

IMAGE COMPRESSION AND PACKET VIDEO

DC Image



Washington DC Image  
Scaled 10X

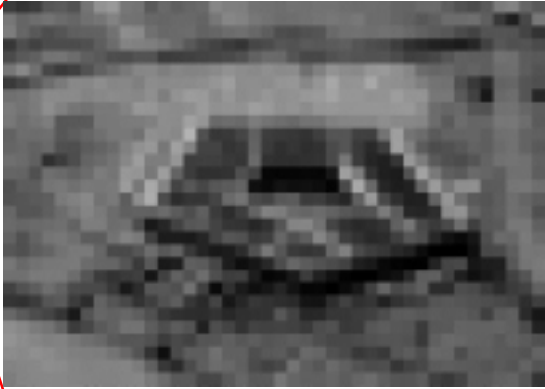
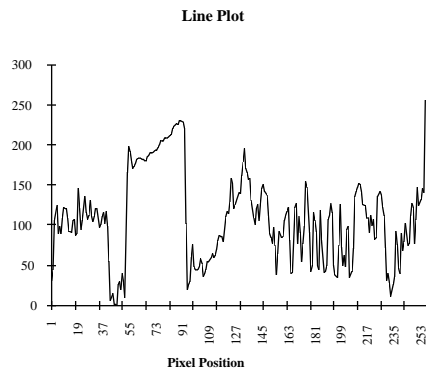


IMAGE COMPRESSION AND PACKET VIDEO

Data Profile

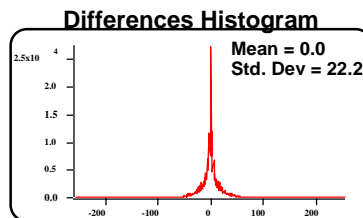
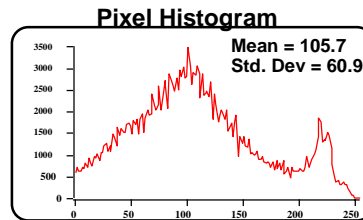


IMAGE: DC.img



## Difference Image Statistics

- Look at the histogram of the difference image



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## Histogram Statistics

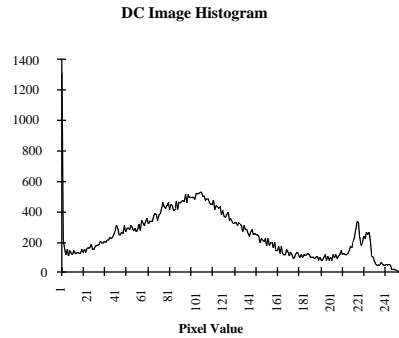
- The histogram of the difference signal is highly peaked at origin which can be exploited by the Huffman coder.
- If we look at the histogram of a typical picture, it would be roughly a gaussian distribution.
  - Difference Histogram is roughly a Laplacian distribution

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## Image Histogram



IMAGE: DC.img



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## Predictive Coding

- Two predictive coders are discussed in the literature (e.g. Anil Jain Fundamentals of Image Processing)
  - Differential Pulse Code Modulation (DPCM)
  - Delta Modulation (DM)
- Predictive Coders are among the simplest of lossless and lossy coders
  - Hardware requirements are minimal and have been built using older technology.

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## Predictive Coding

- Basic idea: Predictive coding exploits the spatial correlation between adjacent pels (previously coded information) to make an approximate prediction of the sample to be encoded.
  - The differential signal resulting from the subtraction of the prediction from the actual value of the pel is quantized into a set of L discrete amplitude levels (lossy mode).
  - These levels could then be Huffman encoded.

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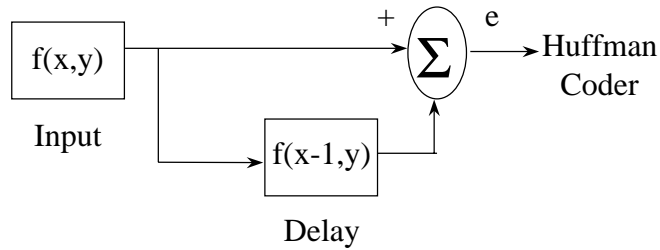
## Predictive Coder

- A Lossy predictive encoder consists of two required, and one optional component:
  - Predictor; linear predictors, adaptive predictors
  - Quantizer; uniform, non-uniform, adaptive quantizer
  - Entropy encoder (optional); Huffman coder, arithmetic coder, rice coder (NASA standard)

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## Differential Coder

- Consider the coder shown below:



- A simple coder is designed by obtaining the differences in gray level of adjacent pixels

## Differential Coder

- Original Pixel Values

$f(x-2,y)$     $f(x-1,y)$     $f(x,y)$    One line of a picture  
 $0 \leq f(x,y) \leq 255$

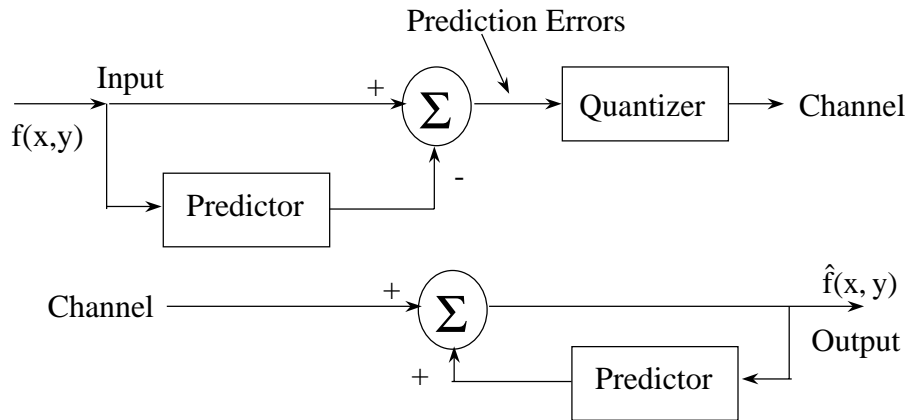
- Difference signal  $e(x,y) = f(x,y) - f(x-1,y)$

$e(x-2,y)$     $e(x-1,y)$     $e(x,y)$    The error signal  
 $-256 \leq e(x,y) \leq 255$

$$e(x,y) = f(x,y) - f(x-1,y)$$

$$e(x-1,y) = f(x-1,y) - f(x-2,y)$$

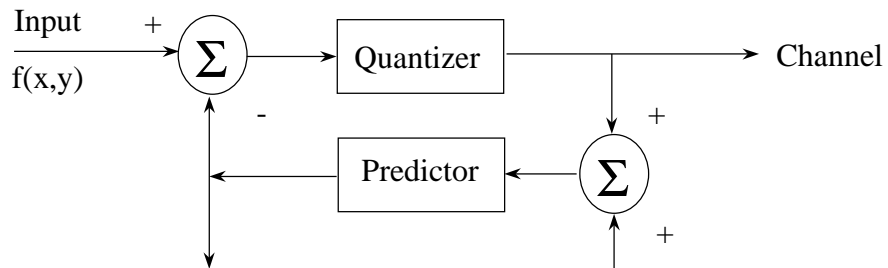
## Feed Forward Differential Coder System



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## Differential Pulse Code Modulation (DPCM)

- Problem: Quantization Error Propagation
- Solution: Quantization Error Feedback



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## DPCM Coding

- In DPCM coding, the analog signal is sampled at the Nyquist rate
  - The predictor makes a prediction using the previously transmitted pixels for the current frame (Intraframe Prediction)
  - The predictor could use pixels from the previous fields (Intraframe Prediction)
    - » Intraframe Prediction requires previous frame storage.
  - Predictors using pels from the same scan line are called 1-D predictors, from the previous scan lines are called 2-D predictors

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## Linear Predictors

- First-Order Predictor:

$$\hat{x}_i = \rho x_{i-1} + (1 - \rho)\mu$$

$\rho$  = CORRELATION COEFFICIENT

$$0.85 \leq \rho \leq 0.95$$

$\mu = E[x]$  = mean

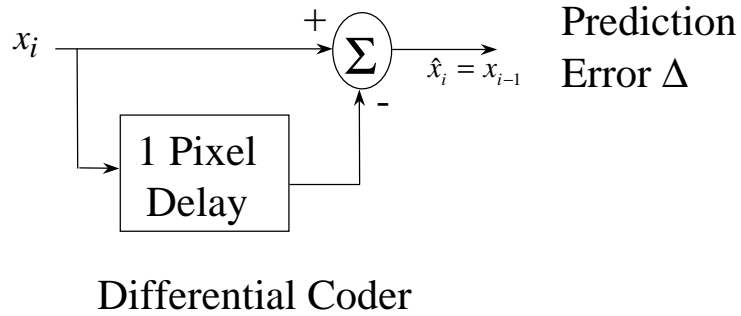
MINIMIZES  $E[(x_i - \hat{x}_i)^2]$

FOR  $\rho \approx 1 \Rightarrow \hat{x}_i = x_{i-1}$

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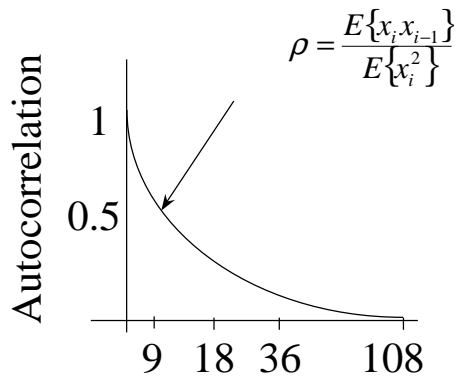
## Linear Predictor

■ Resulting Predictor:



## Linear Predictor

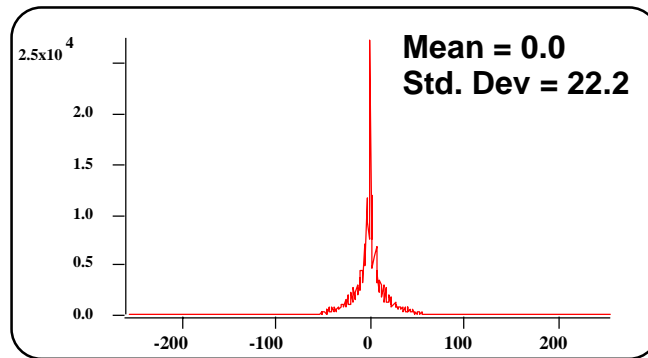
■ Correlation between the neighboring pixels



Correlation between pixels decreases as distance between them increase

## Predictor Error Probability Density

- The probability density function or histogram of prediction error is shown:



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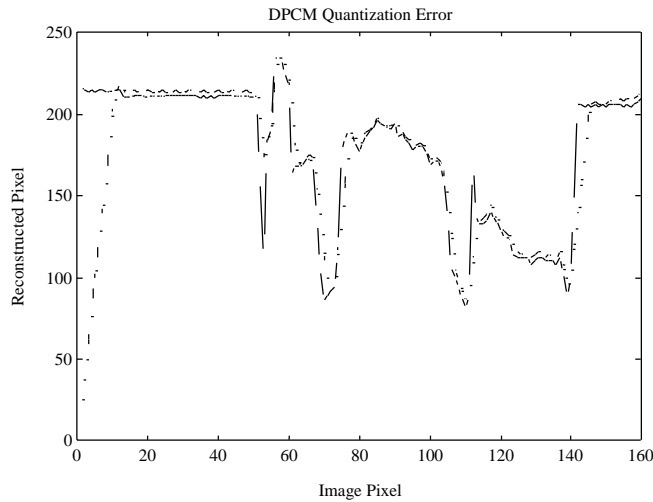
## Lossy DPCM Error

DPCM quantizer based on Lloyd-Max 3 bit table from text

Data taken from first line of image.

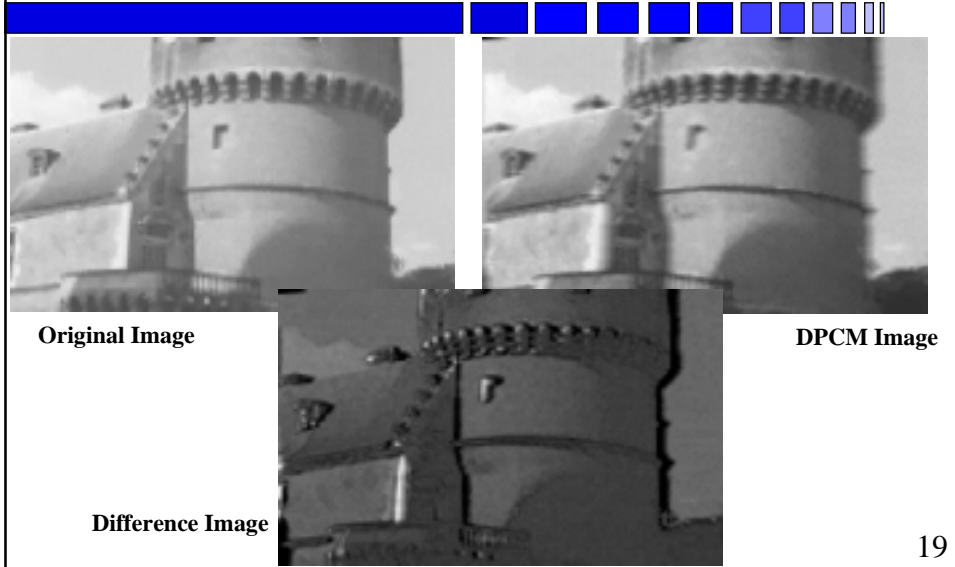
Example of:

- Slope Overload
- Granular Noise
- Edge Busyness



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## Lossy DPCM Example



## Lossy DPCM Results

- Image DPCM Lossy Compressed 1.3bpp
- Original Data 6.93 bpp
- Lossless DPCM rate: 2.67 bpp
- $rmse = 10.84$
- $mse = 117.6$

## Minimum Mean Square Predictor

- A conditional mean predictor chooses

$$\hat{b}_N = \sum_{b_N} b_N P(b_N | b_1 \dots b_{N-1})$$

which can be shown to minimize the mean square prediction error

$$E(b_N - \hat{b}_N)^2$$

where E is the statistical averaging or expectation operator.

## Minimum Mean Square Predictor

- A linear predictor chooses

$$\hat{b}_N = \sum_{i=1}^{N-1} \alpha_i b_i$$

where  $\{b_i\}$  are the previous encoded pixels and  $\{\alpha_i\}$  are the weighting coefficients that are chosen according to some performance criterion.

- For example, a “Maximum Likelihood” predictor chooses  $\hat{b}_N$  to be that value of  $b_N$  that maximizes the conditional probability  $P(b_N | b_1 \dots b_{N-1})$

## Minimum Mean Square Predictor

- A minimum mean square linear predictor can easily be derived by differentiating  $E(b_N - \hat{b}_N)^2$  with respect to each  $\alpha_j$  and setting the result to zero.
- This yields

$$\frac{\partial}{\partial \alpha_j} \left\{ E \left( b_N - \sum_{i=1}^{N-1} \alpha_i b_i \right)^2 \right\} = 0 \text{ for } j = 1 \dots M-1$$

$$-2E \left[ b_N b_j - \sum_{i=1}^{N-1} \alpha_i b_i b_j \right] = 0$$

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## Minimum Mean Square Predictor

- Letting  $d_j = E(b_N b_j)$  and  $r_{ij} = E(b_i b_j)$  and using column matrices  $D, \bar{\alpha}$  and square matrix  $R$  we get

$$D - R\bar{\alpha} = 0$$

The matrix  $R$  is known as the correlation matrix, and  $r_{ij}$  is the correlation between Pels  $b_i$  and  $b_j$

- If the correlation matrix is nonsingular, then

$$\alpha = R^{-1}d$$

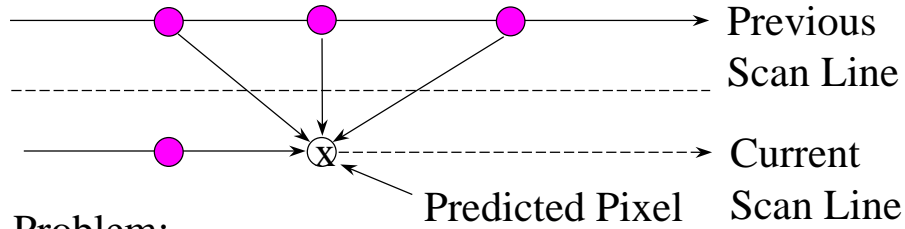
Otherwise, many solutions exist.

- Each having the same mean square prediction error:  
 $E(b_N - \hat{b}_N)^2$

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## Other Linear Predictors

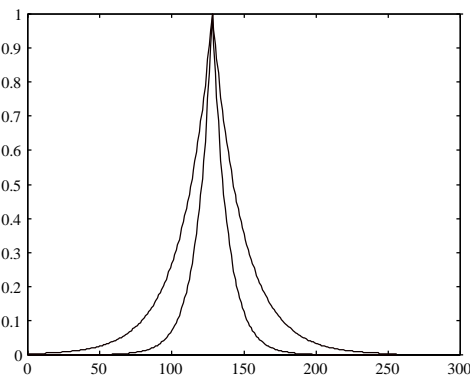
### ■ 2-Dimensional Prediction



Problem:

Interlace in most systems reduces line-to-line correlation

## Two Dimensional Predictor



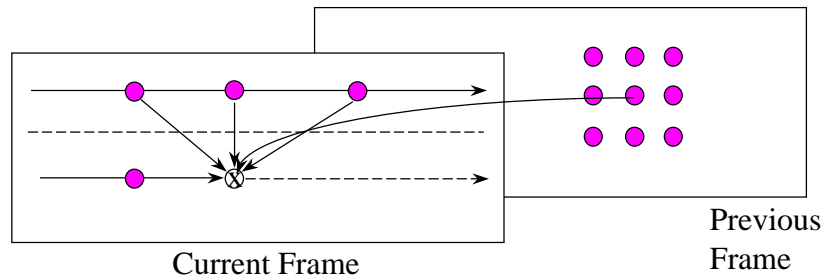
$$\hat{x}_{i,j} = \rho_1 x_{i-1,j-1} + \rho_2 x_{i-1,j} + \rho_3 x_{i-1,j+1}$$

Histogram of error signal for the one and two-D predictor

■ Two Dimensional predictor produces smaller error signals than one dimensional predictor

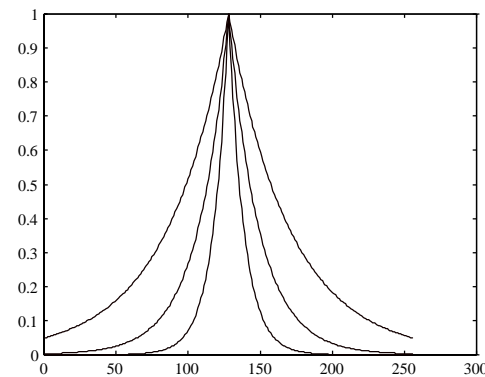
## Frame Difference Predictor

- For scenes with low detail and small motion, frame difference prediction appears to be the best.
- In scenes with higher detail and motion, intrafield prediction does better than frame prediction.



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## 3-D Prediction (Frame to Frame)



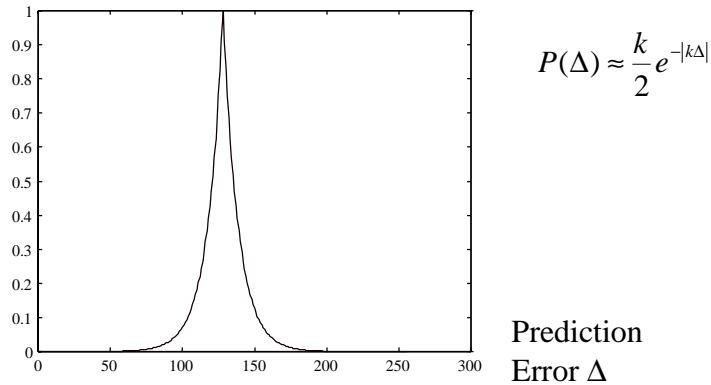
$$\hat{x}(i, j, t) = \rho_1 x(i, j, t-1) + \rho_2 x(i, j-1, t) + \rho_3 x(i-1, j, t) + \rho_4 x(i-1, j-1, t) + \rho_5 x(i-1, j+1, t)$$

- If there is no motion  $x(i, j, t) = x(i, j, t-1)$  but when there is a motion, an error signal is produced.
- Previous frame has to be stored

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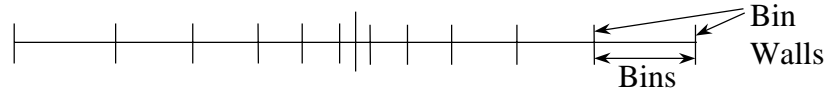
## Statistics of Prediction Errors

- Histogram of error signal approximates an exponential distribution



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## Non-Uniform Quantizer



- Small errors for highly probable small D, large errors for unlikely, large D
  - Small average quantization error
- Large errors occur near sharp edges where eye has lowered sensitivity (Masking)
- Quantizer outputs tend to be equi-probable (Permits efficient, fixed length code)
- Edge Business

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## Quantizer Design For DPCM

- Consider an 8-level quantizer, the probability of occurrence of the quantized difference signals is not uniform, small differences are more numerous than large differences.
  - The output of the 8-level quantizer is coded by a Huffman Coder.
- In the 8-level quantizer, the levels are non-uniform

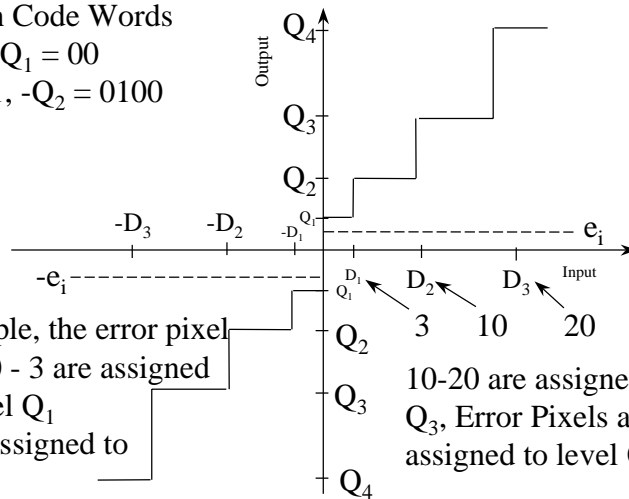
## Quantizer Design For DPCM

Huffman Code Words

$Q_1 = 1, -Q_1 = 00$

$Q_2 = 011, -Q_2 = 0100$

$Q_3 =$

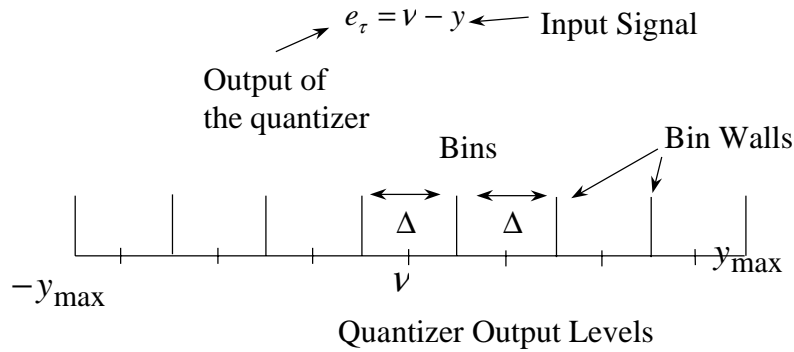


For example, the error pixel between 0 - 3 are assigned to the level  $Q_1$   
 3-10 are assigned to level  $Q_2$

10-20 are assigned to level  $Q_3$ , Error Pixels above 20 are assigned to level  $Q_4$

## Quantization Error

- The quantization error is the difference the quantizer output and input



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## Quantization Error

- Let  $\Delta = \text{bin width} = \frac{2y_{\max}}{2^R}$   
 $\therefore \max e_{\tau} = \frac{\Delta}{2}$

- The mean square error

$$e_{\tau}^2 = \int_{v-\Delta/2}^{v+\Delta/2} (y-v)^2 P(y) dy = \frac{\Delta^2}{12} = \frac{1}{3} y_{\max}^2 2^{-2R}$$

- If rms value of the input  $y$  is given by

$$y_{rms} = \sqrt{\int y^2 dy}$$

- Signal-to-Quantization noise ratio:  $Q_{SNR} = \frac{y_{rms}}{e_{\tau rms}}$   
 or in dB  $SNR = 10 \log \frac{y_{rms}}{e_{\tau}}$

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