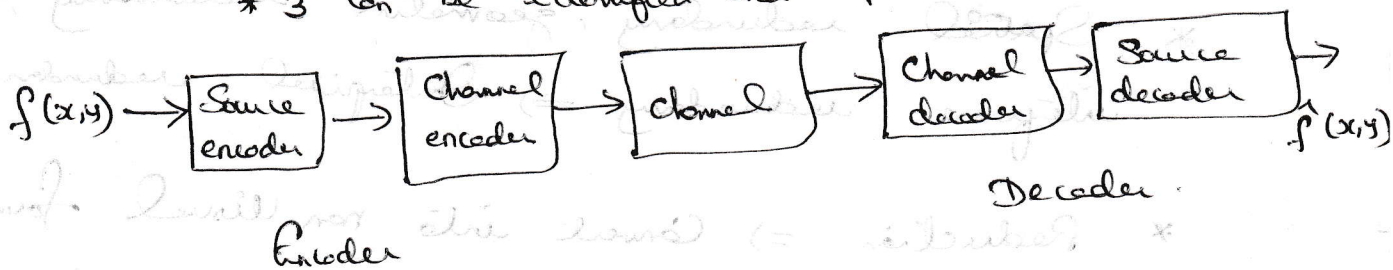


fig ①: General Compression system model.

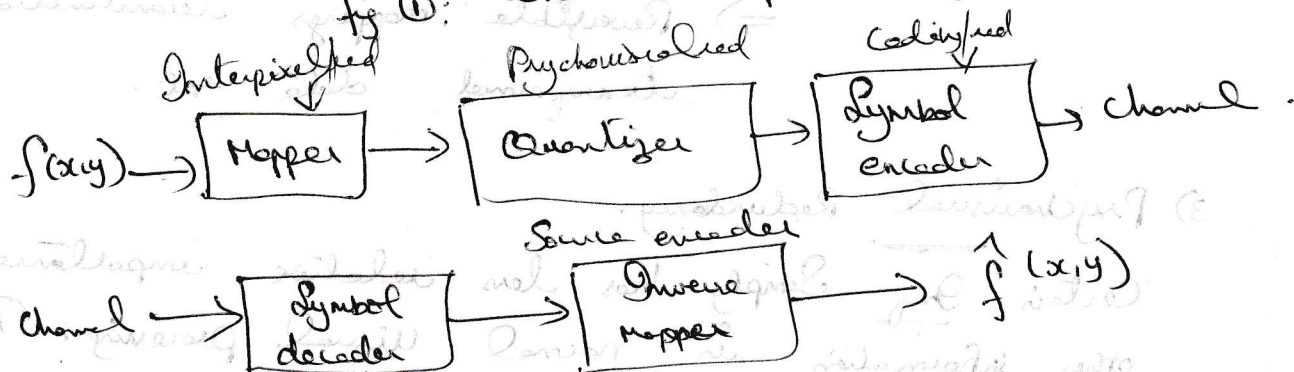


fig ②:

## 1) Coding redundancy.

\* Assigning fewer bits to more probable gray levels than to the less probable ones achieves data compression  $\Rightarrow$  Variable length coding.

\* If the gray levels of an image are coded in a way that uses more code symbols than absolutely necessary to represent each gray level  $\Rightarrow$  Coding redundancy.

## 2) Interpixel Redundancy :-

\* Value of any given pixel can be predicted from the value of its neighbors, Imp caused by individual pixels relatively small. \* Much of the visual contribution of a single pixel to an image is redundant, it could have been guessed on the basis of values of its neighbors  $\Rightarrow$

\* Spatial redundancy, geometric redundancy and interframe redundancy  $\Rightarrow$  Interpixel redundancy / Temporal redundancy

\* Reduction  $\Rightarrow$  Convert into non visual format.

$\Rightarrow$  Remove Interpixel redundancy  $\Rightarrow$  mappings

$\Rightarrow$  Reversible mappings reconstructed from transformed data set.

## 3) Psychovisual Redundancy.

Certain Imp simply has less relative importance than other information in normal visual processing. This Imp is said to be psychovisually redundant.

- \* Human perception of the Inf of an image  $\Rightarrow$  Does not involve quantitative analysis of every pixel value in the image.
- \* ~~General~~, Observer searches for distinguishing features such as edges & ) textural regions and mentally combines them into recognizable groupings.
- \* Unlike coding and interpixel redundancy, Psychovisual redundancy is associated with real & ) quantifiable visual information.
- \* Its elimination is ~~not~~ possible only because the Inf itself is not essential for normal visual processing.
- \* Since the elimination  $\Rightarrow$  Loss of quantitative Inf  $\Rightarrow$  Quantization.
- \* Quantization  $\Rightarrow$  Mapping of a broad range of I/p values to a limited no of o/p values.
- \* It is an irreversible operation (Visual Inf is lost),  $\Rightarrow$  Quantization results in lossy data comp.

# Image Compression Models.

## Source Encoder.

- \* Reducing (or) eliminating coding redundancy.
- \* Interpixel (or) Psychovisual.

### i) Mapper.

- \* Transforms o/p data into a (non visual) format designed to reduce interpixel redundancy in o/p image.
- \* Operation is reversible  $\Rightarrow$  may or may not reduce directly amount of data required to represent the image.
- \* Run length coding  $\rightarrow$  Example

### ii) Quantizer

- \* Reduces accuracy of mapper o/p in accordance with predetermined fidelity criterion.
- \* It reduces Psychovisual redundancy in the o/p image.
- \* Operation is irreversible.
- \* It is omitted  $\Rightarrow$  loss free compression is achieved.

### iii) Symbol Encoder.

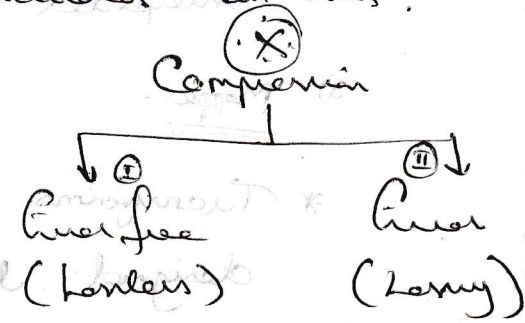
- \* It creates a fixed (or) VLC to represent Quantizer
- \* It reduces coding redundancy.

## Source decoder

- \* Symbol decoder & Inverse mapper  $\Rightarrow$  performs reversible operation.
- \* Quantization results in irreversible info loss, an Inverse Quantize block is not included in this.

## Channel Encoder & Decoder

- \* Hamming code used.



## Elements of Inf Theory.

- \* Measuring Information

$$I(E) = \log \frac{1}{P(E)} = -\log P(E)$$

Self Inf of an event.

## I Error Free Compression (Lorless Compression)

- Appln  $\Rightarrow$  Archival of medical (or) business documents
- $\Rightarrow$  Satellite imagery.

- \* It reduces inter pixel redundancies & Coding redundancies.

Types:-

- 1) Variable length coding  $\rightarrow$  VLC  $\left\{ \begin{array}{l} \rightarrow$  Huffman coding \\  $\rightarrow$  Arithmetic Coding \end{array} \right.
- 2) Lempel - Ziv - Welch (LZW) coding
- 3) Bit-plane coding  $\left\{ \begin{array}{l} \rightarrow$  Bit-plane decomposition. \\  $\rightarrow$  Constant area coding \\  $\rightarrow$  ID RLC \\  $\rightarrow$  2D RLC \end{array} \right.
- 4) Lorless predictive coding  $\left\{ \begin{array}{l} \rightarrow$  Context tracing and coding. \end{array} \right.

## II Lossy Compression:-

- 1) Lossy predictive coding  $\rightarrow$  DM  $\rightarrow$  DPCM  $\left[ \begin{array}{l} \text{Optimal prediction} \\ \text{Optimal Quantization} \end{array} \right.$
- 2) Transform coding  $\rightarrow$  Transform Selection  
Subimage Size Selection  
Bit allocation  $\left\{ \begin{array}{l} \text{Tonal coding implementation} \\ \text{Threshold coding implementation} \end{array} \right.$
- 3) Wavelet coding  $\rightarrow$  Wavelet Selection  
Decomposition Selection  
Quantizer design.

## Image Compression Standards

- 1) Binary image compression standards  $\rightarrow$  1D Compression
- 2) Continuous tone still image compression standards  $\rightarrow$  2D Compression  $\left\{ \begin{array}{l} \text{JPEG} \\ \text{JPEG 2000} \end{array} \right.$
- 3) Video compression standards  $\rightarrow$  DPCM/DCP  $\rightarrow$  MPEG/MPEG4

## Lossless Compression Method

### 1) Variable length coding (VLC):-

#  
util

- 3) Coding the individual elements within all blocks identically.
- 4) Adding special shift-up and shift-down symbols to identify each block.

(FCP), (FIT), (LZ)

# Arithmetic Coding

\* It generates non block codes.

\* one to one correspondence between Source Symbols and code words does not exist.

Adv  
Easy to implement a system with multiple arithmetic codes.  
One the computational machinery to implement one is developed all that is needed to setup

an entire sequence of source symbols is assigned a single arithmetic code word. The code word itself defines an interval of real numbers between 0 and 1.

\* No. of symbols in the msg ↑, Interval used to rep. it becomes smaller & no. of binary units used to rep. the Interval becomes larger.

Each symbol of msg reduces the size of the Interval according to its prob. of occurrence.

\* Entire half of the open Interval  $[0, 1)$ . → Starting boundary

\* Three decimal digits are used to represent five-symbol message.

## LZW Coding - [Lempel Ziv Welch]

\* The technique called LZW coding assigns fixed length code words to variable length sequences of source symbols but requires no a priori knowledge of the probability of occurrence of the symbols to be encoded.

\* (GIF), (TIFF) & (PDF).

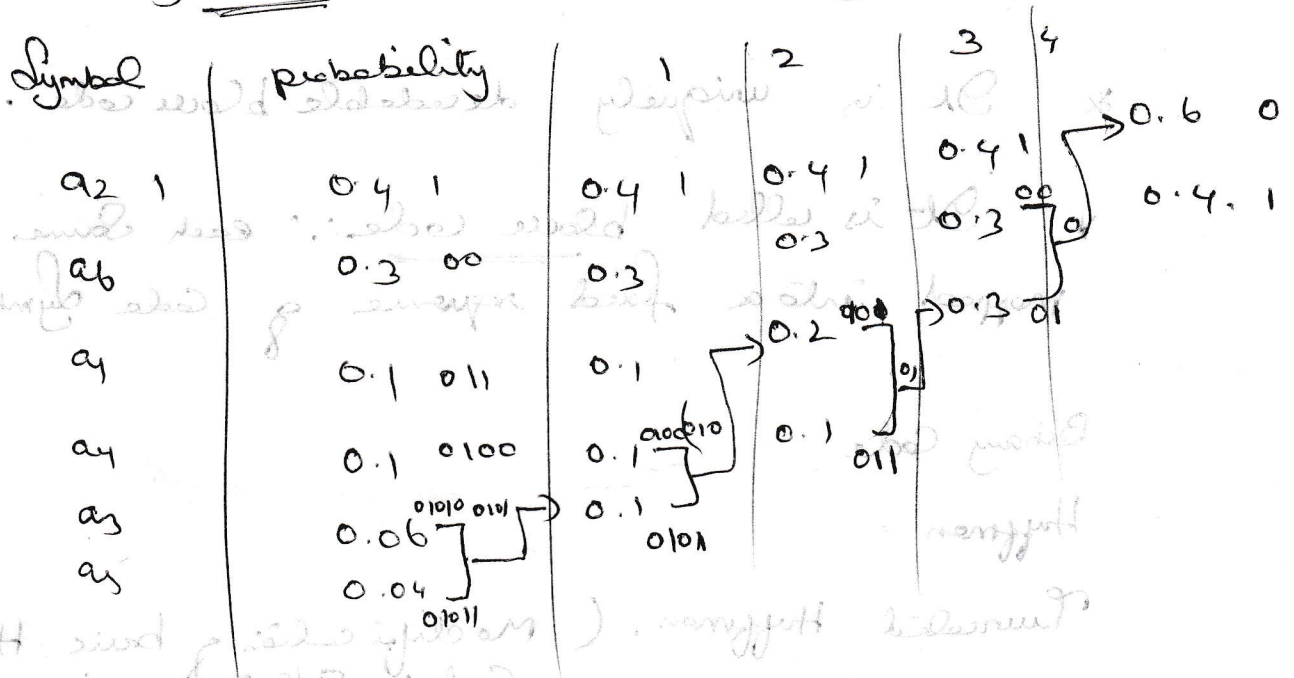
Free Compression :- [ \* Reduce coding redundancy is Simplest Approach ] (1)

Variable length coding

\* Coding redundancy normally present in any natural binary coding of the gray levels in an image.

\* Most popular technique  $\rightarrow$  Huffman \*

Original Source      Source reduction



Entropy (H)

$H =$  Average No of bits in codeword (L)

Redundancy (R) = 1 - code efficiency

$= 1 - H$

$H = P_1 \log_2 (1/P_1) + P_2 \log_2 (1/P_2) + \dots$

$L_{avg} = P_1 (\text{codeword})_1 + P_2 (\text{codeword})_2 + \dots$

H, L<sub>avg</sub>  $\Rightarrow$  units is bits/Symbol

$L_{avg} = 2.2 \text{ bits/symbol}$

$R_{entropy} = 2.14 \text{ bits/symbol}$

Huffman Code efficiency is  $\frac{0.973}{1} = 97.3\%$

Redundancy =  $0.026 \Rightarrow 2.6\%$

\* It creates the optimal code for a set of symbols and probabilities subject to the constraint that the symbols be coded one at a time.

\* It is uniquely decodable block code.

\* It is called block code: each source symbol is mapped into a fixed sequence of code symbols.

Binary Code

Huffman

Truncated Huffman. (Modification of basic Huffman Coding Strategy - Truncated Huffman Coding)  $\rightarrow$  prefix code.

Bz-Code  $\rightarrow$   $\pi$ , near optimal

Binary Shift. VLC.  $\rightarrow$  obeys power law of the form.

Huffman Shift.  $\rightarrow$  2 Int bits are used per continuation bits.  $P(x_j) = c^{-\beta}$

Shift Codes  $\Rightarrow$  a Shift code is generated by

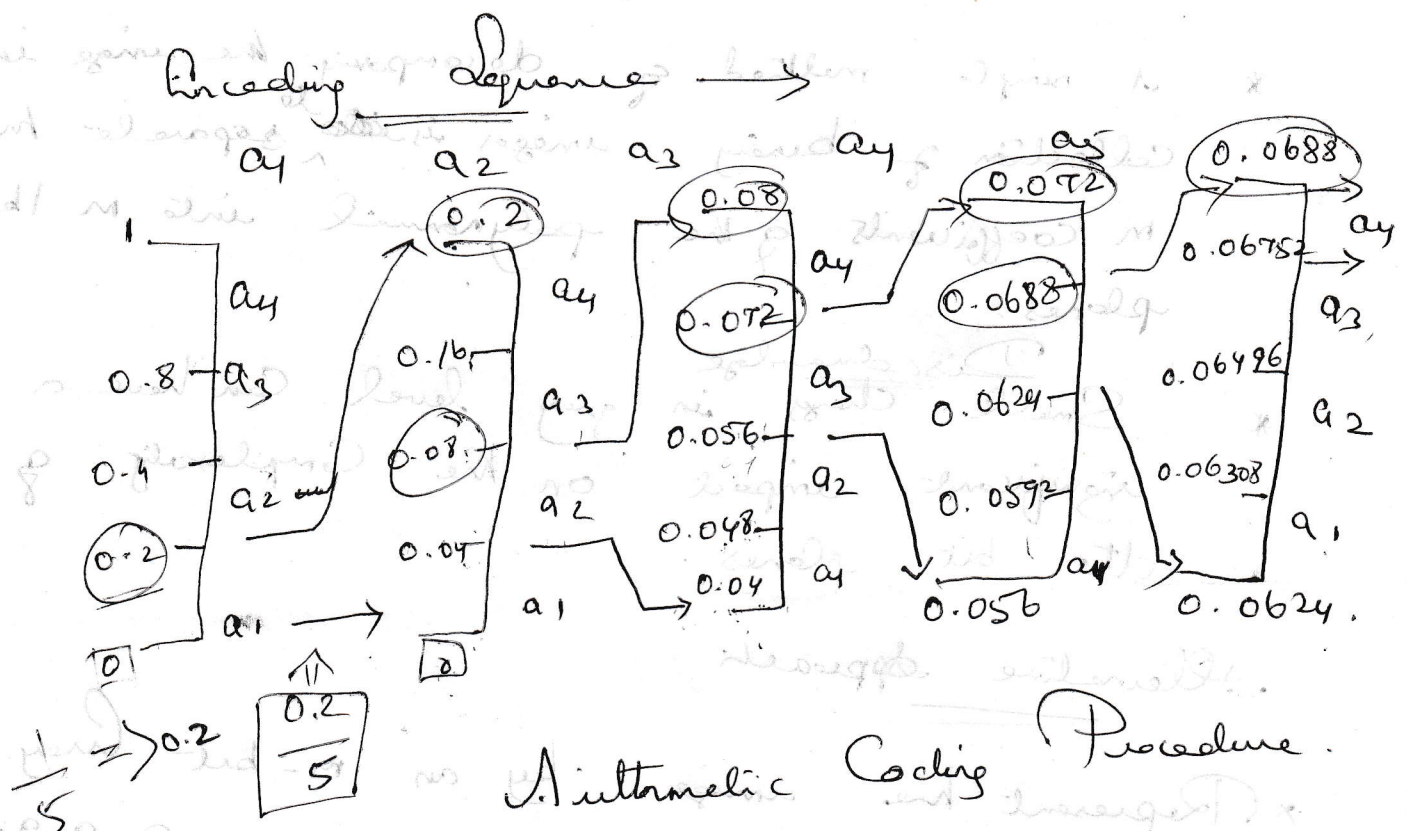
1) arranging the source symbols so that their probabilities are monotonically decreasing.

2) Dividing the total no. of symbols into symbol blocks of equal size.

# contd in previous page

# Arithmetic Coding

\* Here a five symbol sequence  $a_1 a_2 a_3 a_4 a_5$  from a 4 Symbol source is coded. At the start of the coding process  $\rightarrow$  entire half open Interval  $[0, 1)$ .



## Arithmetic Coding Procedure.

Source Symbol	Probability	Initial Subinterval
$a_1$	0.2	$[0.0, 0.2)$
$a_2$	0.2	$[0.2, 0.4)$
$a_3$	0.4	$[0.4, 0.8)$
$a_4$	0.2	$[0.8, 1.0)$

- \* 3 decimal digits are used to five Symbol message.
- \* This translates into  $\frac{3}{5}$  (or) 0.6 decimal bits/symbol and compares favorably with the entropy of the source.
- \* rounding strategy needed for this

## Bits plane coding:-

- \* It reduces Intrinsic redundancy.
- \* Decomposing a multilevel image into a series of binary images and Compressing each binary image via one of several well known binary Compression methods.
- \* A simple method of decomposing the image into collection of binary images is to separate the  $m$  coefficients of the polynomial into  $m$  1-bit planes.
- \* Disadvantage Small changes in gray level can have a significant impact on the complexity of the bit planes.

## Alternative approach

\* Represent the image by an  $m$ -bit Gray Code.

\* The  $m$  bit Gray code  $g_{m-1} \dots g_2 g_1 g_0$  that corresponds to polynomial (1) can be

Computed from.

$$a_{m-1} 2^{m-1} + a_{m-2} 2^{m-2} + a_{m-3} 2^{m-3} + \dots + a_1 2^1 + a_0 2^0 \quad (1)$$
$$g_i = a_i \oplus a_{i+1}, \quad 0 \leq i \leq m-2$$
$$g_{m-1} = a_{m-1}$$

- \* It has the unique property that successive codes words differ in only one bit position.
- \* Small changes in gray level are less likely to

\* when the gray level 127 and 128 are adjacent for instance, only 7<sup>th</sup> bit plane will contain a 0 to 1 transition, ∴ the binary codes that correspond to 127 and 128 are 11000000 and 01000000, respectively.

Constant Area Coding [CAC]

\* A simple but effective method of compressing a binary image (or) bit plane is to use special code words to identify large areas of contiguous 1's or 0's.

\* CAC ⇒ Image divided into blocks of size  $m \times n$  pixels ⇒ All white, all black (or) mixed intensity.

\* Most probable (or) frequently occurring category is then assigned the 1 bit code word 0 & the other 2 categories are assigned the 2 bit codes 10 & 11.

\* Compression ⇒ 1 bit or 2 bit code word.

→ Mixed intensity category is used as a prefix ⇒ followed by  $m \times n$  bit pattern of the block.

\* Solid white areas as 0 & all other blocks by a 1 followed by the bit pattern of the block.

This approach ⇒ white/black Shifting (WBS).

\* Effective method of modification ⇒ Code Solid white lines as 0's and all other lines with a 1 followed by the normal WBS Code Sequence.

\* Iterative approach  $\Rightarrow$  For 2D blocks, a solid white image is coded as 0 and all other images are divided into subblocks that are assigned a prefix as a 0 & all other images are divided into subblocks followed by a 0  $\Rightarrow$  Solid white.

### RLC

#### 1D RLC

\* Effective alternative to CAC represent each row of an image (or) bit plane by a sequence of lengths that describe successive runs of black & white pixels  $\Rightarrow$  RLC  $\Rightarrow$  facsimile (FAX) coding.

- \* Value of a run are i) To specify values first run of each row
- ii) To assume that each row begins with a white run, where run length may in fact be zero.

Entropy of black run length =  $H_0$ .

" " " white " " =  $H_1$ .

Approximate RLC entropy of the image is

$$H_{RLC} = \frac{H_0 + H_1}{L_0 + L_1}$$

where  $L_0$  &  $L_1$  denote the average values of black & white run lengths respectively.

\* It provides an average no. of bits / pixel required to code the run lengths in a binary image using a VLC.

2D Run length coding:-

\* Relative address coding (RAC)  $\rightarrow$  Based on the principle of tracking the binary transitions that begin & end each black & white run.

\* It requires adoption of a convention for determining run values.

\* In addition, irregular transitions at the start & end of each line, as well as an irregular stability line.

Contour tracing and coding:-

\* Relative address coding  $\Rightarrow$  Intensity transitions that make up the contours is a binary image.

\* Represent each contour by a set of boundary points or by a direct contour tracing.

\* Yet  $\rightarrow$  predictive differential Quantizing (PDQ)  
 $\Rightarrow$  Demeritates both approaches.  
 $\Rightarrow$  Scan line oriented contour tracing procedure.

\* The front & back contours of each object of an image are traced simultaneously to generate a sequence of pairs  $(\Delta', \Delta'')$ .

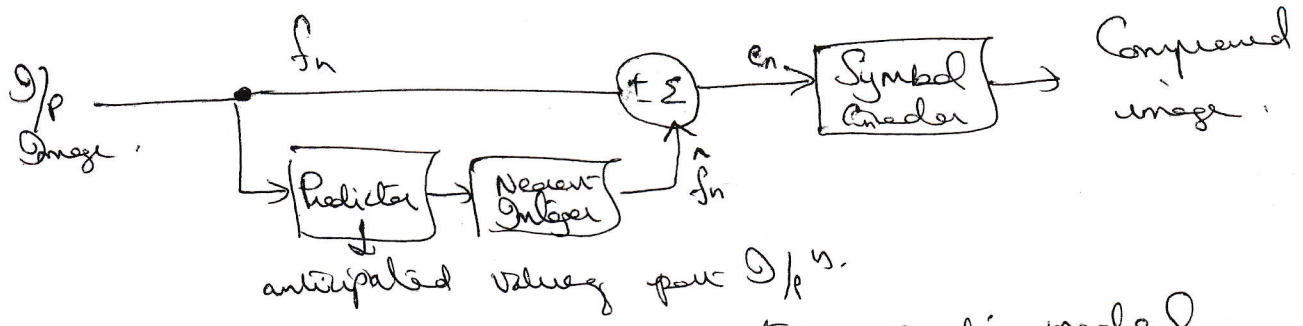


# Lossless predictive coding

\* It does not require decomposition of an image into collection of bit planes.

\* It is based on eliminating the interpixel redundancies of closely spaced pixels by collecting & coding only new information in each pixel.

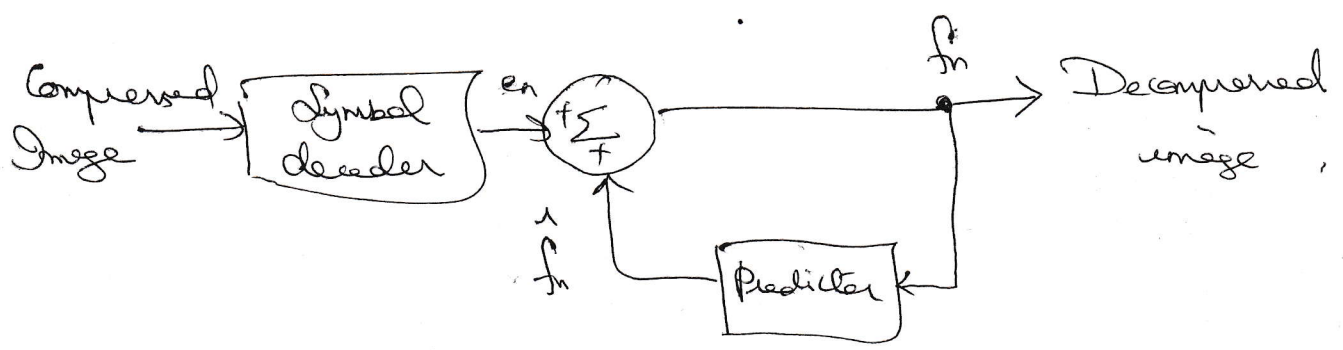
\* New Inq of a pixel is defined as diff between actual & predicted value of that pixel.



anticipated values from D/p's.

Lossless predictive coding model

a) encoder



prediction error

$$e_n = f_n - \hat{f}_n \quad \text{--- (1)}$$

$$f_n = e_n + \hat{f}_n \quad \text{--- (2)}$$

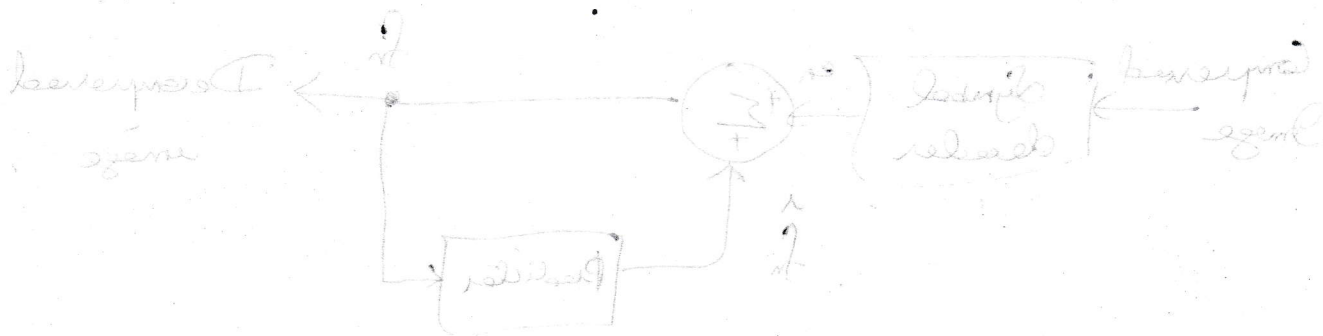
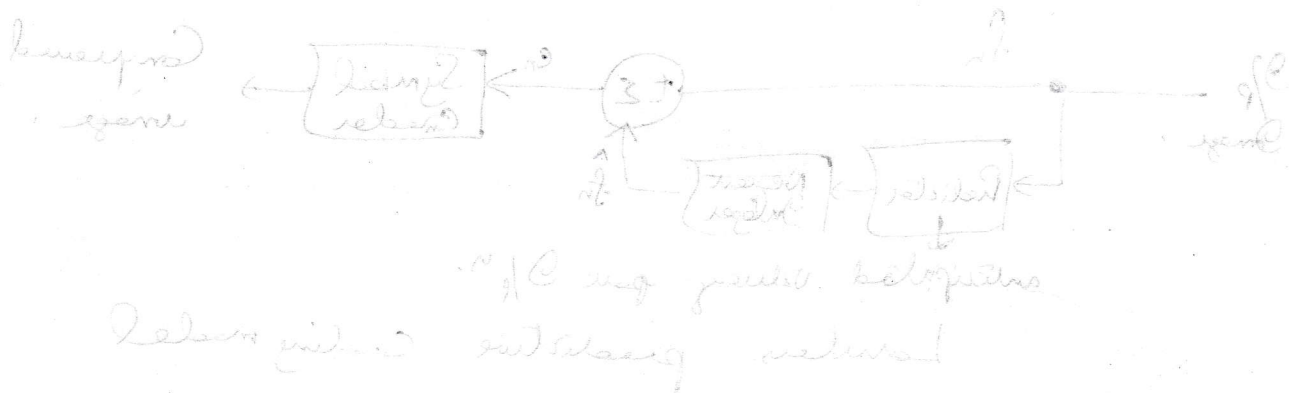
$$\hat{f}_n = \text{round} \left[ \sum_{i=1}^n \alpha_i f_{n-i} \right] \quad (3)$$

$$\hat{f}_n(x, y) = \text{round} \left[ \sum_{i=1}^m \alpha_i f(x, y-i) \right] \quad (4)$$

(4) Coeffs to be evaluated for the first  $m$  pixels of each line  $\Rightarrow$  Huffman code.

$\Delta$  predictor  $\Rightarrow$  previous pixel predictor.

$\Delta$  contextually predictive coding procedure applied to as  $\Delta$  on Predictive pixel coding.



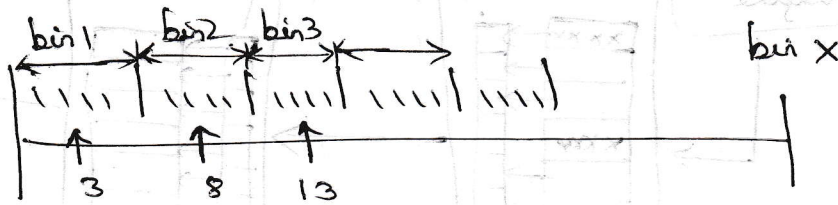
$$\textcircled{1} \quad \hat{f}_n - \hat{f}_{n-1} = \hat{e}_n$$

$$\textcircled{2} \quad \hat{f}_n + \hat{e}_n = \hat{f}_{n+1}$$

# Vector Quantization

①

- \* It is a popular technique for image compression.
- \* Consider a no of bins as shown in Fig. (a).



Example for vector Quantization.

- \* For each bin  $\Rightarrow$  an average value (or) reference value  
Average values for the bins 1, 2, 3 are 3, 8, 13 etc.
- \* Grouping  $\Rightarrow$  Data has 1 to 20 no's grouped to 4 groups.
- \* Mapping of accurate data (or) values into inaccurate values is called Quantization. It can be extended to vector data & that procedure is called as vector quantization.
- \* It maps  $k$  dimensional vectors in the vector space  $R^k$  into a finite set of vectors  $Y$  in the vector space  $R^k$ .  
vector  $Y$  can be given by the equation.  
$$Y = \{y_i : i = 1, 2, \dots, n\} \quad \text{--- (a)}$$
- \* Each vector  $y_i$  is called a code vector (or) code word & the set of all codewords is called a codebook.
- \* Associated with each codeword  $y_i$ , is a nearest neighbor region called Voronoi region & is defined  
$$V_i = \{x \in R^k : \|x - y_i\| \leq \|x - y_j\| \text{ for all } j \neq i\}$$

②

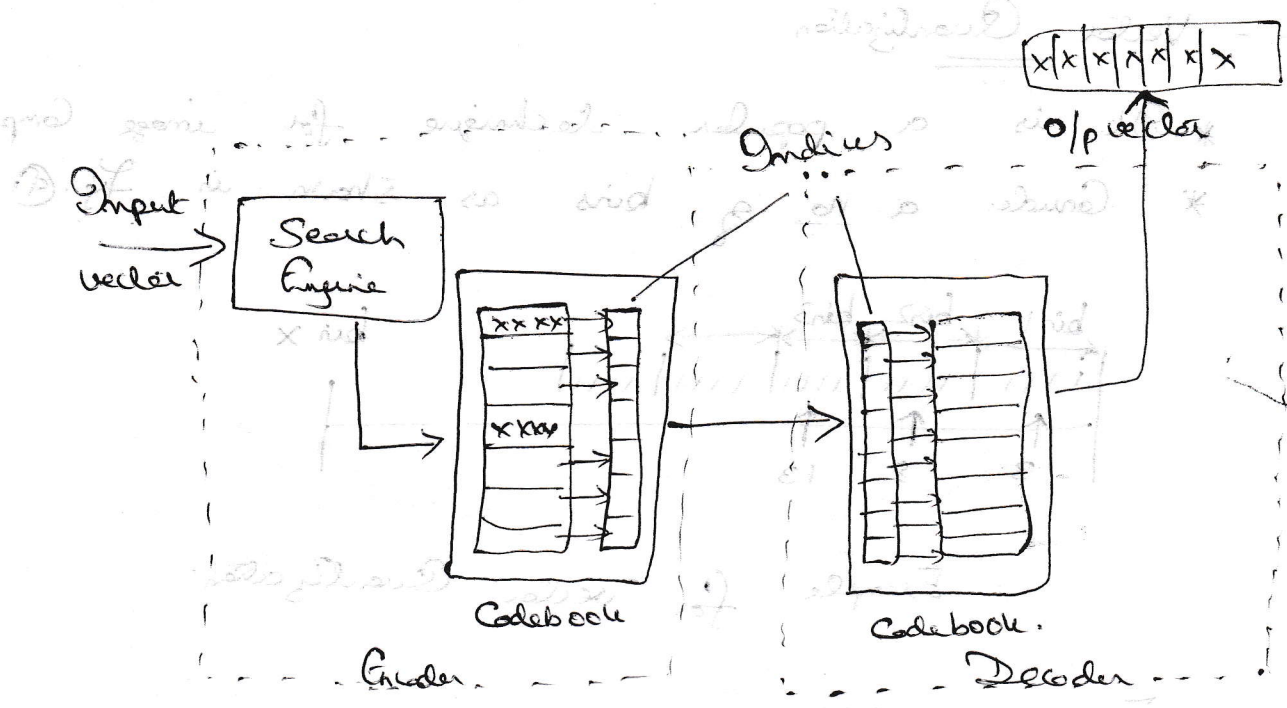


Fig 1: Image Compression technique using vector quantization.

Handwritten notes in Hindi describing the process of vector quantization. The text explains how an input vector is compared against a codebook to find the closest match, which is then used to reconstruct the image. The notes mention the 'Search Engine' and 'Codebook' components.

Additional handwritten notes in Hindi, including a mathematical formula:  $V = \{x \in R^N : \|x - y\| = \min_{z \in C} \|x - z\|\}$ . This formula defines the set of vectors  $V$  that are closest to a given input vector  $x$  within a codebook  $C$ .

# Image Compression Standards.

(2)

## Binary Image Compression Standards.

\* 1D Compression.

\* 2D Compression.

Most widely  $\Rightarrow$  Image Compression Stds are CCITT Group 3 & 4 Stds for binary image

Compression  $\Rightarrow$  utilised in FAX.

Group 3  $\Rightarrow$  non adaptive, 1D RLC,  $\Rightarrow$  RAC.

Group 4  $\Rightarrow$  Simplified Version of Group 3 Stand  
2D Coding is allowed.

$\Rightarrow$  Non adaptive

Both are used  $\Rightarrow$  typed & handwritten text.

$\Rightarrow$  data expansion.

JBIGA overcome  $\Rightarrow$  [Joint Bi-level Imaging Group].

CCITT & ISO  $\Rightarrow$  adopted & prepared several other  $\Rightarrow$  JBIGA1  $\Rightarrow$  adaptive

JBIGA2  $\Rightarrow$  2 to 4  $>$  JBIGA1.

## 1D Compression :-

Group 3 Compression  $\Rightarrow$  line  $\Rightarrow$  Coded as a series  
VLC  $\Rightarrow$  Run lengths of alternating white & black  
runs in a left to right scan of the line.

① \* RL < 63 => Terminating Code [Modified Huffman Code]  
 RL > 63 => Makeup Code in conjunction with a terminating code.

\* while run length code word => 00110101  
 => while run of length zero.

\* Finally EOL => end of line code word  
 e 00000000000001 => End of a sequence of image by 6 consecutive EOLs

2 D Compression -> Chap 3 & 4

-> line by line => position of blocks while (1) while to block run transition is coded w.v.t

position of a reference element as => coding line

-> Previously coded line is called reference line

\* Current reference element & associated chaining elements => 2 tests are performed select one of 3 possible modes => Run mode

Vertical mode  
 Horizontal mode

Continuous Low Still image Compression Standards

=> Several Cont line.

=> Lossy transform

# Video Compression standards

- \* 1) Video Coding Compression Std  $\Rightarrow$  H.261.
- \* 2) Multimedia Std.  $\Rightarrow$  multimedia video Compression Std  $\Rightarrow$  MPEG1, MPEG2 & MPEG4

## MPEG4

- 1) Improved Video Compression Efficiency.
- 2) Content based interactivity.
- 3) universal access.

Three basic types of encoded o/p frames.

- 1) I-frame.
- 2) P-frame.
- 3) B-frame.

# Lovry Compression

## Lovry Predictive Coding

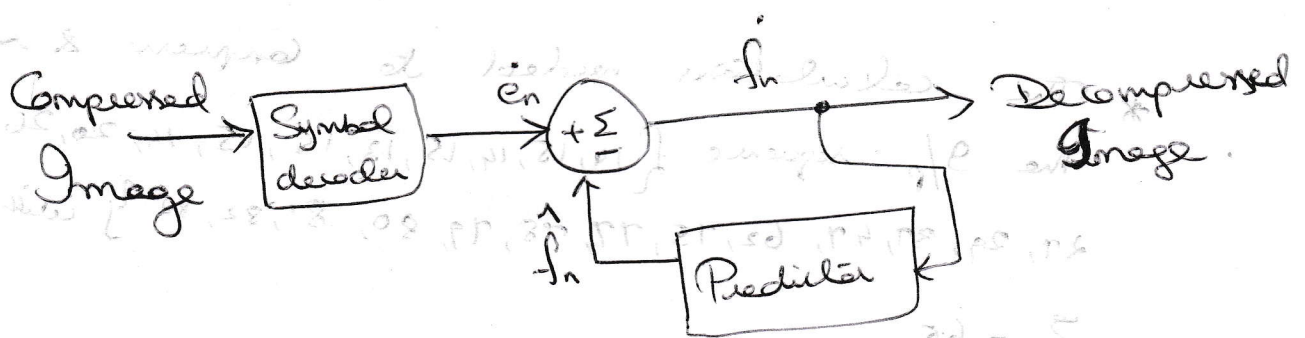
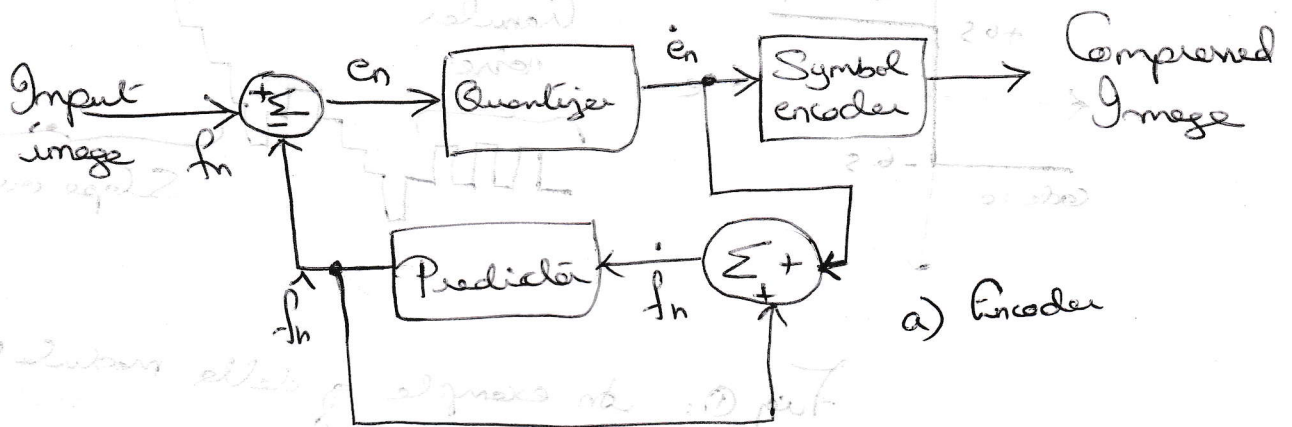


Fig ① - A Lovry Predictive Coding Decoder  
b) Decoder

$$\hat{f}_{n+1} = \hat{e}_n + \hat{f}_n \quad \text{--- ①}$$

### Delta modulation :-

DM is a simple  $\Rightarrow$

well known form of Lovry predictive coding in which predictor & Quantizer

$$\hat{f}_n = \alpha \hat{f}_{n-1}$$

$$e_n = \begin{cases} +1 & \text{for } e_n > 0 \\ -1 & \text{otherwise} \end{cases}$$

$\alpha$  is a prediction G. efficient (normally  $< 1$ ).

$\xi$  is a +ve constant.

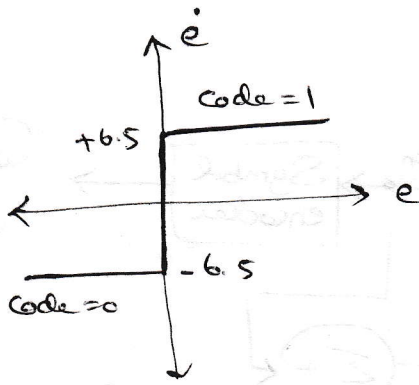


Fig 1: An example of delta modulation.

\* The calculations needed to compare & reconstruct the g/p sequence  $\{14, 15, 14, 15, 13, 15, 15, 14, 20, 26, 27, 28, 27, 27, 29, 37, 47, 62, 75, 77, 78, 79, 80, 81, 82, 82\}$  with  $\alpha=1$  &  $\xi=6.5$

\* Process begins with error free transfer of first g/p level. Initial condition  $\hat{f}_0 = f_0 = 14$  established at both encoder & decoder.

\* when  $n=1$   $\hat{f}_1 = (1) (14) = 14$ ,  $e_1 = 15 - 14 = 1$ ,  $e_1 = +6.5$ .  
 ( $\because e_1 > 0$ ),  $\hat{f}_1 = 6.5 + 14 = 20.5$ . & the resulting

reconstruction error is  $(15 - 20.5) = -5.5$  granular noise.   
 rapid change over.   
 $\Rightarrow$  g/p layers change   
 $\Rightarrow$  Slope overload occurs.

\* when  $\xi$  was too large  $\Rightarrow$  g/p's smallest changes as in relatively smooth region from  $n=20$  to  $n=7$ .   
 $\Rightarrow$  granular noise

## Optimal Prediction

(2)

\* It is used in most predictive coding apps, minimizes the encoder's mean-square prediction error.

$$E \{ e_n^2 \} = E \{ [f_n - \hat{f}_n]^2 \} \quad \text{--- (1)}$$

subject to the constraint that

$$\hat{f}_n = \hat{e}_n + \hat{f}_{n-1} \approx e_n + f_{n-1} = f_n \quad \text{--- (2)}$$

$$\hat{f}_n = \sum_{i=1}^M \alpha_i f_{n-i} \quad \text{--- (3)}$$

\* i.e. optimization criterion  $\Rightarrow$  minimize the mean-square prediction error  $\Rightarrow$  Quantization error is assumed to be negligible.  $\hat{e}_n \approx e_n$  and the prediction is constrained to a linear combination of  $m$  previous pixels.

\* These restrictions are not essential, but they simplify the analysis  $\Rightarrow$  Computational complexity of the predictor.

\* Resulting predictive coding approach is referred to as DPCM [differential pulse code modulation].

## Optimal Quantization

Staircase quantization function  $g(s) = \gamma(s)$  shown in fig is an odd function of  $s$   $\Rightarrow \gamma(-s) = -\gamma(s)$ . by  $t/2$  values of  $s_i$  &  $t_i$

Values of  $S_i$  &  $t_i \Rightarrow$  first quantile of the graph.

\* Breake points  $\Rightarrow$  line discontinuities  $\Rightarrow$  Decision & reconstruction levels of the quantizer.

\*  $S$  is considered to be mapped to  $t_i$  if it lies in the half-open interval  $(S_i, S_{i+1}]$ .

$$\int_{S_{i-1}}^{S_i} (S - t_i) P(S) dS = 0, \quad i = 1, 2, \dots, L/2 \quad \text{--- (1)}$$

$$S_i = \begin{cases} 0 & i=0 \\ \frac{t_i + t_{i+1}}{2}, & i=1, 2, \dots, L/2 - 1 \\ \infty & i=L/2 \end{cases} \quad \text{--- (2)}$$

$$S_{-i} = -S_i \quad \& \quad t_{-i} = -t_i. \quad \text{--- (3)}$$

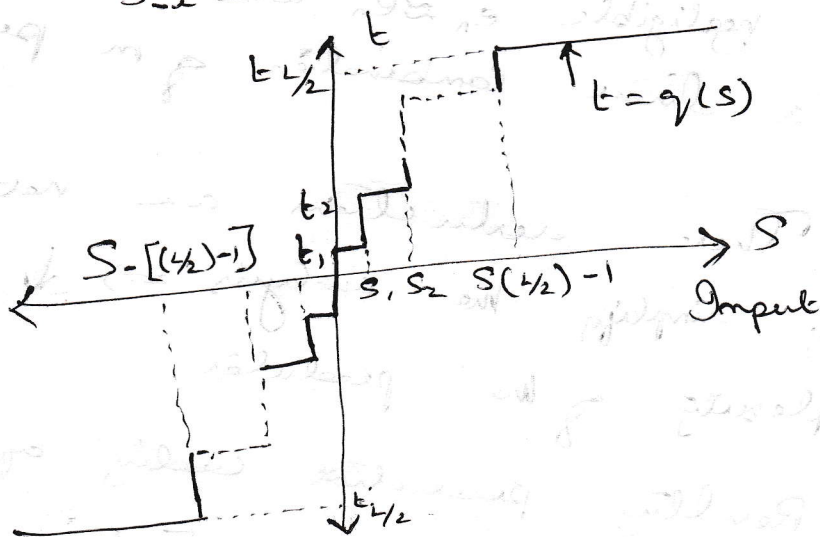


Fig (1): A Typical Quantization function.

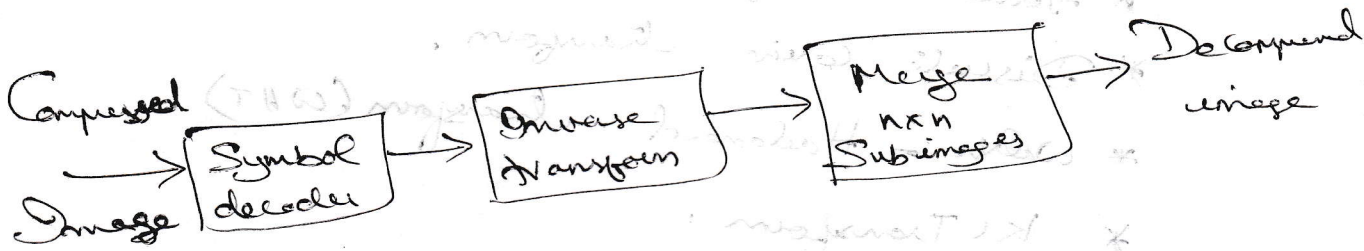
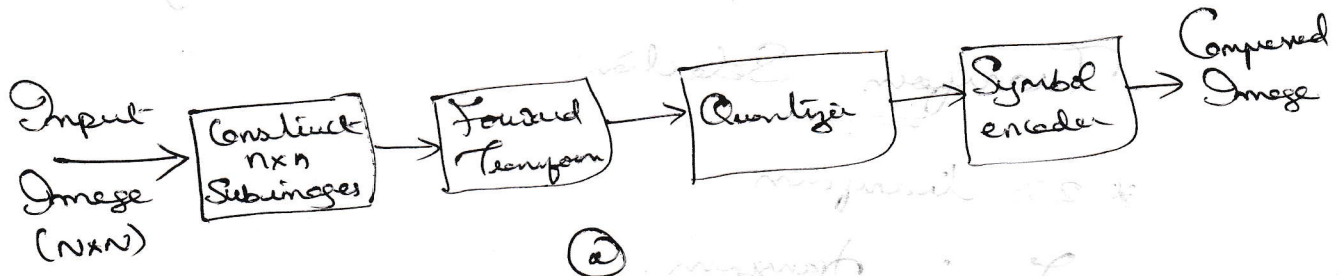
\* For any  $L$ , the  $S_i$  &  $t_i$  that satisfy eqn (1) to (3) are optimal in the mean square error sense.

$\Rightarrow$   $L$ -level Lloyd max quantizer.

# Transform Coding

\* Predictive coding techniques  $\Rightarrow$  Spatial domain methods.  
 \* modifying the transform of the image.

\* Transform coding  $\Rightarrow$  A reversible linear transform [Fourier transform]



## A Transform Coding System

a) Encoder

b) Decoder

\* Goal of transform coding  $\Rightarrow$  Decompose the pixels of sub image (a) to pack as much info as possible into the smaller no of transform Coeffts.

\* Quantization  $\Rightarrow$  Selectively eliminates  $\Rightarrow$  more Coarsely Quantizes the Coefficients that carry less info.  
 $\Rightarrow$  These Coefficients have the smallest impact on reconstructed Subimage Quality.

\* All of the transform encoding steps  $\Rightarrow$  low image content.

$\Rightarrow$  Adaptive transform coding.

\* Fixed for all Subimages  $\Rightarrow$  Non adaptive transform coding.

### Transform Selection

\* 2D transform

\* Fourier transform.

\* Discrete cosine transform.

\* Walsh - Hadamard transform (WHT).

\* KLT transform.

### Subimage Size [Allocation] selection.

\* Significant factor affecting transform coding error & Computational complexity is Subimage size.

\* Applies  $\Rightarrow$  images  $\xrightarrow{\text{Subdivided}}$  so that the correlation (redundancy) between adjacent Subimages is reduced to some acceptance level, so that  $n$  is an integer power of 2  $\Rightarrow n$  is the Subimage dimension.

\* Level of compression & Computational Complexity  $\uparrow$  as Subimage size  $\uparrow$ .

\* Popular Subimage sizes are  $8 \times 8$  and  $16 \times 16$ .

# Bit allocation

\* The reconstruction error associated with  $\Rightarrow$  Truncated series expansion of eqn.  $\Rightarrow$  a number & relative importance of transform Co-efficients are discarded, as well as precision that is used to represent the retained Co-efficients.

## Co-efficients

\* Transform coding system  $\Rightarrow$  Retained Co-efficients are selected on the basis of

a) Max Variance  $\rightarrow$  Zonal Coding.

b) Max Mag  $\rightarrow$  Threshold Coding.

Zonal coding  $\Rightarrow$  Information Theory  $\Rightarrow$  uncertainty.

$\Rightarrow$  Replacing Max Variance by 1.

" min " by 0.

## The overall process of Threshold Coding

\* Zonal Coding  $\rightarrow$  Replacing Inf as uncertainty.

\* Zonal coding  $\rightarrow$

It is implemented by a single fixed value of all

## Sub-images

\* Threshold Coding  $\rightarrow$  It is inherently adaptive i.e. adaptive transform coding.

Three ways.

- 1) A single global threshold  $\rightarrow$  all subimages.
- 2) Different threshold  $\rightarrow$  each subimage.
- 3) Threshold  $\rightarrow$  function of the location of each Co-efficient within the subimage.

\* In the first level of compression  $\Rightarrow$  varies from image to image to

\* In second level Lagrange coding  $\rightarrow$  No of Co-eff discarded for each subimage.

$\rightarrow$  Code rate is constant.

\* III<sup>rd</sup> technique  $\rightarrow$  Variable Code rate but open threshold and Quantization.

$$\hat{T}(u,v) = \text{round} \left[ \frac{T(u,v)}{Z(u,v)} \right]$$

$T(u,v)$  is a thresholded & quantized app of  $T(u,v)$  &  $Z(u,v)$  is an element of transform

normalization away.

$$k_c - \frac{c}{2} \leq T(u,v) \leq k_c + \frac{c}{2}$$

# Wavelet Coding

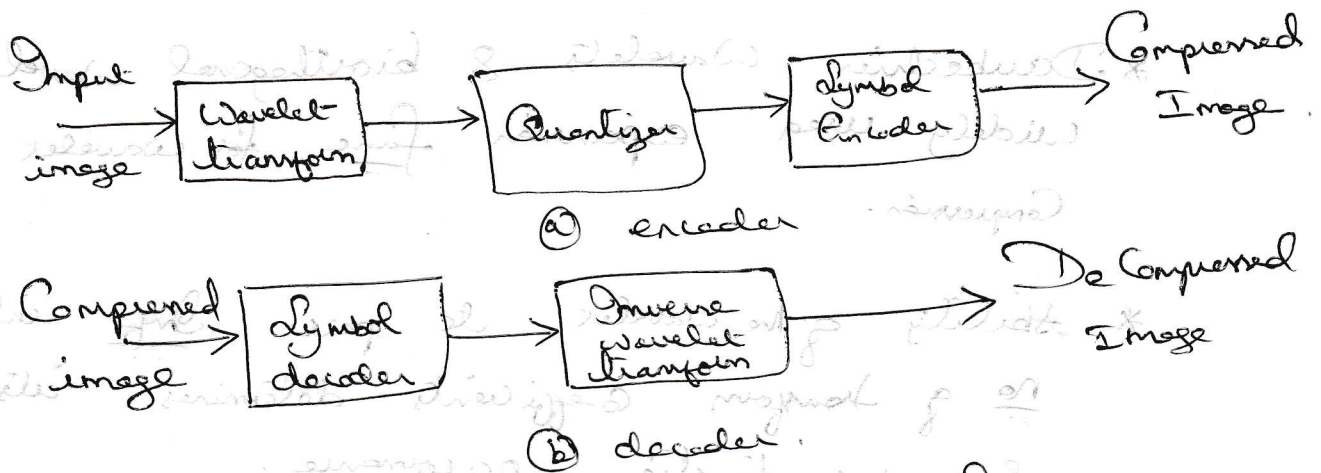


Fig ① wavelet Coding System.

## Wavelet Coding

\* It is based on the idea that the  $C$ -coefficients of a transform that decorrelates the pixels of an image can be coded more efficiently than the original pixels themselves.

\* Transform's basis functions  $\rightarrow$  Wavelets — Pack most of the important visual info into a small no. of  $C$ -coefficients, the remaining  $C$ -coefficients can be quantized coarsely or truncated to zero with image distortion.

- \* Diff. between wavelet & transform
- \* It is computationally efficient & inherently lossless. [basis functions are limited in duration], Subdivision of the signal image is unnecessary.
- \* Removal of subdivision eliminates the blocking artifact that characterizes DCT based app at high compression ratios.

## Wavelet selection

\* Daubechies Wavelets & biorthogonal wavelet  $\rightarrow$  widely used expansion func for wavelet based Compression.

\* Ability of the wavelet to pass Info into a small No of transform coefficients determines its Compression & reconstruction performance.

## Decomposition level Selection

\* No. of transform decomposition levels.

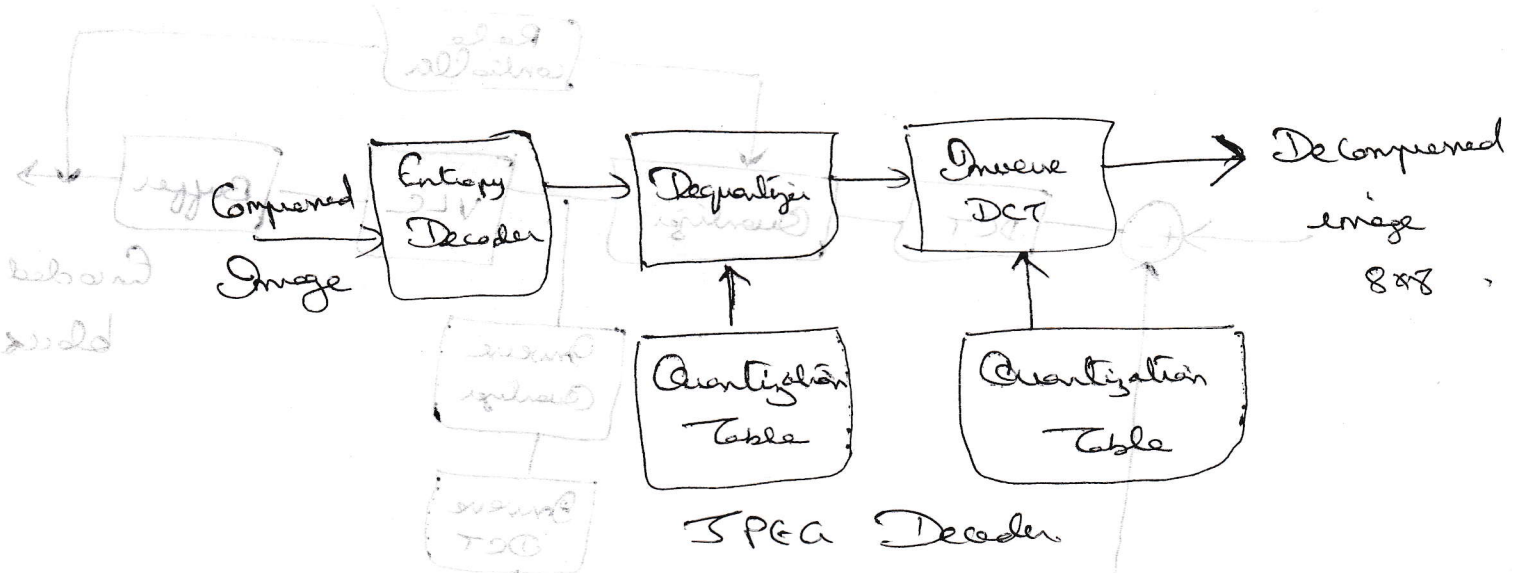
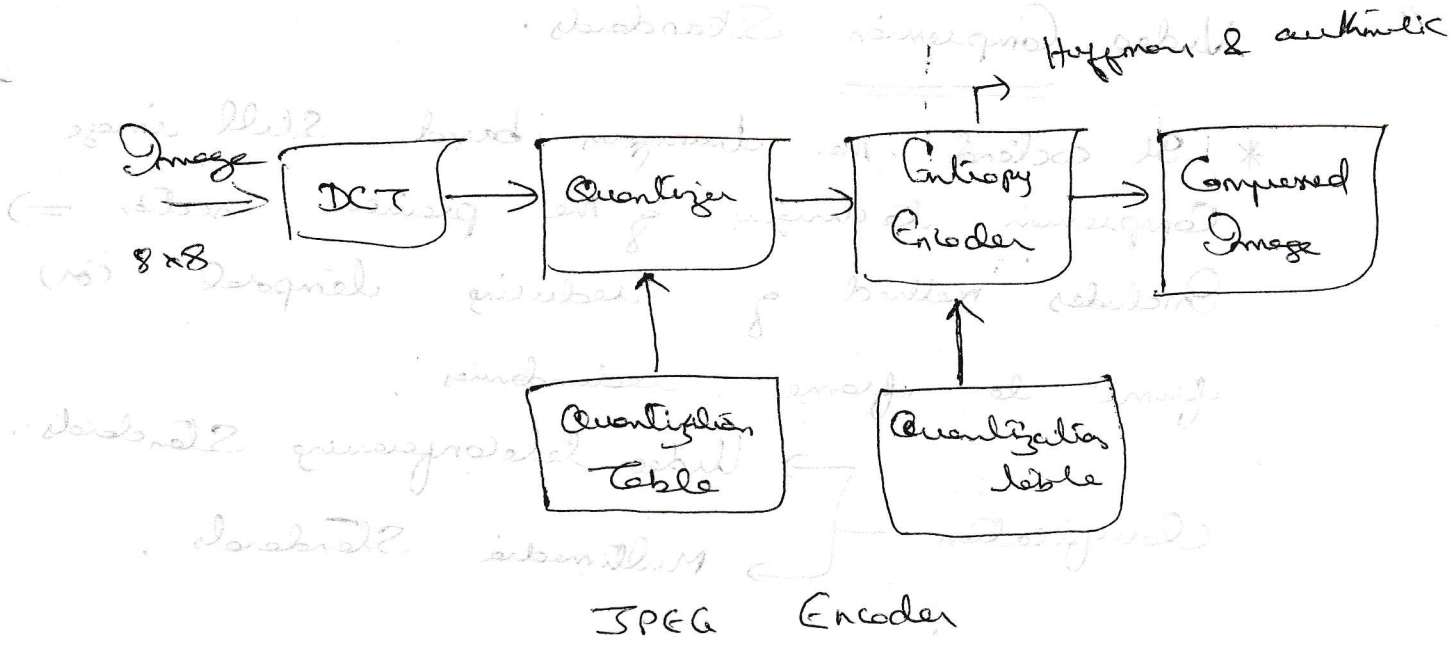
\* P-Scale wavelet transform involves P filter bank iterations, No. of operations in the computation of the forward and inverse transform  $\uparrow$  with No. of decomposition levels.

## Quantizer design

\* Layered filter affecting wavelet Coding Compression reconstruction  $\times$  is coefficient quantization.

\* Quantizers are uniform  $\rightarrow$  1) Introduce an enlarged quantization interval around zero called a dead zone (0) 2) adapting the size of the quantization interval from Scale cos Scale.

\* Selected Quantization levels must be trans to decoder with encoded image bit stream.



\* DC Coefficients & AC Coefficients

↓

JPEG

- 3 different Coding Systems
- i) Lossy baseline Coding System → Based on DCT & adequate for most compression apps.
  - ii) Extended Coding System → greater compression higher precision progressive recon apps!
  - iii) Lossless independent Coding System → Reversible Compression.

# Video Compression Standards.

\* It extends the transform based, still image compression techniques of the previous section  $\Rightarrow$  Includes method of reducing temporal (or) frame to frame redundancies.

Classification  $\left\{ \begin{array}{l} \text{Video teleconferencing Standards.} \\ \text{Multimedia Standards.} \end{array} \right.$

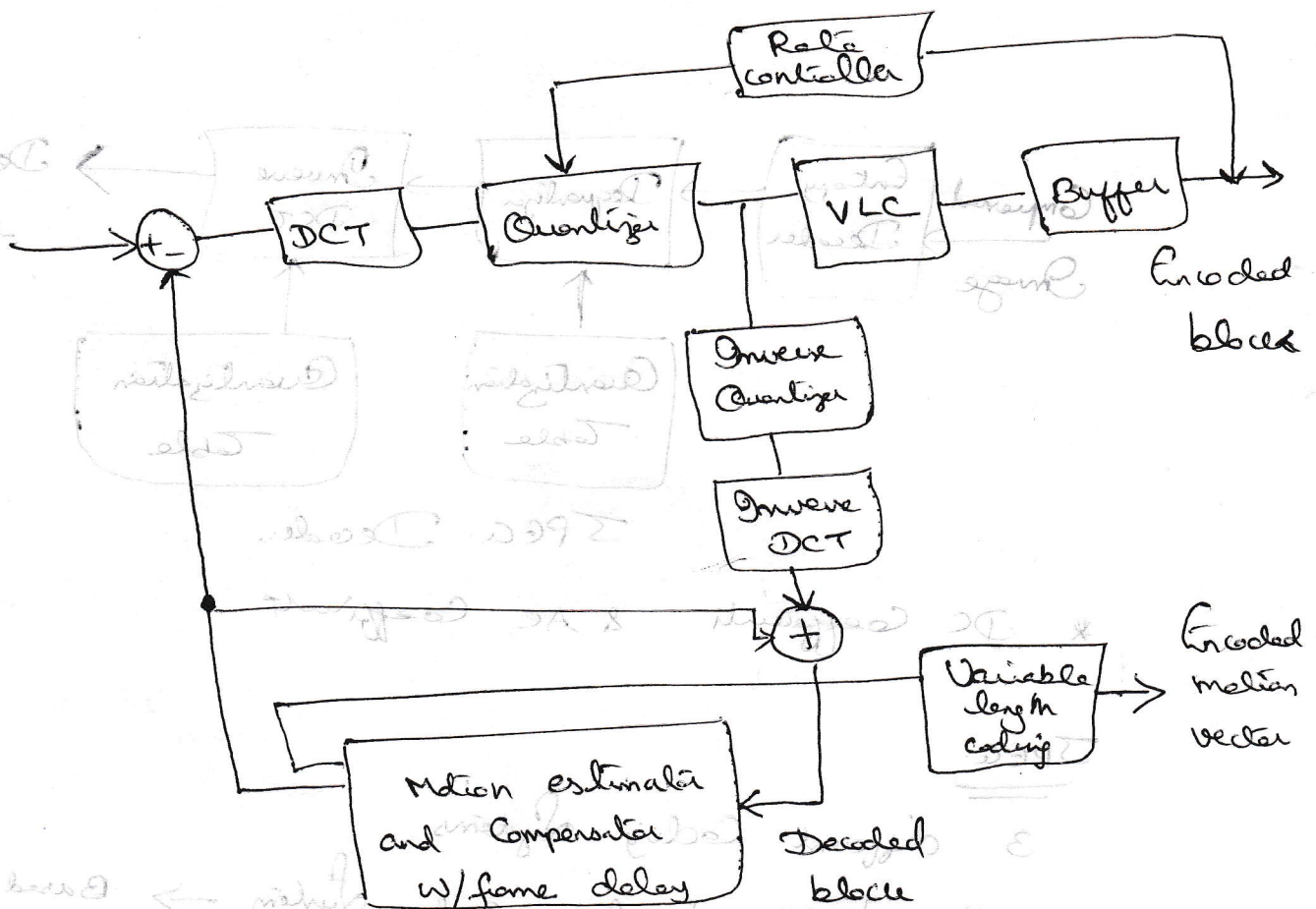


Fig 0: Basic DPCM / DCT Encoder for motion Compensated video Compression.

Basic types of encoded o/p frames

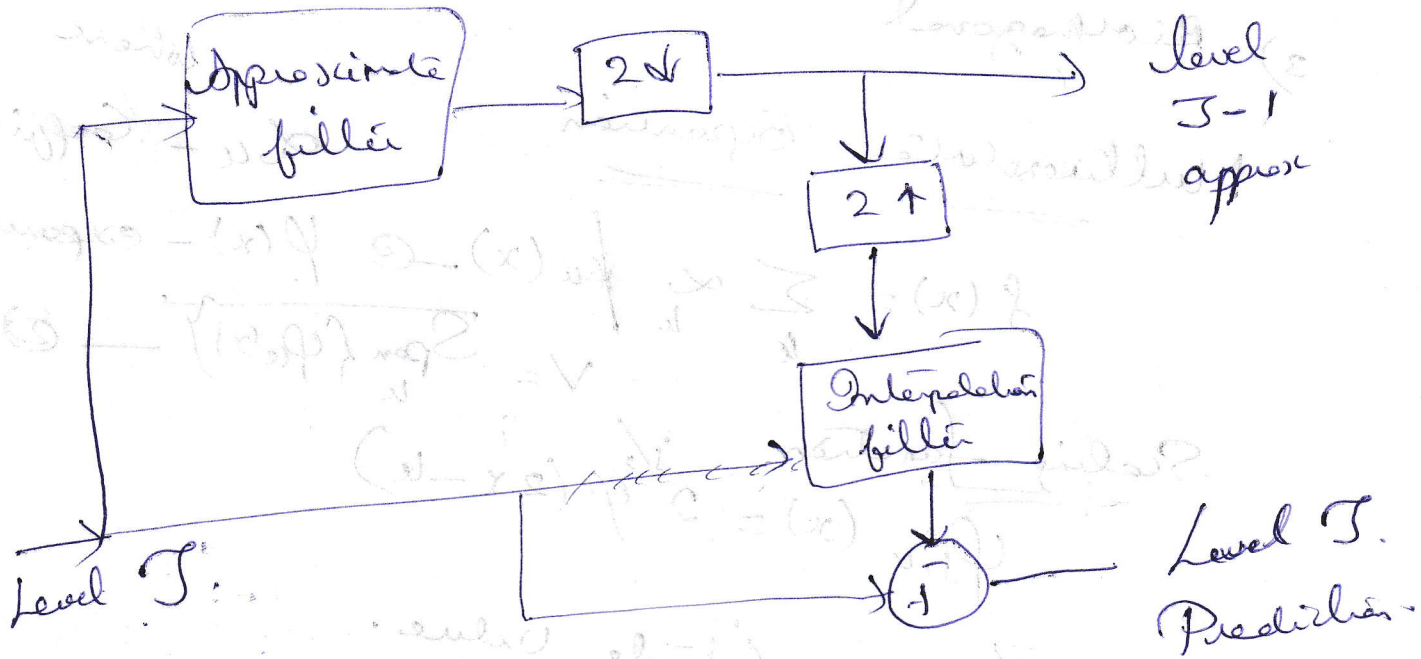
- i) Intra frame (or) independent frame (I-frame)
- ii) Predictive frame (P-frame)
- iii) Bi directional frame (B-frame).

# Wavelets

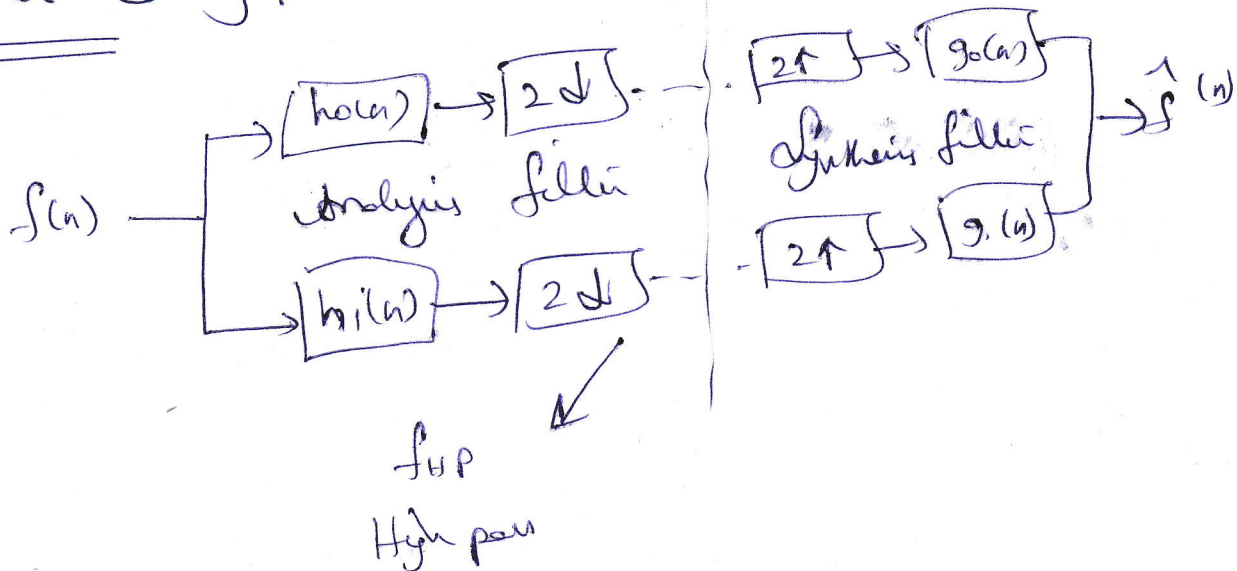
→ Image Pyramids

→ upsampling  $\Rightarrow f_2 \uparrow(n) = f(n/2)$

→ Downsampling  $\Rightarrow f_2 \downarrow(n) = f(2n)$



## Subband Coding:-



/// by for L&H for Component delivered

element,  $\rightarrow$

Conditions  $\rightarrow$   $\phi(x) = \phi(x) + d$   $\leftarrow$  polynomial  $\rightarrow$

- 1) orthogonal  $\rightarrow$   $\phi(x) = \phi(x) + d$   $\leftarrow$  polynomial  $\rightarrow$
- 2) orthogonal
- 3) Biorthogonal

Multi-resolution Expansion

where  $\alpha_k$  - Coefficient

$$f(x) = \sum_k \alpha_k \phi_k(x) \quad \text{--- ①} \quad \phi(x) - \text{expansion}$$

$$V = \text{Span} \{ \phi_k(x) \} \quad \text{--- ②}$$

Scaling function

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

where  $2^{j/2} \rightarrow$  amplitude value.  
 $j \rightarrow$  Integer.

