



# *Fundamentals of Multimedia*

## Lecture 6

# Lossy Data Compression

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# Outcomes of Lecture 5

- *VariableLength Coding*
  - ◆ Shannon-Fano Algorithm
  - ◆ Huffman Coding Algorithm
- *Lossless Compression in JPEG images.*
  - ◆ Differential Coding
  - ◆ Lossless JPEG

# Outline

- *Lossy Compression*
  - ◆ Definition
  - ◆ Distortion measure
- *Quantization*
  - ◆ Uniform scale quantization
  - ◆ Nonuniform scale quantization

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# Lossy Compression

- *Lossy compression yields a much higher compression ratio*
  - ◆ Multimedia compression implementations generally are combination of lossy and lossless compression
  - ◆ The compression and decompression processes induce information loss
  - ◆ The recovered file from the compressed data is a close approximation of its original
- *Distortion measures*
  - ◆ How close an approximation is to its original



# Distortion Measures

- *Mean square error (MSE)*

$$\sigma^2 = \frac{1}{N} \sum_{n=1}^N (x_n - y_n)^2$$

- *Signal to noise ratio (SNR)*

$$SNR = 10 \log_{10} \frac{\sigma_x^2}{\sigma_d^2}$$

- *Peak signal to noise ratio (PSNR)*

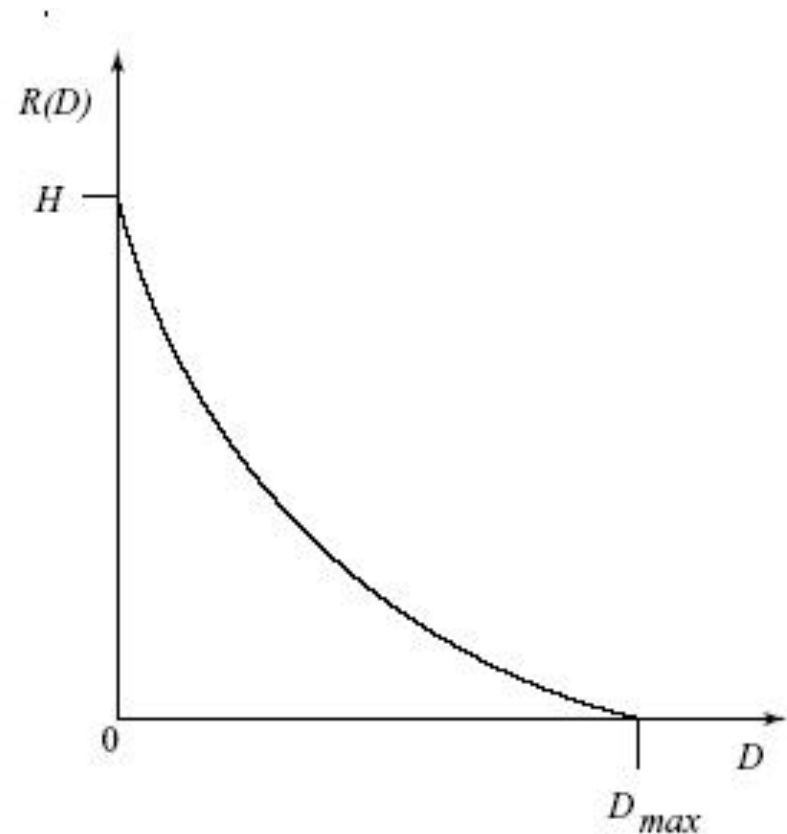
$$PSNR = 10 \log_{10} \frac{x_{peak}^2}{\sigma_d^2}$$

# Example of Distortion Measures

- *Example data*
  - ◆ Original data: {12 16 16 12 12 8 8 12 }
  - ◆ Compressed data: {8 12}
  - ◆ Recovered data:{12 12 12 12 12 12 12 12}
- *Calculation of distortion measures*
  - ◆ MSE= 8
  - ◆ SNR=12.78
  - ◆ PSNR=15
  - ◆ Typical values for the PSNR in lossy image and video compression are between **30 and 50**, where higher is better.

# The Rate Distortion Theory

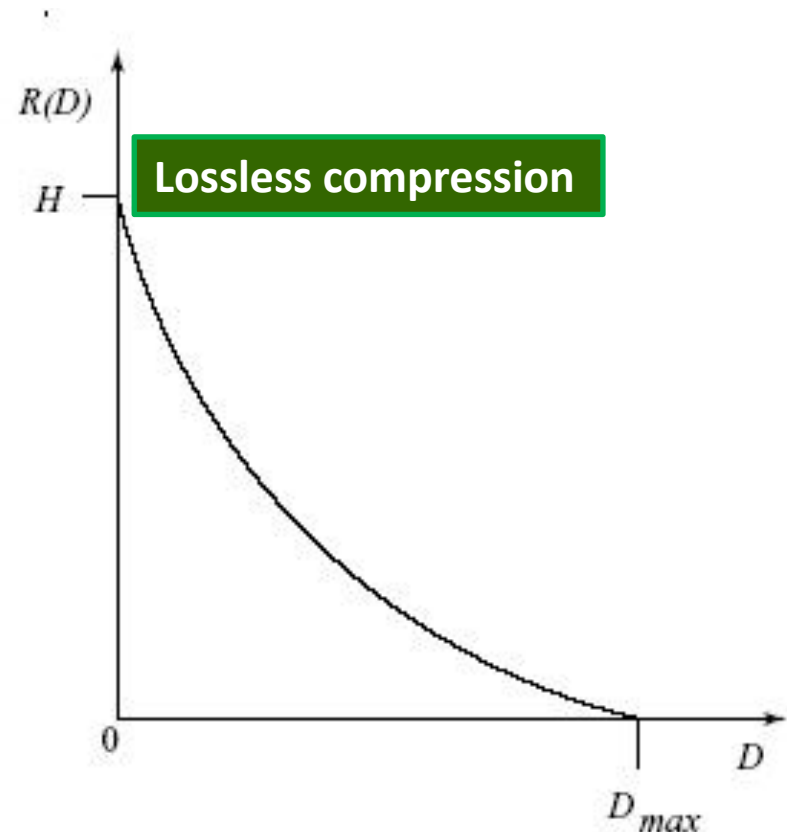
- *Lossy compression always involves a tradeoff between Rate and Distortion*
  - ◆ **Rate:** Average number of bits required to represent each symbol
  - ◆ If  $D$  is a tolerable amount of distortion,  $R(D)$  specifies the lowest rate at which the source data can be encoded while keeping the distortion bounded above by  $D$ .





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**Max distortion with nothing is coded**

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  - ◆ Uniform scale quantization
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# Quantization

- *Quantization is the heart of any lossy compression scheme*
- *Reduce the number of distinct output values to a much smaller set*
  - ◆ Original: { 0,1,2, ... 64 ... 128 ... 192 ... 255 }
  - ◆ Smaller set: {0, 64, 128, 192 }
- *Two different forms of quantization*
  - ◆ Partition the domain of input values into equally (unequally) spaced intervals
  - ◆ Uniform scalar quantization (**equally**)
  - ◆ Nonuniform scalar quantization (**unequally**)

# Example of Quantization

- *Original data*

- ◆ Total 64 data D:  $\{ d_1, d_2, \dots \dots d_{64} \}$

- ◆ Total 16 values X:  $\{ 0, 1, \dots \dots 16 \}$

- ◆ Probabilities of each value  $f(x)$ :

$$S = \{ 0 \quad 1 \quad 2 \quad 3 \quad 4 \quad 5 \quad 6 \quad 7 \quad 8 \quad 9 \quad 10 \quad 11 \quad 12 \quad 13 \quad 14 \quad 15 \}$$
$$P = \left\{ \frac{16}{64} \quad \frac{16}{64} \quad \frac{4}{64} \quad \frac{4}{64} \quad \frac{4}{64} \quad \frac{4}{64} \quad \frac{0}{64} \quad \frac{0}{64} \quad \frac{0}{64} \quad \frac{0}{64} \quad \frac{0}{64} \quad \frac{2}{64} \quad \frac{4}{64} \quad \frac{0}{64} \quad \frac{2}{64} \quad \frac{8}{64} \right\}$$

- *Compressed data Y:*  $\{ y_1, y_2, y_3, y_4 \}$

- ◆ The number of distinct output values will be decreased from 16 to 4

- ◆ How to choose the boundaries of the intervals and how to choose the values in each intervals

# Uniform Scalar Quantization

- *Partition the domain of input values into equally spaced intervals*

$$S = \{ 0 \ 1 \ 2 \ 3 \ 4 \ 5 \ 6 \ 7 \ 8 \ 9 \ 10 \ 11 \ 12 \ 13 \ 14 \ 15 \}$$

$$P = \left\{ \frac{16}{64} \ \frac{16}{64} \ \frac{4}{64} \ \frac{4}{64} \ \frac{4}{64} \ \frac{4}{64} \ \frac{0}{64} \ \frac{0}{64} \ \frac{0}{64} \ \frac{0}{64} \ \frac{0}{64} \ \frac{2}{64} \ \frac{4}{64} \ \frac{0}{64} \ \frac{2}{64} \ \frac{8}{64} \right\}$$

$$0 \ 1 \ *2 \ 3 / 4 \ 5 \ *6 \ 7 / 8 \ 9 \ *10 \ 11 / 12 \ 13 \ *14 \ 15$$

- $Y = \{1.5 \ 5.5 \ 9.5 \ 13.5\}$
- *Example in image compression*
  - ◆ Divide the RGB cube into equal slides in each dimension
  - ◆ R: 3-bit; G: 3-bit; B: 2-bit;

# Nonuniform Scalar Quantization

- *Partition the domain of input values into unequally spaced intervals*
  - ◆ Concentrate the bits to where is most need
- *Lloyd-Max quantization*
  - ◆ Try to minimize MSE

# Lloyd-Max Quantization

- *If the source distribution is not uniform, we must explicitly consider its probability distribution (probability density function)  $f_x(x)$ . Now we need the correct decision boundaries  $b_i$  and reconstruction values  $y_i$*

Begin

Choose initial level set  $Y_0$

$i = 0$ ;

Repeat

Compute  $B_i$  using

$i = i + 1$

Computer  $Y_i$  using

Until  $| Y_i - Y_{i-1} | < \text{threshold}$

$$b_j = \frac{y_{j+1} + y_j}{2}$$

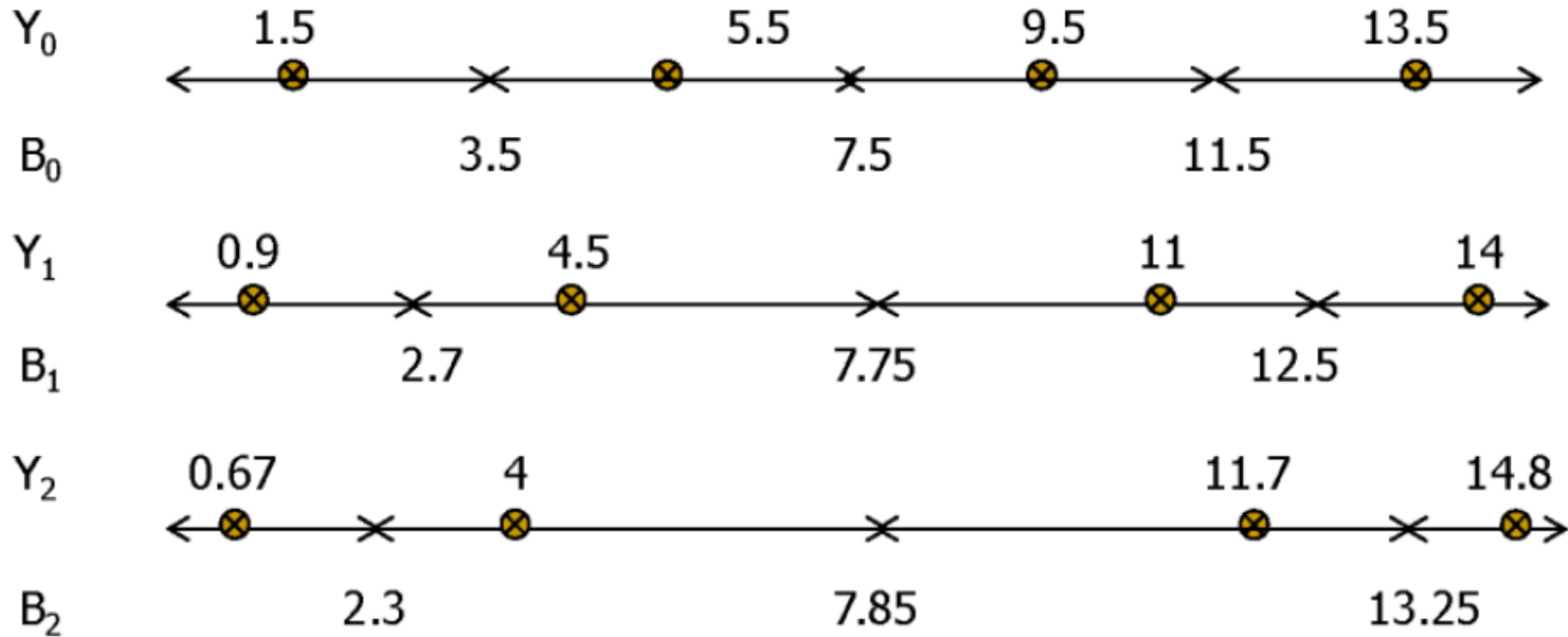
$$y_j = \frac{\int_{b_{j-1}}^{b_j} xf(x)dx}{\int_{b_{j-1}}^{b_j} f(x)dx}$$

End

# Lloyd-Max Quantization

0 1 \* 2 3 / 4 5 \* 6 7 / 8 9 \* 10 11 / 12 13 \* 14 15

$$Y = \{1.5, 5.5, 9.5, 13.5\}$$



0 1 2 / 3 4 5 6 7 / 8 9 10 11 12 13 / 14 15



# Summary

- *Lossy Compression*
  - ◆ Definition
  - ◆ Distortion measure
- *Quantization*
  - ◆ Uniform scale quantization
  - ◆ Nonuniform scale quantization