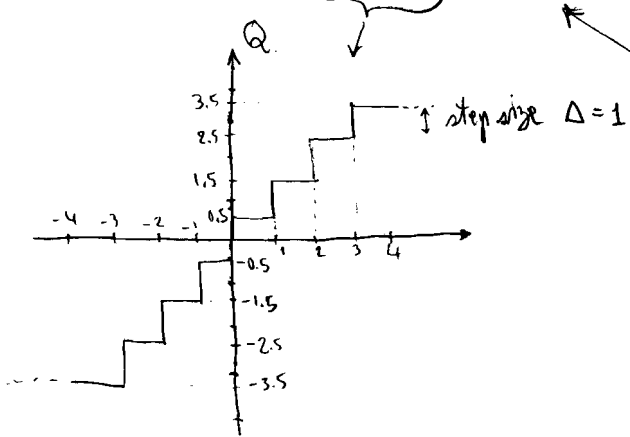


uniform step size $\Delta = \frac{4 \times 2}{2^3} = 1$

1.

input x	$C(x)$	$Q(C(x))$	output $\hat{C}(x)$	$x - \hat{C}(x)$, Compander error	quantizer directly $Q(x)$	$x - Q(x)$ quantizer error
-0.8	-1.6	-1.5	-0.75	-0.05	-0.5	-0.3
1.2	2.133	2.5	1.75	-0.55	1.5	-0.3
0.5	1	1.5	0.75	-0.25	0.5	0
0.6	1.2	1.5	0.75	-0.15	0.5	0.1
3.2	3.467	3.5	3.25	-0.05	3.5	-0.3
-0.3	-0.6	-0.5	-0.25	-0.05	-0.5	0.2



output Ans

$$\begin{aligned} \text{MSE with compander} &= \frac{(0.05)^2 + (0.55)^2 + (0.25)^2 + (0.15)^2 + (0.05)^2 + (0.05)^2}{6} \\ &= \frac{0.395}{6} = 0.06583 \end{aligned}$$

$$\begin{aligned} \text{MSE without compander} &= \frac{(0.3)^2 + (0.3)^2 + (0.1)^2 + (0.3)^2 + (0.2)^2 + 0}{6} \\ &= \frac{0.32}{6} = 0.0533 \end{aligned}$$

Without using compander, input data are mostly quantized in few level. When we apply compander system, all input data equally quantized (after Compressor $C(x)$) then they are mapped back by expander;

(-1)

②

$$\text{From } \bar{C}^{-1}(x) = \frac{\alpha_{\max}}{\mu} \left[(1+\mu)^{\frac{|x|}{\alpha_{\max}}} - 1 \right] \text{sgn}(x).$$

$$\text{Let } \bar{C}^{-1}(x) = y$$

$$\text{For } x > 0 ; \quad y = \frac{\alpha_{\max}}{\mu} \left[(1+\mu)^{\frac{x}{\alpha_{\max}}} - 1 \right]$$

$$\text{sgn}(x) = \begin{cases} 1 & x > 0 \\ -1 & x < 0 \end{cases}$$

$$1 + \frac{y\mu}{\alpha_{\max}} = (1+\mu)^{\frac{x}{\alpha_{\max}}}$$

$$\text{Take ln } \rightarrow ; \quad \ln \left(1 + \frac{y\mu}{\alpha_{\max}} \right) = \frac{x}{\alpha_{\max}} \ln(1+\mu)$$

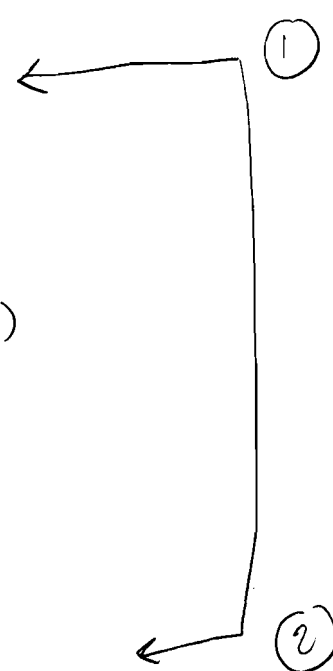
$$x = \alpha_{\max} \frac{\ln \left(1 + \frac{\mu y}{\alpha_{\max}} \right)}{\ln(1+\mu)}$$

$$\text{For } x < 0 ; \quad y = \frac{\alpha_{\max}}{\mu} \left[(1+\mu)^{\frac{-x}{\alpha_{\max}}} - 1 \right] (-1)$$

$$1 - \frac{\mu y}{\alpha_{\max}} = (1+\mu)^{\frac{-x}{\alpha_{\max}}}$$

$$\text{Take ln } ; \quad \ln \left(1 - \frac{\mu y}{\alpha_{\max}} \right) = -\frac{x}{\alpha_{\max}} \ln(1+\mu)$$

$$x = -\alpha_{\max} \frac{\ln \left(1 + \frac{\mu(-y)}{\alpha_{\max}} \right)}{\ln(1+\mu)}$$



Combine case ① and ②
for all x ;

$$x = \alpha_{\max} \frac{\ln \left(1 + \frac{\mu |y|}{\alpha_{\max}} \right)}{\ln(1+\mu)} \text{sgn}(y)$$

Let change variable;

$$y \rightarrow x$$

$$x \rightarrow \bar{C}(x).$$

$$\bar{C}(x) = \alpha_{\max} \frac{\ln \left(1 + \frac{\mu |x|}{\alpha_{\max}} \right)}{\ln(1+\mu)} \text{sgn}(x)$$

Ans

$$\textcircled{3} \quad \hat{x}_{i,j} = a_1 x_{i,j-1} + a_2 x_{i-1,j}$$

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

$$= E[(x_{i,j} - a_1 x_{i,j-1} - a_2 x_{i-1,j})^2]$$

$$\frac{\partial \text{MSQE}}{\partial a_1} = -2 E[(x_{i,j} - a_1 x_{i,j-1} - a_2 x_{i-1,j}) x_{i,j-1}] = 0$$

$$\underbrace{E[x_{i,j} x_{i,j-1}]}_D = a_1 \underbrace{E[x_{i,j-1} x_{i,j-1}]}_A + a_2 \underbrace{E[x_{i-1,j} x_{i,j-1}]}_B \quad \textcircled{1}$$

$$\frac{\partial \text{MSQE}}{\partial a_2} = -2 E[(x_{i,j} - a_1 x_{i,j-1} - a_2 x_{i-1,j}) x_{i-1,j}] = 0$$

$$\underbrace{E[x_{i,j} x_{i-1,j}]}_E = a_1 \underbrace{E[x_{i,j-1} x_{i-1,j}]}_B + a_2 \underbrace{E[x_{i-1,j} x_{i-1,j}]}_C \quad \textcircled{2}$$

solve for a_1, a_2

$$\textcircled{1} \rightarrow a_1 A + a_2 B = D$$

$$\textcircled{2} \rightarrow a_1 B + a_2 C = E$$

$$\begin{bmatrix} A & B \\ B & C \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} D \\ E \end{bmatrix}$$

$$\underline{R} \quad \underline{a} = \underline{z}$$

$$\underline{a} = \underline{R}^{-1} \underline{z}$$

Ans

4.

Find. Partition $P(\hat{A}_0)$

given uniform quantizer. $S = \left\{ \begin{matrix} S_1 & S_2 & S_3 & S_4 \\ (2,2), (2,-2), (-2,2), (-2,-2) \end{matrix} \right\}$

$$S_1 = \{x_6, x_8, x_{10}\}$$

$$S_2 = \{x_2, x_9\}$$

$$S_3 = \{x_1, x_3, x_7\}$$

$$S_4 = \{x_4, x_5\}$$

set $m = 0$

$$\text{Distortion } D_0 = \frac{1}{10} \sum_{i=1}^{10} \min d(x_i, \hat{x}) = 4.6270$$

\downarrow
 $(x_i - \hat{x})^2$

$$\begin{aligned} \min d(x_i, \hat{x}) &= (2-1.1)^2 + (2-0.5)^2 + (2-0.1)^2 + (2-1.7)^2 + (2-0.3)^2 + (2-4.8)^2 \\ &+ (2-0.6)^2 + (-2+0.1)^2 + (2-0.7)^2 + (-2+0.3)^2 \\ &+ (-2+0.5)^2 + (2-1)^2 + (-2+0.8)^2 + (2-0.6)^2 + (-2+0.6)^2 + (2-0.2)^2 \\ &+ (-2+0.7)^2 + (-2+1.2)^2 + (2+0.2)^2 + (-2+0.9)^2 = 4.6270 \end{aligned}$$

Let $D_{-1} = \infty$

$$\frac{D_{-1} - D_0}{D_0} = \frac{\infty - 4.6270}{4.6270} > 0.001 \quad \rightarrow \epsilon \quad \text{continue. Find } P(\hat{A}_1).$$

next page.

④ continue

Partition $P(\hat{A}_1)$ such that $d(x_i; \hat{x}_j) < d(x_i; \hat{x}_k)$; $j \neq k$,

we get.

$$\hat{x}(S_1) = \left(\frac{x_6 + x_8 + x_{10}}{3} \right) = \left(\frac{1.1 + 0.1 + 0.3}{3}, \frac{0.5 + 1.7 + 4.8}{3} \right) = (0.5, 2.333)$$

$$\hat{x}(S_2) = \left(\frac{x_2 + x_9}{2} \right) = \left(\frac{0.6 + 0.7}{2}, \frac{-0.1 - 0.3}{2} \right) = (0.65, 0.2)$$

$$\hat{x}(S_3) = \left(\frac{x_1 + x_3 + x_7}{3} \right) = \left(\frac{-0.5 - 0.8 - 0.6}{3}, \frac{1 + 0.6 + 0.2}{3} \right) = (0.633, 0.6)$$

$$\hat{x}(S_4) = \left(\frac{x_4 + x_5}{2} \right) = \left(\frac{-0.7 - 0.2}{2}, \frac{-1.2 - 0.9}{2} \right) = (-0.45, -1.05)$$

set $m=1$

$$\text{Distortion } D_1 = \frac{1}{10} \sum_{i=1}^{10} \min d(x_i, \hat{x}) = 1.2697$$

$$\frac{D_0 - D_1}{D_1} = \frac{4.627 - 1.2697}{1.2697} = 2.644 > 0.001 \quad \text{continue Find } P(\hat{A}_2)$$

$$P(\hat{A}_2) = \hat{x} P(\hat{A}_1) = \hat{A}_1$$

then $D_2 = D_1$

$$\text{so } \frac{D_1 - D_2}{D_2} = 0 < 0.001 \quad \text{Halt (stop)}$$

Therefore, the codebook is

$$(0.5, 2.333), (0.65, 0.2), (0.633, 0.6), (-0.45, -1.05)$$

Ans