

Uniform quantizer

- Suppose we have a source that can be modeled as a random variable taking values in the interval  $[-4, 4]$  with more probability mass near the origin than away from it. We want to quantize this using the companding quantizer which composes of compressor and 3 bit uniform quantizer. The compressor characteristic we will use is given by the following equation:

$$c(x) = \begin{cases} 2x & \text{if } -1 \leq x \leq 1 \\ \frac{2x}{3} + \frac{4}{3} & x > 1 \\ \frac{2x}{3} - \frac{4}{3} & x < -1 \end{cases}$$

The mapping is shown graphically in figure 1. The inverse mapping is given by

$$c^{-1}(x) = \begin{cases} \frac{x}{2} & \text{if } -2 \leq x \leq 2 \\ \frac{3x}{2} - 2 & x > 2 \\ \frac{3x}{2} + 2 & x < -2 \end{cases}$$

The inverse mapping is shown graphically in figure 2. For this companding quantizer, what are the outputs for the following inputs:  $-0.8, 1.2, 0.5, 0.6, 3.2, -0.3$ . Compare your results with the case when the input is directly quantized with a uniform quantizer with the same number of levels. Comment on your results.

- Two companding characteristics that are widely used today are  $\mu$ -law companding and A-law companding. The  $\mu$ -law companding expander function is given by

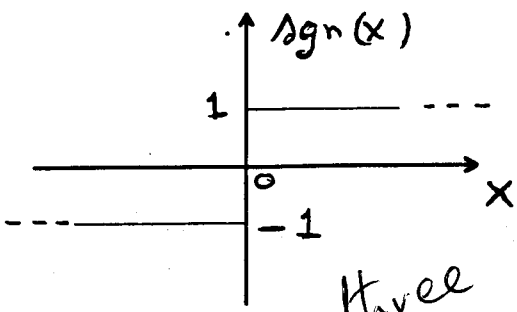
$$c^{-1}(x) = \frac{x_{\max}}{\mu} [(1 + \mu)^{\frac{|x|}{x_{\max}}} - 1] \text{sgn}(x).$$

Show that its compressor function is

SHOW ALL THE STEPS

$$c(x) = x_{\max} \frac{\ln(1 + \mu \frac{|x|}{x_{\max}})}{\ln(1 + \mu)} \text{sgn}(x).$$

where  $\mu$  is a constant,  $x$  is an input value and  $\text{sgn}(x)$  is the function of the form.



3. We want to DPCM-encode images using a ~~two~~-tap predictor of the form

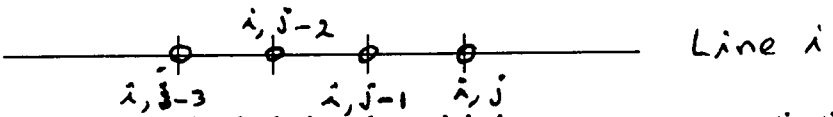
$$\hat{x}_{i,j} = a_1 \times x_{i,j-1} + a_2 \times x_{i,j-2} + a_3 x_{i,j-3}$$

and a four-level quantizer followed by a Huffman coder. Find the equations we need to solve to obtain coefficients  $a_1, a_2$  and  $a_3$  that minimize the mean squared error or

$$MSE = E[(x_{i,j} - \hat{x}_{i,j})^2]$$

and autocorrelation is

$$R_{xx}(k) = E[x_n x_{n+k}] = E[x_n x_{n-k}]$$

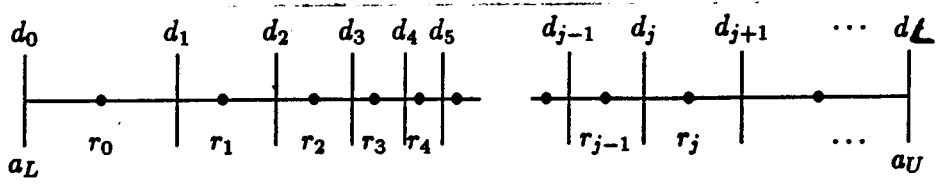


4. Lloyd-Max quantizer is a quantizer that is designed to minimize mean square quantization error (MSQE) or

$$MSQE = E[(f - \hat{f})^2] = \int_{a_L}^{a_U} (f - \hat{f})^2 p(f) df$$

where  $p(f)$  is probability density function of random variable  $f$ . The boundary of inputs are specified as  $a_L$  and  $a_U$ , number of levels is  $L$ .

Derive the constraints for decision level (input level),  $d_i$  and reconstruction level (output level),  $r_i$  of the Lloyd-Max quantizer.



$d_i$  = decision level (input level)  
 $r_i$  = reconstruction level (output level)

# COMPRESSOR

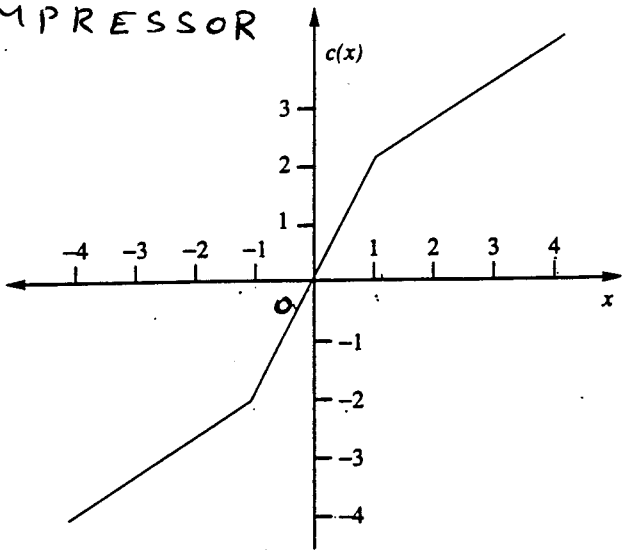


FIGURE 1.

# EXPANDER

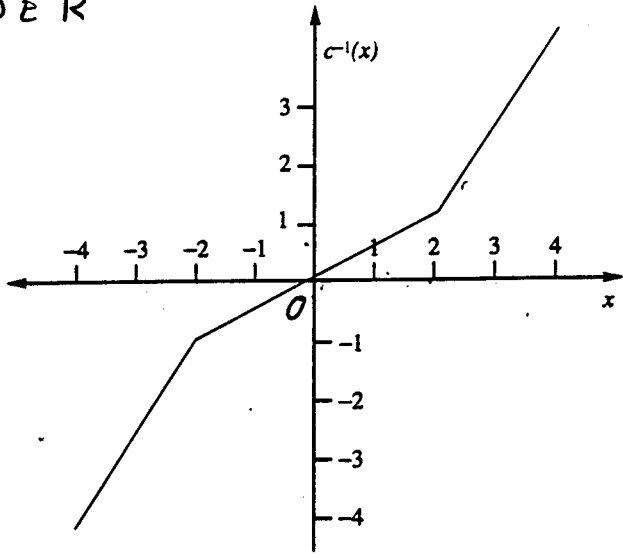


FIGURE 2.