

# Image Coding and Data Compression

Part 2 : Data Compression

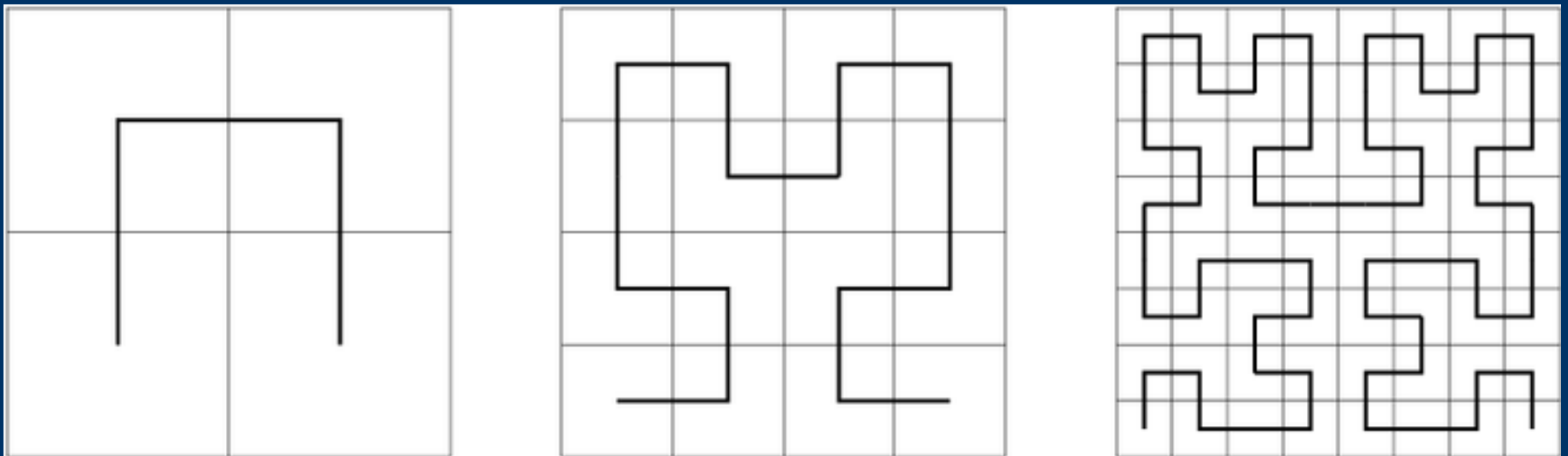
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# Image Scanning

- Space Filling Curve - map a line on to a 2D space
- Peano-Hilbert Curve
- Peano scanning



# Peano-Hilbert Curve

- 2D local statistics are preserved in 1D data stream.
  - Higher correlation between pixels in stream.
  - Scanning and reconstruction is lossless.
  - Improved Image data compression.
  - Continuous and not differentiable.
  - Does not pass through any point more than once.
  - Recursive nature, easy to implement.
  - Image dimensions must be  $2^n \times 2^n$
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# Image Compression Standards

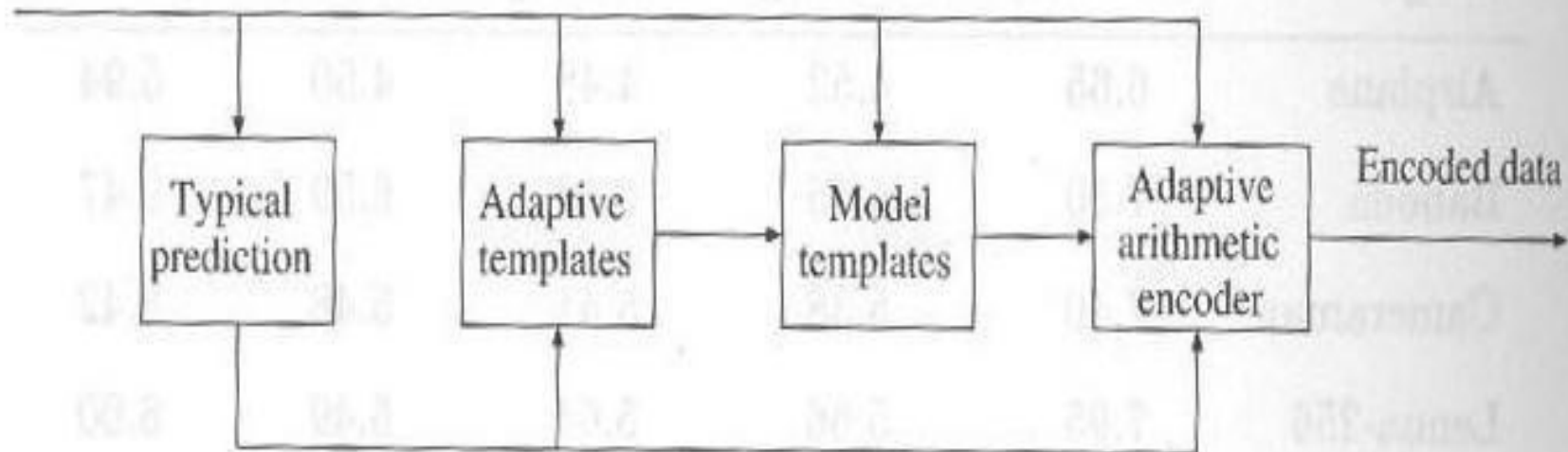
- JBIG – Joint Bi-level Image Experts Group
  - JPEG – Joint Photographic Experts Group
  - MPEG – Moving Pictures Experts Group
  - ACR/NEMA – American College of Radiology/  
National Electrical Manufacturers Association.
  - DICOM – Digital Imaging and Communications  
in Medicine
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# JBIG

- Standard for progressive coding of bi-level images
    - Progressive coding
    - Sequential coding
    - Single-layer coding
      - Prediction step
      - Adaptive template
      - Model template
      - Adaptive arithmetic encoder
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# JBIG

Input bilevel image or bit plane



- 3 levels of decorrelation
- 1D JBIG coding uses run length coding

# Enhanced JBIG

- PSV = 7

$$\tilde{f}(m, n) = \frac{f(m-1, n) + f(m, n-1)}{2}$$

- F1 Transform

$$v_1 = F_1(v) = \begin{cases} 0; & v=0 \\ 2v-1; & v \leq 2^{K-1} \\ 2(2^K - v); & v > 2^{K-1} \end{cases}$$

- Enhanced JBIG - PSV7-F1-JBIG
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# JPEG

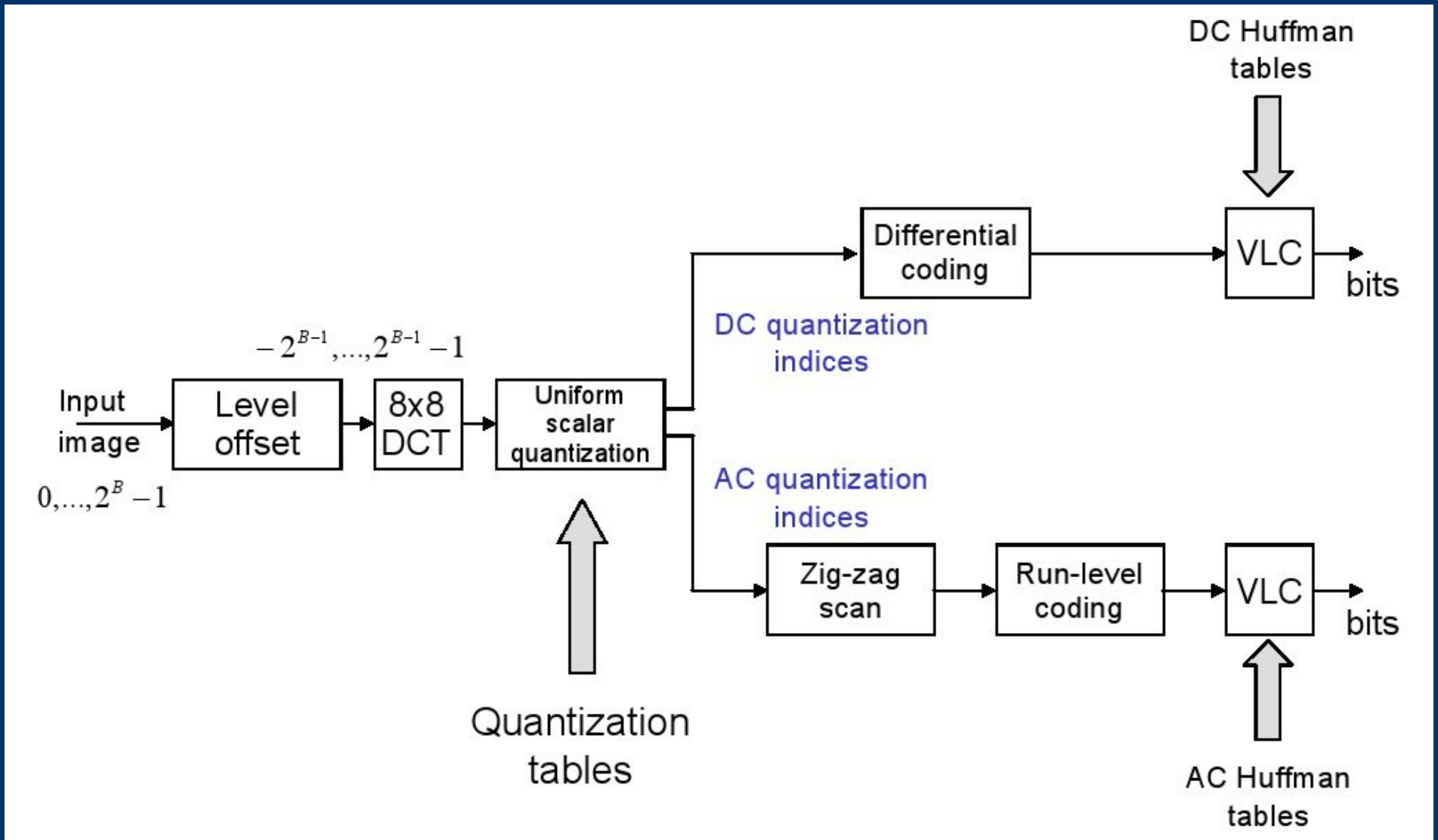
- Flexible standard for monochrome and color image compression.
  - Digital compression and coding of continuous-tone still Images.
  - Work started in mid-1980's
  - Draft international standard 1991
  - Intra-frame coding scheme, optimized for still images.
  - Coding of color components separately, arbitrary color spaces possible, best compression for YCbCr.
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# JPEG

- Variable compression ratio
  - Compression ratio up to 24:1 for ITU-R 601 images without loss of quality
  - Widely used for image exchange, WWW, and digital photography
  - Supports both lossless and lossy coding
    - Baseline coding – Block-wise DCT
    - Extended coding
    - Lossless independent coding
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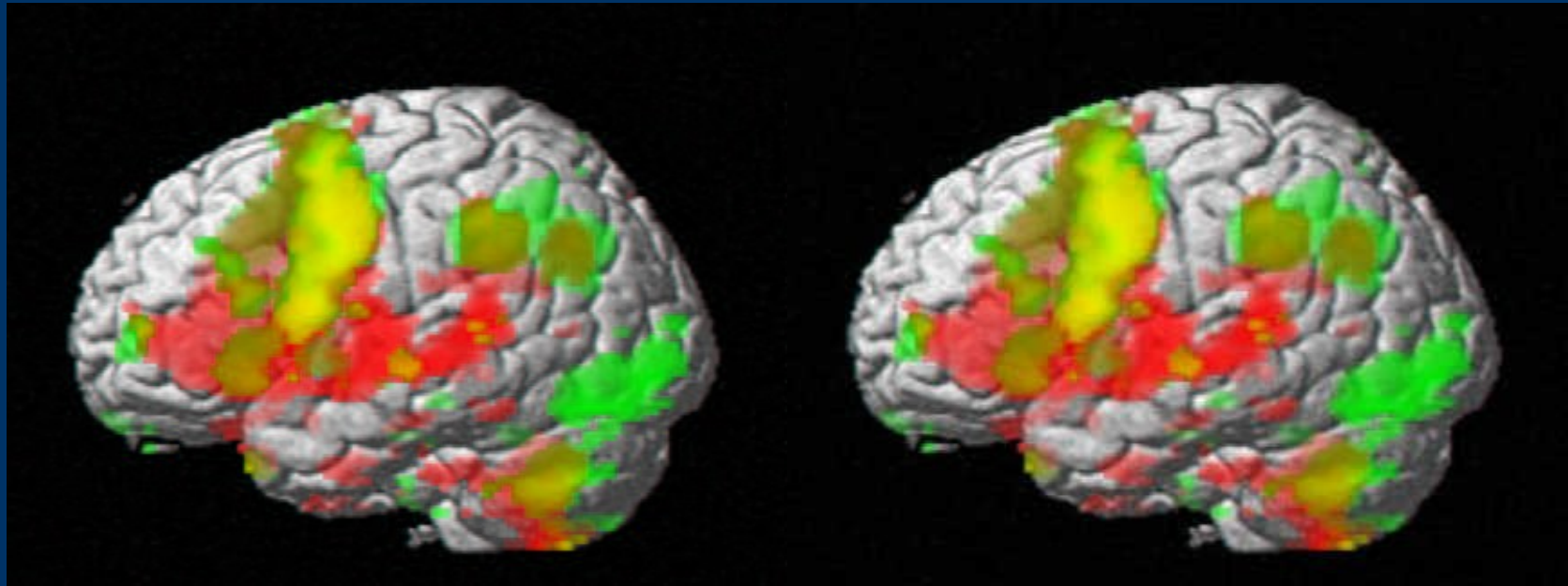
# JPEG





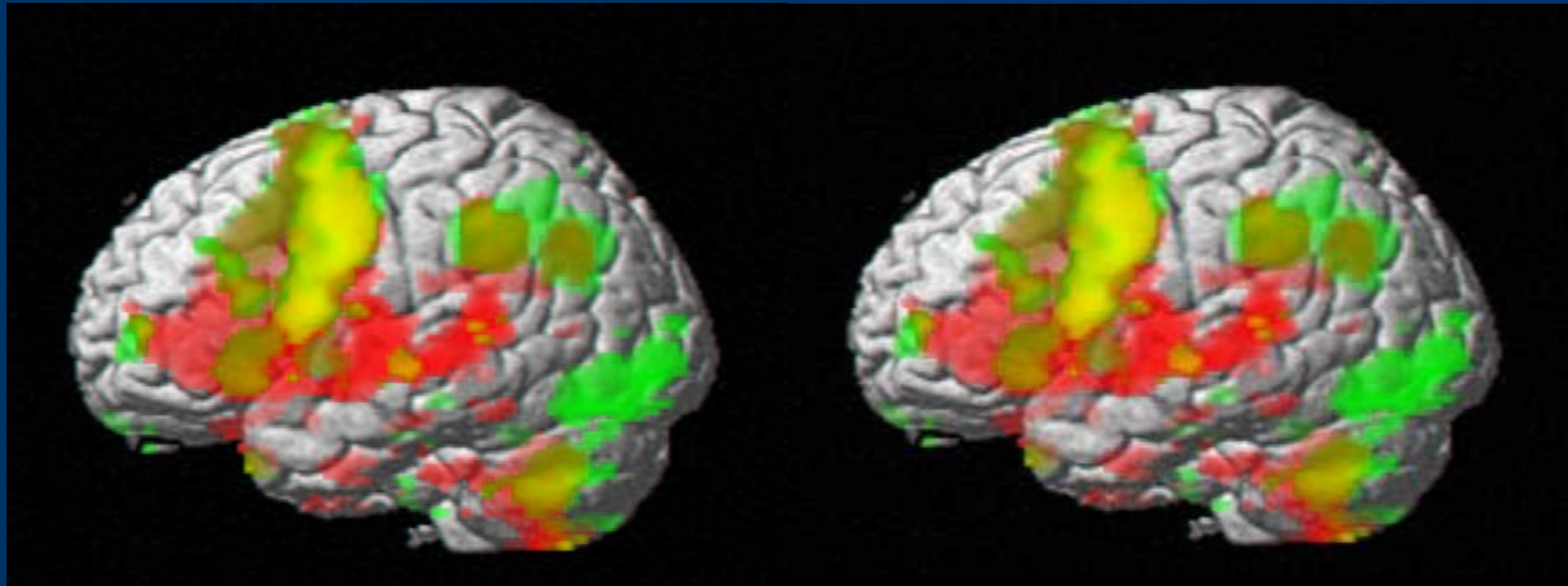


# Illustrations - Baseline JPEG



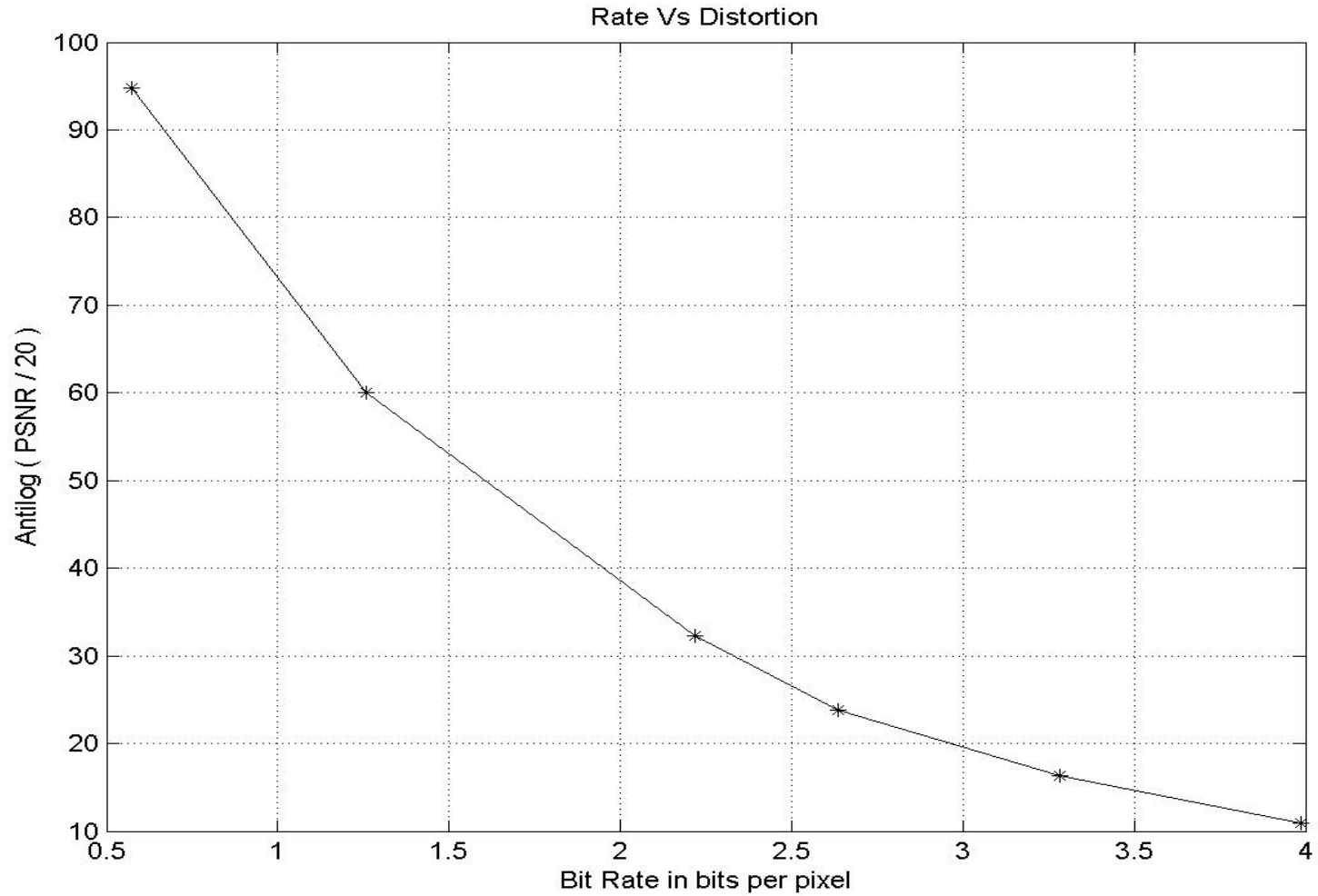
- Original Image :  $368 \times 432 \times 3 = 466$  KB (8 bpp)
  - Transmitted Stream = 21,995 bytes = 21.5 KB
  - Compression Ratio = 22:1
  - PSNR = 35.299 dB , MSE = 19.195
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# Illustrations - Baseline JPEG



- Original Image :  $368 \times 432 \times 3 = 466$  KB (8 bpp)
  - Transmitted Stream = 15,300 bytes = 15 KB
  - Compression Ratio = 31:1, Subsampled Chrominance
  - PSNR = 32.525 dB , MSE = 36.354
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# Illustrations - Baseline JPEG



# Quote

- Perfection is reached,  
not when there is no longer anything to add,  
but when there is no longer anything to take away.  
– Antoine de Saint-Exupery



# Segmentation-based Adaptive Scanning

- Segmentation-based Lossless Image Coding – SLIC
  - Segment the image into nearly homogeneous regions surrounded by contours representing individual objects in the image – Region Growing Approach
  - Encode contour and texture information and discontinuity map and error image separately using gray codes.
  - JBIG Encoding
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# Lower-limit Analysis of Lossless Data Compression

- Lowest limit of bit rate in lossless compression can not be determined practically.
  - Difficult to judge various compression algorithms.
  - Zeroth-order entropy is a useful metric.
  - Higher order entropy values can provide a better estimate to the lower bound bit rate.
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# Memoryless Entropy

- Memoryless source
- Successive pixels are statistically independent
- $M^{\text{th}}$  order entropy

$$H_m(A) = H_m(a_{i_0}, a_{i_1}, a_{i_2}, \dots, a_{i_m})$$

$$= - \sum_{A^m} p(a_{i_0}, a_{i_1}, a_{i_2}, \dots, a_{i_m}) \log_2 p(a_{i_0}, a_{i_1}, a_{i_2}, \dots, a_{i_m})$$

- $H(A) = \frac{H_m(A)}{m+1}$

# Markov Entropy

- Successive pixels are significantly interdependent even after decorrelation
- Source has memory
- Can be modeled as a Markov source

$$H(a_{i_0} / a_{i_1}, a_{i_2}, \dots, a_{i_m}) = - \sum_{A^m} p(a_{i_0}, a_{i_1}, a_{i_2}, \dots, a_{i_m}) \times \log_2 p(a_{i_0}, a_{i_1}, a_{i_2}, \dots, a_{i_m})$$

- $H(a_{i_0} / a_{i_1}, a_{i_2}, \dots, a_{i_m}) \leq H(A)$
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# Estimation of true source entropy

- Order of the model and conditional pdfs are unknown.
  - Higher the order of the conditional probability  $\rightarrow$  Lower resulting entropy  $\rightarrow$  Closer to the true source entropy.
  - Estimate conditional probabilities for an  $m^{\text{th}}$ -order Markov source model with  $2^K$  intensity levels and  $N$  data samples
  - Estimation error  $\epsilon = \frac{2^{mK} - 1}{2N \ln 2}$
  - For a given  $\epsilon$ , derive a higher order parameter by reducing  $K$  - splitting the data bits.
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# Quick Recap

- Piano-Hilbert Curve
  - Image Compression Standards
    - JBIG
    - JPEG
      - Baseline Encoding
      - Rate-Distortion Curve
  - Segmentation based adaptive scanning
  - Lower-limit of data compression
  - Estimation of true source entropy
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# Thank You

Any Questions..?? \*

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\* If there are no questions today, then be prepared to answer some very difficult questions next week..!!