

Image Coding and Data Compression

Part 1 : Image Coding

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The Need for Coding and Compression

- Film technology is outdated and cumbersome.
 - Medical Images are Digital.
 - High Spatial Resolution and Fine grey level quantization are required for medical images.
 - Volumetric data obtained by CT or MRI could be of size $512 \times 512 \times 64$. At 16 bits/voxel, 32 MB of storage space is required.
 - Digital Images may be compressed via image coding and data compression techniques.
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Information Theory

- Entropy is a measure of uncertainty or randomness in an image.
- Higher entropy gives more information about the image.

- Higher Correlation between data



Higher Redundancy



Lesser Entropy



Lesser Information



Redundancy

- Code Redundancy
 - All pixel values do not occur with equal probability
 - Spatial Redundancy
 - Adjacent pixels are correlated
 - Psychovisual Redundancy
 - Precise numbers are not needed to observe important features in the image
 - Reduce redundancy through coding for better compression of images.
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Lossless vs Lossy Compression

- Lossless Coding
 - Original image can be recovered from the coded and compressed image without any loss in information.
 - Lossy Coding
 - Original data cannot be recovered with complete numerical accuracy from the compressed image.
 - Numerically lossy coding may be perceptually or diagnostically lossless.
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Distortion Measure

- Let us define a error Image as

$$e(m,n) = g(m,n) - f(m,n)$$

- RMS value of the error

$$\sqrt{\frac{1}{N^2} \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} [g(m,n) - f(m,n)]^2}$$

- SNR is defined as

$$\frac{\sum_{m=0}^{N-1} \sum_{n=0}^{N-1} g^2(m,n)}{\sum_{m=0}^{N-1} \sum_{n=0}^{N-1} e^2(m,n)}$$

Fundamental Concepts of Coding

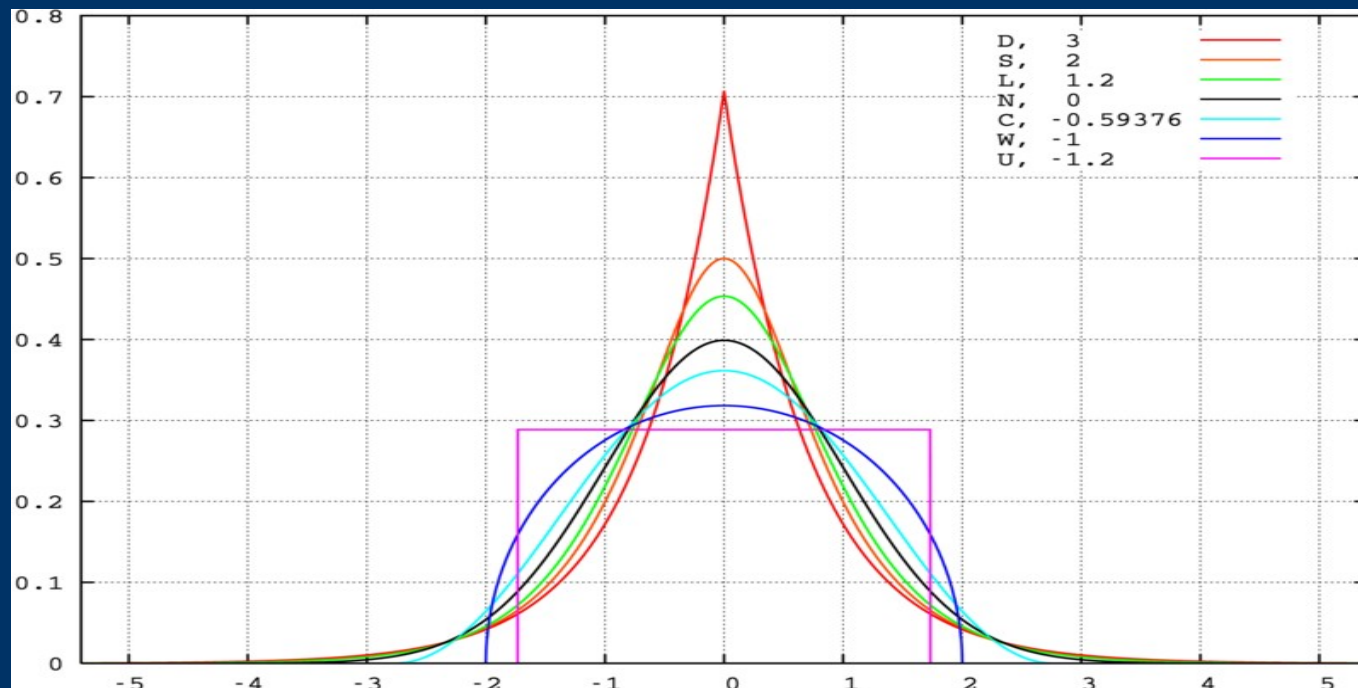
- Alphabet is a set of pre-defined symbols.
 - Word is a finite sequence of symbols from an alphabet.
 - Code is a mapping of words from a source alphabet into words of code alphabet.
 - A code is said to be distinct if each code word is distinguishable from other code words.
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Fundamental Concepts of Coding

- A distinct code is uniquely decodable if every code is identifiable in a sequence of code words.
 - A uniquely decodable code must be decodable on a word to word basis.
 - If no code word is a prefix of another, then it is also instantaneously decodable.
 - A code is optimal if it is instantaneously decodable and has minimum average length for a given pdf.
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Direct Source Coding

- Coding method is directly applied to the pixel values generated by the source.
- Patterns of limited variation and high correlation
- Usually images have non-uniform pdf.



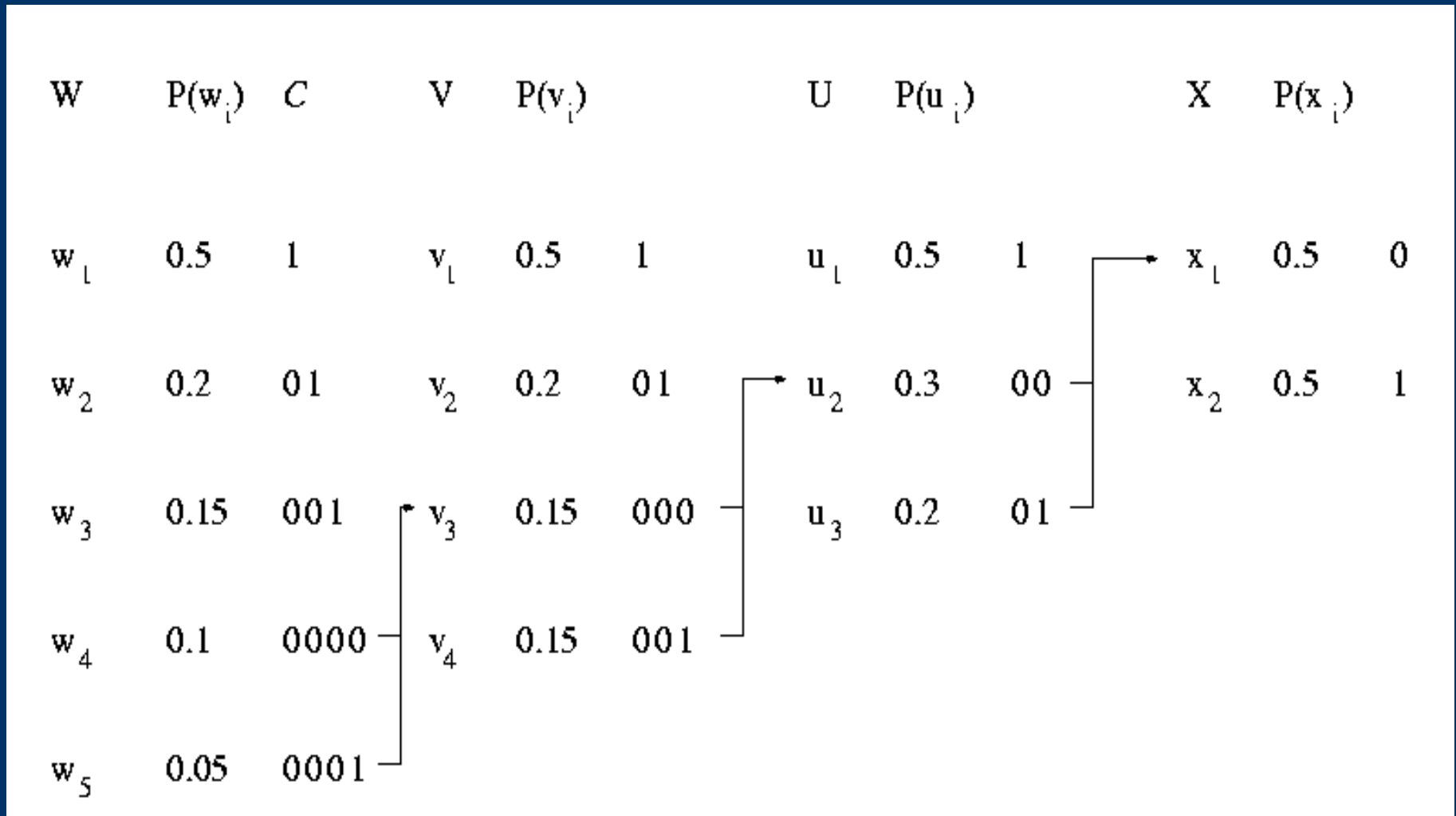
Huffman Coding

- Some pixel values have higher probabilities than other pixel values.
 - Use short code words for pixels with high probabilities of occurrence.
 - Code words are variable in length.
 - Average code word lengths are lesser than that provided by fixed length codes.
 - Codes are uniquely and instantaneously decodable.
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Huffman Coding

- Prior knowledge of pdf is assumed.
 - Huffman code is optimal only for a given source pdf and has to be redesigned for changes in pdf.
 - Huffman coding yields best results for highly non-uniform or concentrated pdf.
 - Without prior decorrelation, the code word length is increasing for sources with several symbols.
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Huffman Coding - Example



Run-length Coding

- Images with high levels of correlation contain strings of repeated values of the same grey level.

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|---|---|---|---|---|---|---|---|---|---|---|---|---|---|---|---|
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 2 | 3 | 2 | 2 | 1 | 2 |
| 0 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 2 | 2 | 3 | 4 | 5 |
| 1 | 0 | 0 | 0 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 2 | 2 | 4 | 6 |

- Run-length coding can be done as follows

Row 1: (1,10),(2,1),(3,1),(2,2),(1,1),(2,1)

Row 2: (0,1),(1,10),(2,2),(3,1),(4,1),(5,1)

Row 3: (1,1),(0,3),(1,8),(2,2),(4,1),(6,1)



Run-length Coding

- Best suited for bi-level images.
 - Images with fine details, intricate texture and high resolution quantization with large numbers of bits per pixel may actually lead to data expansion.
 - Can be advantageously applied to bit planes of grey level and colour images.
 - Errors in run length can cause severe degradation of the reconstructed image due to loss of pixel position.
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Arithmetic Coding

- Represent a string of input symbols by their individual p_l and cumulative probabilities P_l .
 - Source string is represented by code point C_k and an interval A_k .
 - A new symbol is encoded as follows $A_{k+1} = A_k p_l$ and the new code point is defined as $C_{k+1} = C_k + A_k P_l$
 - Each symbol need not have a unique code word as in Huffman coding.
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Lempel-Ziv Coding

- Probabilities of source symbols not known a priori.
 - Pass symbols from source into sub strings or words.
 - Map the sub strings of variable length into uniquely decipherable codes of fixed length.
 - Lempel- Ziv coding may be viewed as a search through a fixed size variable content dictionary for words that match the current string.
 - Lempel-Ziv-Welch (LZW) Coding.
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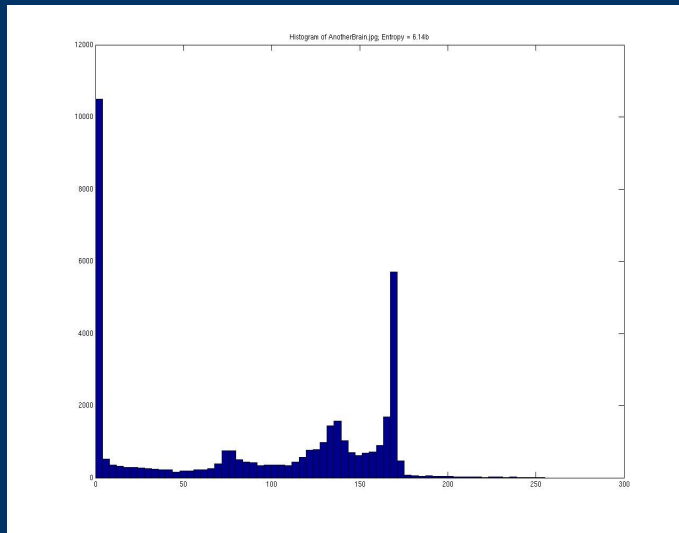
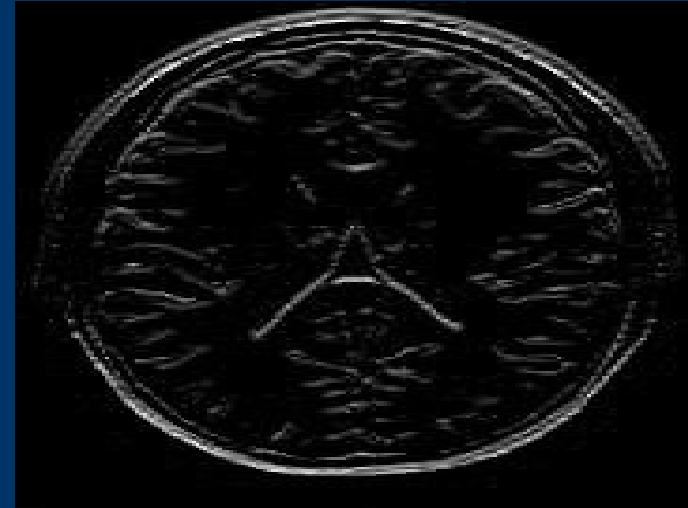
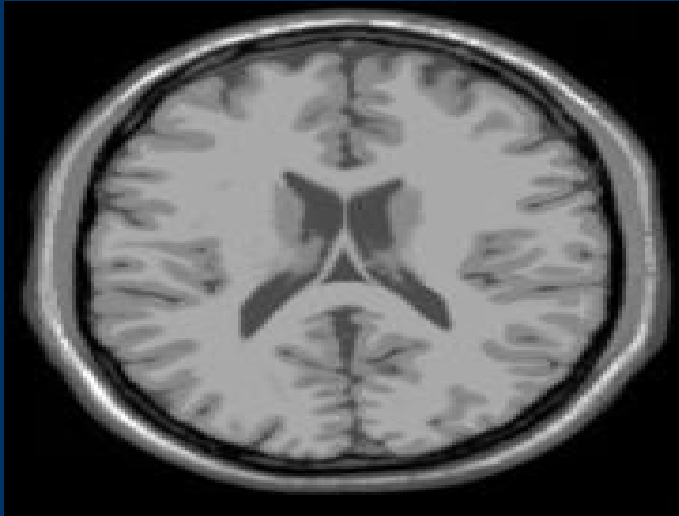
Contour Coding

- Pixels of same grey levels occur in 2D contours or patterns in the image.
 - Information related to all such contours may be used to encode the image.
 - For each contour, encode co-ordinates of starting point, grey levels and sequence of steps needed to trace the contour.
 - A consistent rule is required for repeatable tracing of contours.
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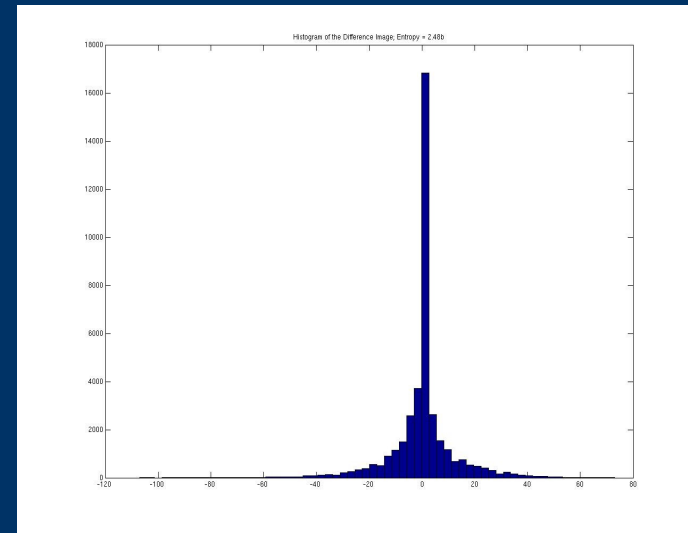
Decorrelation

- Decorrelation is a procedure to remove or reduce redundancy or correlation between pixels.
 - Commonly used decorrelation techniques are
 - Differentiation
 - Transformation to another domain
 - Model based prediction
 - Interpolation
 - Decorrelated data needs to be encoded and transmitted.
 - Coding requirements are significantly reduced.
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Decorrelation



Entropy = 6.14b



Entropy = 2.48b

Transform Coding

- Orthogonal Transforms - Rotation of co-ordinate system in signal space.
 - Purpose of the transform – Decorrelation and Energy Concentration.
 - We have already seen
 - DFT
 - WHT
 - We will now see
 - Discrete Cosine Transform (DCT)
 - Karhunen Loeve Transform (KLT)
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DCT

- DCT is a modification of DFT

- $$F(k, l) = \frac{a(k, l)}{N} \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} f(m, n) \cos \left[\frac{\pi m}{N} (2k+1) \right] \cos \left[\frac{\pi n}{N} (2l+1) \right]$$

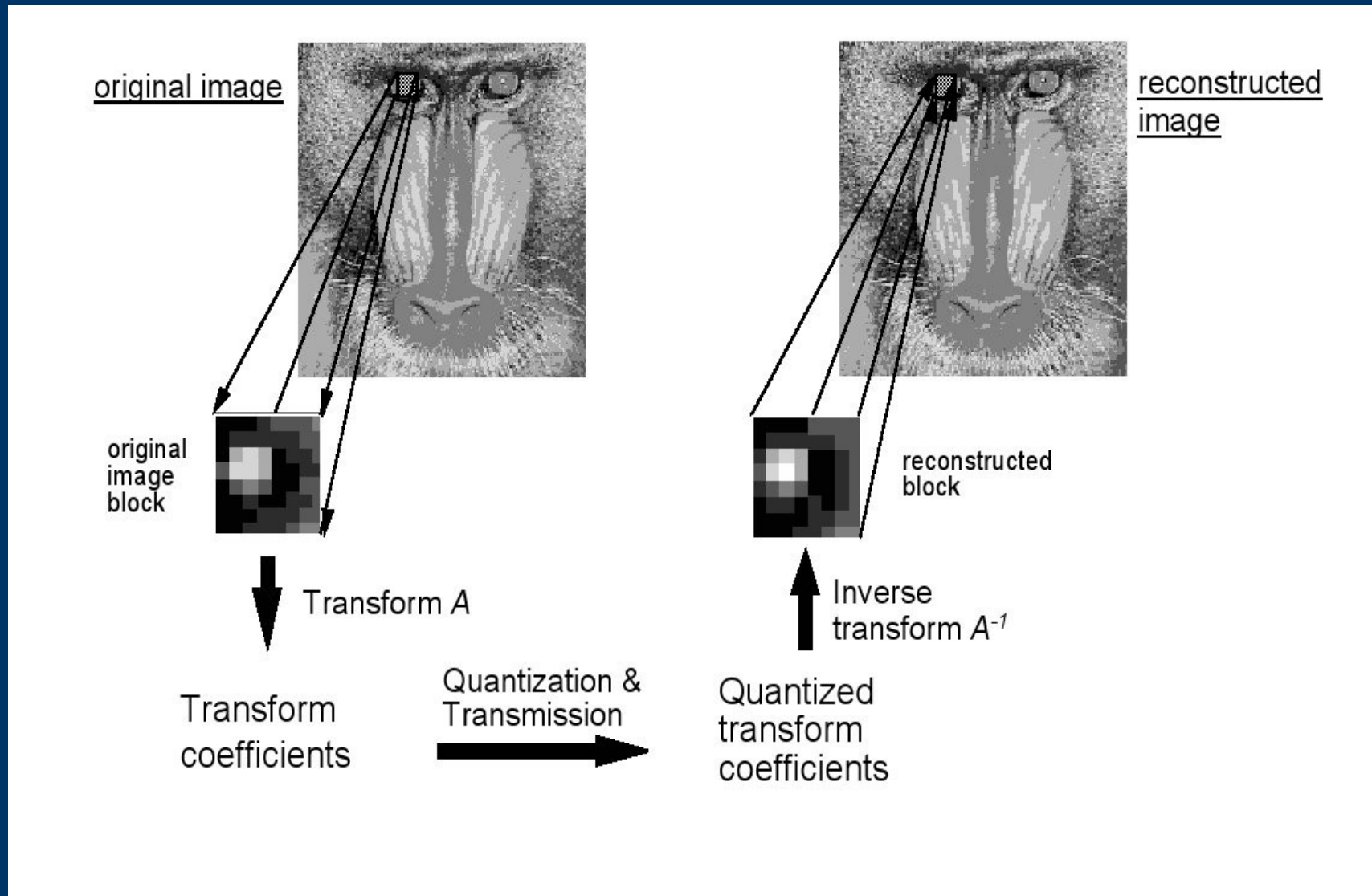
$$\begin{aligned} k &= 0, 1, 2, \dots, N-1 \\ l &= 0, 1, 2, \dots, N-1 \end{aligned} \quad a(k, l) = \begin{cases} 1 & \text{if } (k, l) = (0, 0) \\ \frac{1}{2} & \text{otherwise} \end{cases}$$

- Inverse DCT is given by

$$f(m, n) = \frac{1}{N} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} a(k, l) F(k, l) \cos \left[\frac{\pi m}{N} (2k+1) \right] \cos \left[\frac{\pi n}{N} (2l+1) \right]$$

$$m = 0, 1, 2, \dots, N-1 \quad n = 0, 1, 2, \dots, N-1$$

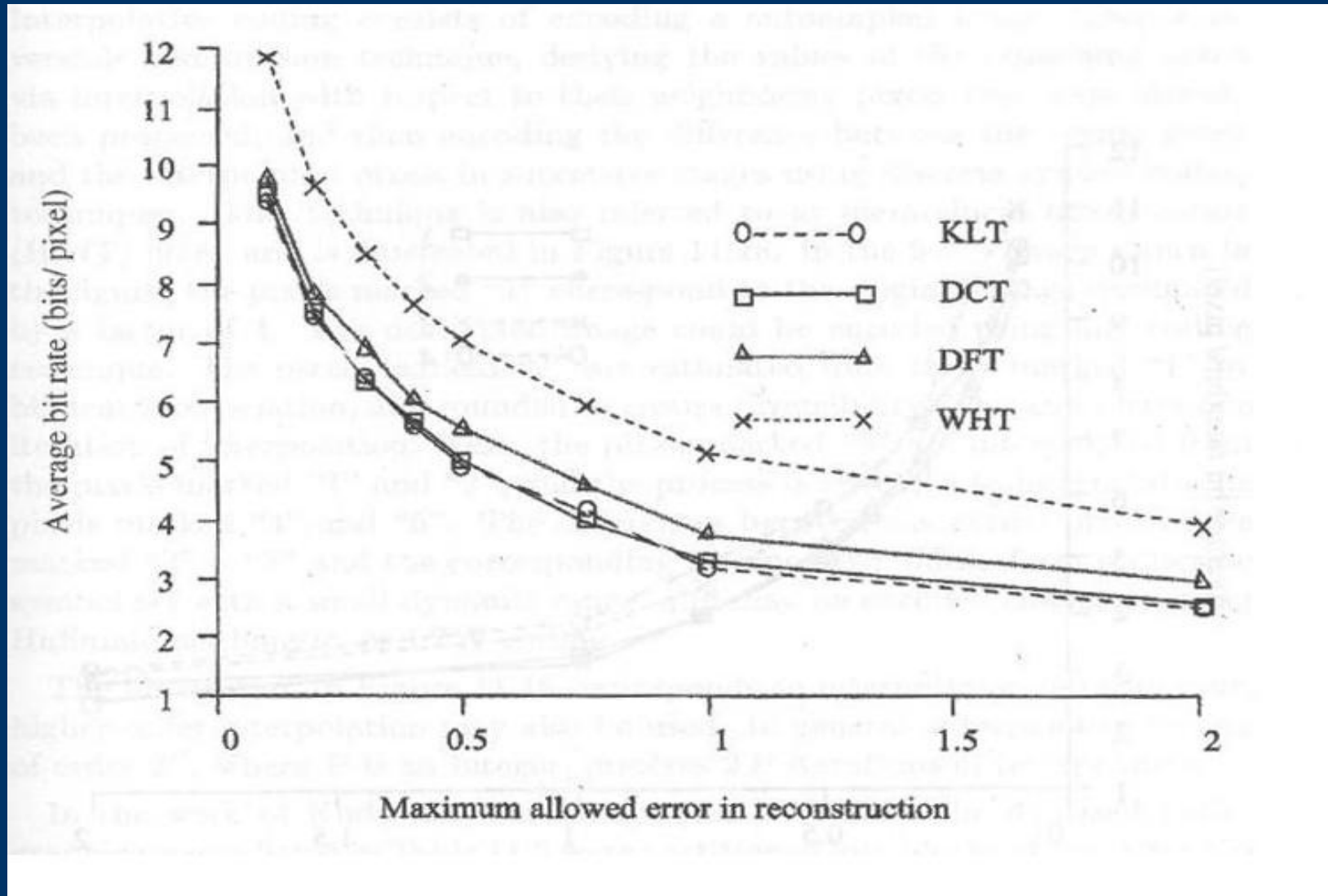
DCT



KLT

- Principal Component Transform or Hotelling Transform or Eigenvector Transform.
 - Based on statistical properties of the image.
 - KLT yields decorrelated transform coefficients (Covariance matrix is diagonal).
 - Basis functions are the eigenvectors of the input covariance matrix.
 - KLT achieves optimum energy concentration.
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Comparison of Transforms



Encoding Transform Coefficients

- Transform coefficients are quantized for encoding irrespective of the transform used.
 - Quantization errors are introduced.
 - Maximum error limit must be derived so that pixels exceeding this limit are encoded separately.
 - Use variable length encoding and bit allocation depending on pdf of transform coefficients
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Interpolative Coding

- Sub sample the image (Decimation).
 - Interpolate the remaining pixels through some suitable interpolation techniques.
 - Compute the differences between the actual and interpolated pixels.
 - Encode the difference using discrete symbol coding techniques.
 - Differences can modelled using Laplacian pdf.
 - Interpolative coding of order 2^P
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Predictive Coding

- Correlation between adjacent samples spatially or temporally – statistical redundancy.
- Sample of a signal may be predicted from previous samples.
- Linear Prediction Model for Images

$$\tilde{f}(m, n) = - \sum_p \sum_q a(p, q) f(m-p, n-q)$$

where $\mathbf{a(p,q)}$ are the 2D LP model coefficients.

- Prediction Error and MSE can be calculated as

$$e(m, n) = f(m, n) - \tilde{f}(m, n) \quad \epsilon^2 = E[e^2(m, n)]$$

Predictive Coding

- Error image $e(m,n)$ has more concentrated pdf which leads to better compression.
- Derive Prediction Coefficients – 2D normal or Yule-Walker equations.

- $$\phi_f(r, s) + \sum_p \sum_q a(p, q) \phi_f(r-p, s-q) = \begin{cases} 0 & (r, s) \in ROS \\ \epsilon^2 & (r, s) = (0, 0) \end{cases}$$

where ϕ_f is the ACF and **ROS** is the region of support.

- Prediction error can be encoded using any discrete symbol coding method.
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Predictive Coding

- Multichannel Linear Prediction is a combination of 1D and 2D Linear Prediction.
 - Certain number of rows of the image can be seen as a collection of multichannel signals.
 - Prediction coefficient matrices can be calculated using Levinson-Wiggins-Robinson algorithm or directly from the image through Burg algorithm.
 - For error free reconstruction at the decoder, the prediction coefficients have to be recomputed.
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Predictive Coding

- Adaptive 2D recursive Least Squares Prediction.
 - LP model with constant prediction coefficients assume stationarity of image generating processes.
 - Assumption is hardly valid for real world images.
 - To overcome this problem,
 - Block computation of prediction coefficients
 - Adapt coefficients recursively to changing statistical characteristics of the image
 - Minimize the weighted sum of prediction errors in the least squares sense.
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Quick Recap

- Lossless data compression is desirable in medical images.
 - Information theoretic considerations.
 - Direct Source Coding
 - Transform Coding
 - Predictive Coding
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Thank You

Any Questions..??

