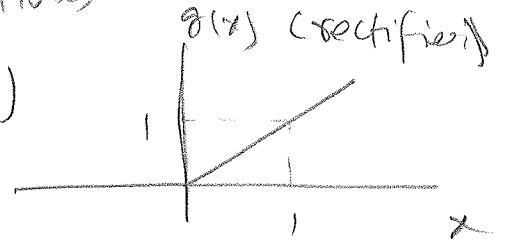


① Input $X \sim N(0, \sigma^2)$ $Y = g(X)$



(a) current $= \frac{Y}{R}$

$$\Pr\left(\left\{\frac{Y}{R} > 1\right\}\right) = \Pr(\{Y > R\}) = \Pr(\{X > R\}) \text{ since } R > 0$$

$$\Pr(\{X > R\}) = \int_R^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2\sigma^2}} dx = Q\left(\frac{R}{\sigma}\right)$$

$$\text{where } Q(\alpha) = \int_{\alpha}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2\sigma^2}} dx = 1 - \Phi(\alpha)$$

(b) power $= \frac{Y^2}{R}$

$$\Pr\left(\left\{\frac{Y^2}{R} > 1\right\}\right) = \Pr(\{Y^2 > R\}) = \Pr(\{X > \sqrt{R}\}) = Q\left(\frac{\sqrt{R}}{\sigma}\right)$$

(c) $\mu_I = E\left[\frac{Y}{R}\right] = \frac{1}{R} E[Xu(X)] = \frac{1}{R} \int_0^{\infty} x f_X(x) dx$

$$= \frac{1}{R} \int_0^{\infty} x \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}} dx = \frac{1}{\sqrt{2\pi}\sigma R} (-\sigma^2) e^{-\frac{x^2}{2\sigma^2}} \Big|_0^{\infty} = \frac{\sigma}{\sqrt{2\pi}R}$$

$$\begin{aligned} E\left[\frac{Y^2}{R^2}\right] &= \frac{1}{R^2} \int_0^{\infty} x^2 \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}} dx = \frac{1}{\sqrt{2\pi}\sigma R^2} \left\{ \left[-x\sigma^2 e^{-\frac{x^2}{2\sigma^2}} \right]_0^{\infty} + \sigma^2 \int_0^{\infty} e^{-\frac{x^2}{2\sigma^2}} dx \right\} \\ &= \frac{\sigma}{\sqrt{2\pi}R^2} \left(\int_0^{\infty} e^{-\frac{x^2}{2\sigma^2}} dx \right) \end{aligned}$$

Suppose $A = \int_0^{\infty} e^{-\frac{x^2}{2\sigma^2}} dx$

$$A^2 = \int_0^{\infty} \int_0^{\infty} e^{-\frac{x^2+y^2}{2\sigma^2}} dx dy = \int_0^{\pi/2} \int_0^{\infty} e^{-\frac{r^2}{2\sigma^2}} r d\theta dr = \frac{\sigma^2 \pi}{2}$$

$$E\left[\frac{Y^2}{R^2}\right] = \frac{\sigma}{\sqrt{2\pi}R^2} \times \sigma \sqrt{\frac{\pi}{2}} = \frac{\sigma^2}{2R^2}$$

$$\sigma^2_I = \text{Var}\left[\frac{Y^2}{R^2}\right] = E\left[\frac{Y^2}{R^2}\right] - \left(E\left[\frac{Y}{R}\right]\right)^2 = \frac{\sigma^2}{2R^2} \left(1 - \frac{1}{\pi}\right)$$

(d) $\mu_P = E\left[\frac{Y^2}{R}\right] = \frac{1}{R} \int_0^{\infty} x^2 \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}} dx = \frac{\sigma^2}{2R}$

From $E[(X - \bar{X})^4] = 3\sigma^4$ and $\bar{X} = 0$

$$\int_{-\infty}^{\infty} x^4 \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}} dx = 3\sigma^4$$

$$\Rightarrow \int_0^{\infty} x^4 \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}} dx = \frac{3}{2} \sigma^4 \quad \text{---}$$

$$E\left[\frac{Y^4}{R^2}\right] = \frac{1}{R^2} \int_0^{\infty} x^4 \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}} dx = \frac{3\sigma^4}{2R^2}$$

$$\sigma^2_P = Var\left[\frac{Y^4}{R^2}\right] = E\left[\frac{Y^4}{R^2}\right] - \left(E\left[\frac{Y^2}{R}\right]\right)^2 = \frac{3\sigma^4}{2R^2} - \frac{\sigma^4}{4R^2} = \frac{5\sigma^4}{4R^2}$$

②

$$\begin{aligned} \textcircled{a} \quad E[X] &= E[X|G_1] \cdot P(G_1) + E[X|G_2] \cdot P(G_2) \\ &= 80 \times 0.6 + 20 \times 0.4 \\ &= 48 + 8 = 56 \end{aligned}$$

$$\begin{aligned} \text{var}(X) &= E[\text{var}(X|G_i)] + \text{var}[E[X|G_i]] \\ &= 20 \times 0.6 + 10 \times 0.4 + (80-56)^2 \times 0.6 + (20-56)^2 \times 0.4 \\ &= 880 \end{aligned}$$

where G_i denotes the groups.

$$\sigma_X = 29.66$$

$$\textcircled{b} \quad f_X(x) = 0.6 \left(\frac{1}{\sqrt{2\pi} \sqrt{20}} e^{-\frac{(x-80)^2}{2 \times 20}} \right) + 0.4 \frac{1}{\sqrt{2\pi} \sqrt{10}} e^{-\frac{(x-20)^2}{2 \times 10}}$$

X is NOT Gaussian here

$$P(X > 56 + 29.66) = 0.6 \left(1 - \Phi \left(\frac{56 + 29.66 - 80}{\sqrt{20}} \right) \right) + 0.4 \left(1 - \Phi \left(\frac{56 + 29.66 - 20}{\sqrt{10}} \right) \right)$$

$P(A)$

$$= 0.6 (1 - \Phi(1.265)) + 0.4 (1 - \Phi(20.76))$$

$$\approx 0.0617$$

$$P(B) = P(\bar{x} - \sigma_X < X \leq \bar{x} + \sigma_X)$$

$$= 0.6 \left(\Phi(1.265) - \Phi(-5.366) \right) + 0.4 \left(\Phi(20.76) - \Phi(11.38) \right)$$

$$\approx 0.53826$$

$$P(C) = P(\bar{x} - \sigma_X < X \leq \bar{x}) \dots$$

Proceed further on similar lines.

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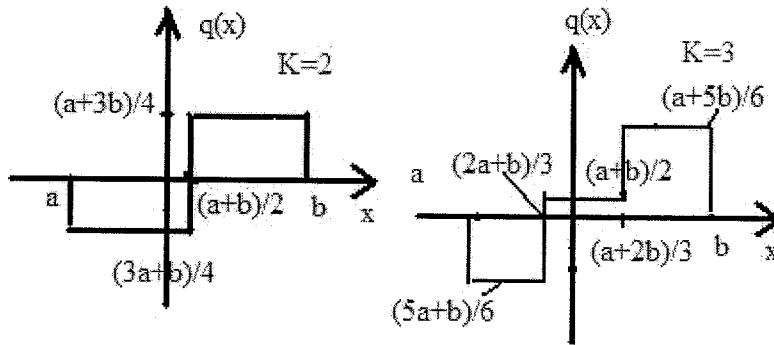


(a) If the quantizer has K levels, then the i^{th} interval will be

$$x \in \left(a + (i-1) \times \frac{(b-a)}{K}, a + i \times \frac{(b-a)}{K} \right), i = 1, \dots, K$$

When x is in that interval, the output $Y = q(X)$ is equal to the midpoint of the interval.

$$y = a + \frac{(b-a)(2i-1)}{2K}$$

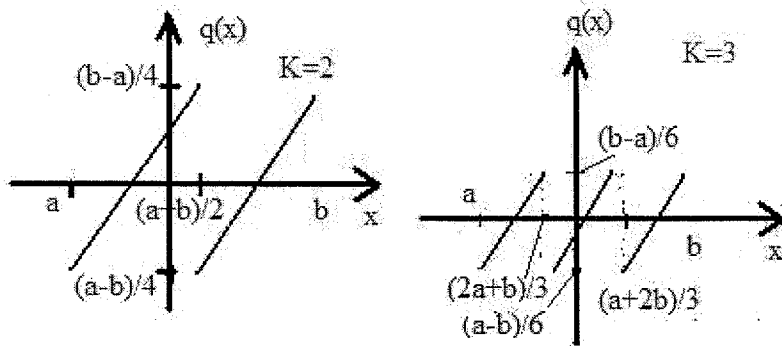


(b) $g(x) = x - q(x)$

In the i^{th} interval,

$$x \in \left(a + (i-1) \times \frac{(b-a)}{K}, a + i \times \frac{(b-a)}{K} \right), i = 1, \dots, K$$

$$g(x) = x - \left[a + \frac{(b-a)(2i-1)}{2K} \right]$$



(c) for general K

$$Y = q(X)$$

$$f_Y(y) = \sum_{i=1}^K \frac{1}{K} \delta \left(y - \left[a + \frac{(b-a)(2i-1)}{2K} \right] \right)$$

$$E = g(X)$$

$$f_E(e) = f_X(x) \left| \frac{dx_1}{de} \right| + \dots + f_X(x) \left| \frac{dx_K}{de} \right|, E \in \left[\frac{(a-b)}{2K}, \frac{(b-a)}{2K} \right]$$

$$= 0, \text{ else}$$

$$f_E(e) = \begin{cases} \frac{K}{(b-a)}, & E \in \left[\frac{(a-b)}{2K}, \frac{(b-a)}{2K} \right] \\ 0 & , \text{ else} \end{cases}$$

$$(d) E[E^2] = \int_a^b g(x)^2 f_X(x) dx = K \int_a^{a + \frac{1}{K}(b-a)} \left(x - \left(a + \frac{b-a}{2K} \right) \right)^2 \frac{1}{(b-a)} dx$$

$$= \frac{K}{(b-a)} \frac{1}{3} \left[x - \left(\frac{(2K-1)a+b}{2K} \right) \right]^3 \Big|_a^{a + \frac{b-a}{K}}$$

$$= \frac{K}{3(b-a)} \left\{ \left[a + \frac{b-a}{K} - \left(\frac{(2K-1)a+b}{2K} \right) \right]^3 - \left[a - \left(\frac{(2K-1)a+b}{2K} \right) \right]^3 \right\}$$

$$= \frac{K}{3(b-a)} \left[\left(\frac{b-a}{2K} \right)^3 - \left(\frac{a-b}{2K} \right)^3 \right]$$

$$= \frac{(b-a)^2}{12K^2}$$

Alternatively, E is uniform in $\left[\frac{a-b}{2K}, \frac{b-a}{2K} \right]$

$$E[E] = 0$$

$$E[E^2] = \text{var}[E] = \frac{(b-a)^2}{12K}$$

(4)

ii) Poisson Process.

$$\begin{aligned}
 (a) \Pr(T > 5) &= \int_5^{\infty} \frac{1}{t} e^{-\frac{t}{5}} dt \\
 &= - \int_5^{\infty} e^{-\frac{t}{5}} d\left(-\frac{t}{5}\right) \\
 &= -e^{-\frac{t}{5}} \Big|_5^{\infty} \\
 &= e^{-1}
 \end{aligned}$$

$$\begin{aligned}
 \text{Alternatively, } \Pr(T > t) &= 1 - \Pr(T \leq t) \\
 &= 1 - F_T(t) \\
 &= 1 - (1 - e^{-\frac{t}{5}}) \\
 &= e^{-1}
 \end{aligned}$$

$$\begin{aligned}
 (b) \Pr(T > 10) &= \int_{10}^{\infty} \frac{1}{t} \left(\frac{t}{5}\right) e^{-\frac{t}{5}} dt \\
 &= \frac{1}{5} \int_{10}^{\infty} t e^{-\frac{t}{5}} d\left(-\frac{t}{5}\right) \\
 &= -\frac{1}{5} \int_{10}^{\infty} t e^{-\frac{t}{5}} = -\frac{1}{5} t e^{-\frac{t}{5}} \Big|_{10}^{\infty} + \frac{1}{5} \int_{10}^{\infty} e^{-\frac{t}{5}} dt \\
 &= 2e^{-2} - \int_{10}^{\infty} e^{-\frac{t}{5}} d\left(-\frac{t}{5}\right) \\
 &= 2e^{-2} - e^{-\frac{t}{5}} \Big|_{10}^{\infty} \\
 &= 3e^{-2}
 \end{aligned}$$

$$(c) \Pr(T > 10 | T > 5) = \Pr(T > 5) = e^{-1} \quad \text{Memoryless Property.}$$

$$\begin{aligned}
 \text{Alternatively } \Pr(T > 10 | T > 5) &= \frac{\Pr(T > 10)}{\Pr(T > 5)} = \frac{\int_{10}^{\infty} \frac{1}{t} e^{-\frac{t}{5}} dt}{e^{-1}} \\
 &= e^{-1}
 \end{aligned}$$

$$\begin{aligned}
 (d) \Pr(T_2 > 20 | T_2 > 10) &= \frac{\Pr(T_2 > 20)}{\Pr(T_2 > 10)} = \frac{\int_{20}^{\infty} \frac{1}{t} \left(\frac{t}{5}\right) e^{-\frac{t}{5}} dt}{3e^{-2}} \\
 &= \frac{5e^{-4}}{3e^{-2}} = \frac{5}{3}e^{-2}
 \end{aligned}$$

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(e) $E[T | T > 5] = 5 + E[T] = 10$ Memoryless Property.

Alternatively, $f_T(t | T > 5) = \frac{f_T(t)}{Pr(T > 5)} = \frac{\frac{1}{5} e^{-\frac{t}{5}}}{e^{-1}} \quad t > 5$

$$\begin{aligned} E[T | T > 5] &= \int_5^{\infty} t f_T(t | T > 5) dt \\ &= \int_5^{\infty} t e^{+1} \frac{1}{5} e^{-\frac{t}{5}} dt \\ &= -e \int_5^{\infty} t e^{-\frac{t}{5}} d(-\frac{t}{5}) \\ &= -e \int_5^{\infty} t d(e^{-\frac{t}{5}}) \\ &= -e t e^{-\frac{t}{5}} \Big|_5^{\infty} + e \int_5^{\infty} e^{-\frac{t}{5}} dt \\ &= 5 + 5 = 10 \end{aligned}$$

(f) $E[T_2 | T_2 > 10]$

$f_{T_2}(t | T_2 > 10) = \frac{f_{T_2}(t)}{Pr(T_2 > 10)} = \frac{\frac{1}{5}(\frac{t}{5}) e^{-\frac{t}{5}}}{3e^{-2}} \quad t > 10$

$$\begin{aligned} E[T_2 | T_2 > 10] &= \int_{10}^{\infty} t f_{T_2}(t | T_2 > 10) dt \\ &= \frac{e^2}{75} \int_{10}^{\infty} t^2 e^{-\frac{t}{5}} dt \\ &= \frac{e^2}{75} \left[-5 \int_{10}^{\infty} t^2 e^{-\frac{t}{5}} d(-\frac{t}{5}) \right] \\ &= \frac{e^2}{75} \left[-5 \int_{10}^{\infty} t^2 d(e^{-\frac{t}{5}}) \right] \\ &= \frac{e^2}{75} \left[-5 t^2 e^{-\frac{t}{5}} \Big|_{10}^{\infty} + 5 \int_{10}^{\infty} e^{-\frac{t}{5}} 2t dt \right] \\ &= \frac{e^2}{75} \left[500e^{-2} - 50 \int_{10}^{\infty} e^{-\frac{t}{5}} d(-\frac{t}{5}) \right] \\ &= \frac{e^2}{75} \left[500e^{-2} - 50 t e^{-\frac{t}{5}} \Big|_{10}^{\infty} + 50 \int_{10}^{\infty} e^{-\frac{t}{5}} dt \right] \\ &= \frac{e^2}{75} \left[500e^{-2} + 500e^{-2} - 50 \int_{10}^{\infty} e^{-\frac{t}{5}} d(-\frac{t}{5}) \right] \\ &= \frac{e^2}{75} (1250e^{-2}) = \frac{50}{3} \end{aligned}$$

(ii) Bernoulli Process

$$(a) P_r(T \geq 5) = 1 - P_r(T \leq 4)$$

$$p = \frac{1}{5}$$

$$= 1 - \sum_{t=1}^4 p(1-p)^{t-1} = 0.41$$

$$\text{Alternatively, } P_r(T \geq 5) = \sum_{t=5}^{\infty} p(1-p)^{t-1} = \frac{1-5p+10p^2-10p^3+5p^4-p^5}{1-p} = 0.41$$

(b) Let T_2 be a 2nd order negative binomial r.v.

$$P_r(T_2 \geq 10) = 1 - P_r(T_2 \leq 9)$$

$$= 1 - \sum_{t_2=2}^9 \binom{t_2-1}{2-1} (1-p)^{t_2-2} p^2 = 0.44$$

$$\text{Alternatively, } P_r(T_2 \geq 10) = \sum_{t_2=10}^{\infty} \binom{t_2-1}{2-1} (1-p)^{t_2-2} p^2 = p^2 \sum_{t_2=10}^{\infty} \binom{t_2-1}{2-1} (1-p)^{t_2-2}$$

$$= \frac{(-1+p)^{10}(1+8p)}{(1-p)^2} = 0.44$$

by Memoryless Property.

$$(c) P_r(T \geq 10 | T \geq 5) = P_r(T \geq 10 | T > 4) = P_r(T \geq 6)$$

$$= 1 - \sum_{t=1}^5 p(1-p)^{t-1} = 0.33$$

$$\text{Alternatively } P_r(T \geq 10 | T \geq 5) = \frac{P_r(T \geq 10)}{P_r(T \geq 5)} = \frac{\sum_{t=10}^{\infty} p(1-p)^{t-1}}{\sum_{t=5}^{\infty} p(1-p)^{t-1}} = 0.33$$

$$(d) P_r(T_2 \geq 20 | T_2 \geq 10) = \frac{P_r(T_2 \geq 20)}{P_r(T_2 \geq 10)} = \frac{1 - \sum_{t_2=2}^{19} \binom{t_2-1}{2-1} (1-p)^{t_2-2} p^2}{1 - \sum_{t_2=2}^9 \binom{t_2-1}{2-1} (1-p)^{t_2-2} p^2}$$

$$= 0.19$$

$$(e) E[T|T \geq 5] = E[T|T \geq 4] = E[T] + 4 = \frac{1}{p} + 4 = 9.$$

by Memoryless Property

$$\text{Alternatively, } P_r(T=t|T \geq 5) = \frac{P_r(T=t)}{P_r(T \geq 5)} = \frac{p(1-p)^{t-1}}{\sum_{t=5}^{\infty} p(1-p)^{t-1}}$$

$$= \frac{(1-p)^{t-5} p}{1 - 5p + 10p^2 - 10p^3 + 5p^4 - p^5}$$

$$E[T|T \geq 5] = \sum_{t=5}^{\infty} t P_r(T=t|T \geq 5) = \frac{1+4p}{p} = 9$$

$$f) E[T_2] = ?$$

$$P_r(T_2 = t_2 | T_2 \geq 10) = \frac{P_r(T_2 = t_2)}{P_r(T_2 \geq 10)} = \frac{\binom{t_2-1}{2-1} (1-p)^{t_2-2} p^2}{1 - \sum_{t=2}^9 \binom{t-1}{2-1} (1-p)^{t-2} p^2} \quad t_2 \geq 10$$

$$= (t_2-1) \cdot 0.09 \cdot 0.8^{t_2-2}$$

$$E[T_2 | T_2 \geq 10] = \sum_{t_2=10}^{\infty} t_2 P_r(T_2 = t_2 | T_2 \geq 10) = 15.54.$$