

$$\boxed{2.9} \quad \mu_Q = n \cdot p = (400)(0.025) = 10$$

$$\sigma_Q = \sqrt{npq} = 3.1225$$

∴ Here,

$$z^* = h_1 / \sigma_Q \approx 3.2$$

$$j^* = (\mu_Q - k_1) / \sigma_Q \approx 0.64$$

Assume the normality of the binomial dist. of Q .

Now, using table A.4 in the text, linear interpolation can be used to get $ARL \approx 6$.

∴ An approximate $ARL \approx 6$

$\boxed{2.11}$

(a). From table 4.7, for All-OK $ARL = 370$ and off-target $ARL = 5$, we get the value of $\sqrt{n} \delta / \sigma \gg 1.42$

$$\text{Here, } \delta / \sigma = (800 - 750) / 60 = 5/6$$

$$\Rightarrow n \gg 2.9. \text{ So take } n = 3$$

$$\text{Now } \mu_Q = 800, \sigma_Q = \sigma / \sqrt{n} = 34.65$$

$$\text{Now, } k_2 = k^{opt} = \mu_Q - \delta/2 = 800 - 25 = 775$$

$$z = (\mu_Q - k_2) / \sigma_Q = 0.72 \text{ and All-OK } ARL = 370, \text{ from table A.5}$$

$$\text{of text } \Rightarrow z \approx 3.02$$

$$\Rightarrow h_2 = z \cdot \sigma_Q = 104.6$$

$$(b). \text{ Want } n \ni \left. \begin{aligned} P(\bar{x} < \# \mid x_i \stackrel{iid}{\sim} N(800, 60^2)) &= \frac{1}{370} = 0.0027 \\ \text{and } P(\bar{x} < \# \mid x_i \stackrel{iid}{\sim} N(750, 60^2)) &= \frac{1}{5} = 0.2 \end{aligned} \right\}$$

$$\Rightarrow \left. \begin{aligned} \Phi\left(\frac{\# - 800}{60/\sqrt{n}}\right) &= 0.0027 = \Phi(-2.78) \\ \text{and } \Phi\left(\frac{\# - 750}{60/\sqrt{n}}\right) &= 0.2 = \Phi(-0.84) \end{aligned} \right\} \Rightarrow \left. \begin{aligned} \# - 800 + 2.78 \times \frac{60}{\sqrt{n}} &= 0 \\ \# - 750 + 0.84 \times \frac{60}{\sqrt{n}} &= 0 \end{aligned} \right\} (*)$$

$$(*) \Rightarrow n = 5.42 \approx 6, \# = 728.35 \approx 728 \Rightarrow n = 6 \text{ and } LCL \approx 728.$$

2.12

(a). $k_1 = \mu_0 + \delta/2 = 45 + 0.7/2 = 45.35$
 $k_2 = - = 44.65$

$h = (k_1 - \mu_0) / \sigma_0 = 0.35 / 0.7 = 0.5$

From table 4.5 of text, we get $z_h = 5.07$

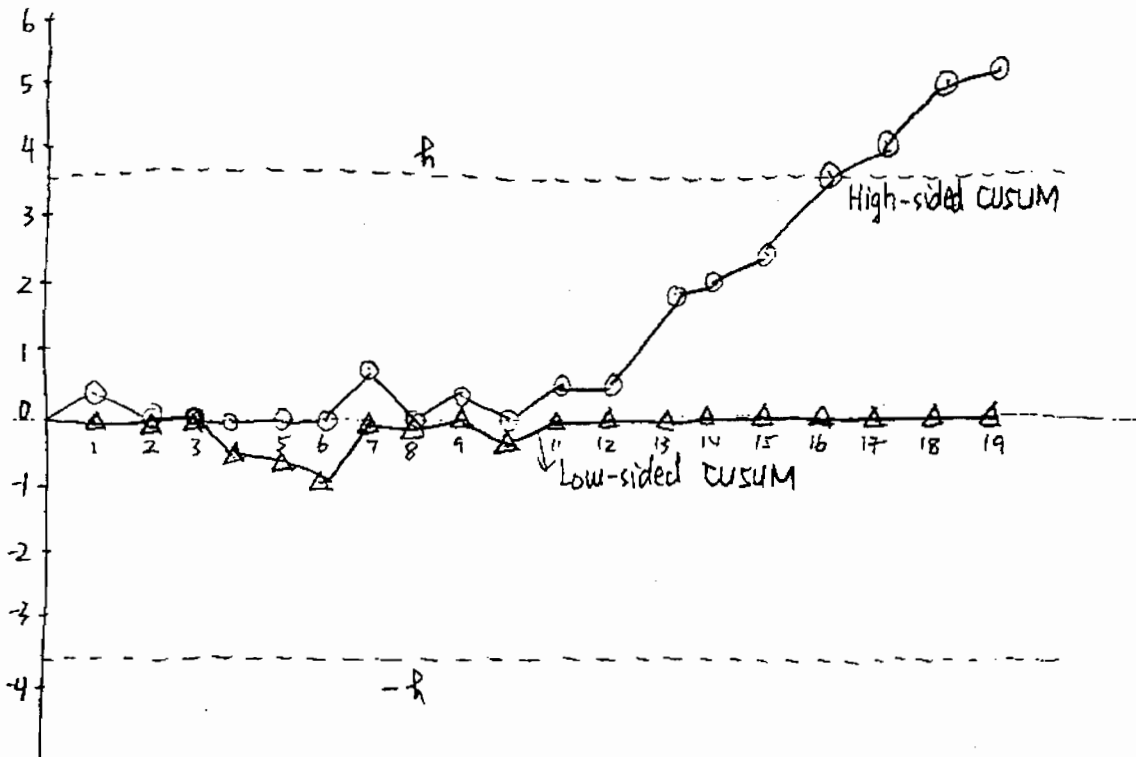
$\therefore h = z_h \cdot \sigma_0 = (5.07)(0.7) = 3.549$

\therefore The required CUSUM scheme is given by the parameters:

$u_0 = 0, k_1 = 45.35, k_2 = 44.65, h = 3.549$

(b). Use a 0-headstart 2-sided CUSUM

$U_i = \max(0, (\delta_i - k_1) + U_{i-1}), L_i = \min(0, (\delta_i - k_2) + L_{i-1})$



Yes, the high-side CUSUM plot shows that there is a signal at the 16th sample. Low-side CUSUM plot doesn't reveal any signal

2.14 Running Gan's cusum ARL program.

$\mu_R = 0$:

$L_1(u) = 63.9, \quad L_2(u) = 63.9, \quad L_1(0) = 68.19, \quad L_2(0) = 68.19$

$ARL_{combined} = \frac{L_1(0)L_2(u) + L_2(0)L_1(u) - L_1(0)L_2(0)}{L_1(0) + L_2(0)} = \underline{\underline{29.805}}$

$\mu_R = 1$:

$L_1(u) = 4.07, \quad L_2(u) = 10839.89, \quad L_1(0) = 5.42, \quad L_2(0) = 10861.33$

$ARL_{combined} = \underline{\underline{4.05728}}$


2.25 Here, $X \sim \text{Poisson}(\lambda)$.

$f_{\lambda}(x) = \frac{e^{-\lambda} \cdot \lambda^x}{x!}, \quad x = 0, 1, 2, \dots$

To get ARL, use Markov Chain approach. Consider the 6 states =

- $S_1 = \{ X_{t-4}, X_{t-3}, X_{t-2}, X_{t-1} \} = \{ 0, 0, 0, 0 \}$
- $S_2 = \{ 0, 0, 0, 1 \}$
- $S_3 = \{ 0, 0, 1, 0 \}$
- $S_4 = \{ 0, 1, 0, 0 \}$
- $S_5 = \{ 1, 0, 0, 0 \}$
- $S_6 = \text{alarm}$

The transition matrix is:

	S_1	S_2	S_3	S_4	S_5	S_6	
S_1	$e^{-\lambda}$	$\lambda e^{-\lambda}$	0	0	0	$1 - (e^{-\lambda} + \lambda e^{-\lambda})$	$ARL = L_1$ 
S_2	0	0	$e^{-\lambda}$	0	0	$1 - e^{-\lambda}$	
S_3	0	0	0	$e^{-\lambda}$	0	$1 - e^{-\lambda}$	
S_4	0	0	0	0	$e^{-\lambda}$	$1 - e^{-\lambda}$	
S_5	$e^{-\lambda}$	0	0	0	0	$1 - e^{-\lambda}$	
S_6	0	0	0	0	0	1	

Let $\underline{L} = (L_1, \dots, L_5)^T$, $L_i = \text{mean \# transitions required to move from state } S_i \text{ to alarm } (S_6)$.
 $R = \text{upper left } 5 \times 5 \text{ submatrix of transition matrix. Compute } \underline{L} = (I_5 - R)^{-1} \cdot \underline{1}_5$

2.26

$$(a). ARL = \frac{1}{1 - p(-2) - p(-1) - p(0) - p(1) - p(2)}$$

(b). Define the states S_i 's of a Markov Chain as follows :

$$S_i = "X_t = i", \quad i = -2, -1, 0, 1, 2$$

$$\text{and } S_a = "|X_t| \geq 3"$$

The transition matrix is :

	S_{-2}	S_{-1}	S_0	S_1	S_2	S_a
S_{-2}	$p(0)$	$p(1)$	$p(2)$	$p(3)$	$p(4)$	$1 - \sum_0^4 p(j)$
S_{-1}	$p(-1)$	$p(0)$	$p(1)$	$p(2)$	$p(3)$	$1 - \sum_{-1}^3 p(j)$
S_0	$p(-2)$	$p(-1)$	$p(0)$	$p(1)$	$p(2)$	$1 - \sum_{-2}^2 p(j)$
S_1	$p(-3)$	$p(-2)$	$p(-1)$	$p(0)$	$p(1)$	$1 - \sum_{-3}^1 p(j)$
S_2	$p(-4)$	$p(-3)$	$p(-2)$	$p(-1)$	$p(0)$	$1 - \sum_{-4}^0 p(j)$
S_a	0	0	0	0	0	1

where R = upper left 5×5 submatrix of the above transition matrix

Now, the ARL will be the 3rd element of

$$\underline{L} = (I^{5 \times 5} - R)^{-1} \underline{1}^{5 \times 1} \quad (\text{since } \underline{L} = L(-2), L(-1), L(0), L(1), L(2))$$

2.27

$$(a). L(u) = 1 \cdot \{P(X_2 \geq UCL) + P(X_2 \leq LCL)\} + \int_{LCL}^{UCL} (1 + L(v)) f_{X_2|X_1}(v) dv$$

(where $f_{X_2|X_1}(v)$ = pdf of a normal dsn with mean μ , variance σ^2)

$$= 1 + \int_{LCL}^{UCL} L(v) f_{X_2|X_1}(v) dv$$

$$(b). ARL = 1 \cdot \{P(X_1 \leq LCL) + P(X_1 \geq UCL)\} + \int_{LCL}^{UCL} (1 + L(v)) f_{X_1}(v) dv$$

$$= 1 + \int_{LCL}^{UCL} L(v) \cdot f_{X_1}(v) dv \quad \left[\text{where } f_{X_1}(v) = \text{pdf of } N(0, \frac{\sigma^2}{1-p^2}) \right]$$

2.30

6

$$(a). L(u) = 1 + L(0) F(k_1 - u) + \int_0^h L(y) f(y + k_1 - u) dy$$

$$= 1 + L(0) (1 - e^{-u-k_1}) I(k_1 \geq u) + \int_0^h L(y) e^{-(y+k_1-u)} \cdot I(y \geq u - k_1) dy$$

$$L'(u) = -L(0) \cdot e^{-u-k_1} \cdot I(k_1 \geq u) + \int_0^h L(y) e^{-(y+k_1-u)} \cdot I(y \geq u - k_1) dy$$

Now, if $k_1 \geq u \geq 0$,

$$L'(u) = -L(0) e^{-u-k_1} + \int_0^h L(y) e^{-(y+k_1-u)} dy$$

$$= L(u) - L(0) - 1 \left(\begin{array}{l} \text{--- In this case, } L(u) = 1 + L(0)(1 - e^{-u-k_1}) \\ \text{--- } + \int_0^h L(y) e^{-(y+k_1-u)} dy \end{array} \right) \text{--- } (*)$$

If $k_1 < u$,

$$L(u) = 1 + \int_{u-k_1}^h L(y) e^{-(y+k_1-u)} dy$$

$$L'(u) = \int_{u-k_1}^h L(y) e^{-y-k_1+u} dy - L(u-k_1)$$

$$= L(u) - 1 - L(u-k_1) \text{--- } (**)$$

(*) & (**) gives the desired result.

$$(b). \text{ Here, } k_1 = 1.5, h_1 = 4.0, m = 8, a_i = \frac{(2i-1)h}{2m}, w_i = \frac{h_1}{m} = 0.5,$$

$$f(x) = e^{-x}, F(x) = 1 - e^{-x} \quad \forall x > 0.$$

$$\text{Note } F(x) = 0 \text{ if } x \leq 0$$

$$f(x) = 0 \text{ if } x \leq 0.$$

Using the argument in Section 2.3 of the notes,

$$R = \begin{pmatrix} F(k_1 - a_1) + w_1 f(k_1) & w_2 f(a_2 + k_1 - a_1) & \dots & w_m f(a_m + k_1 - a_1) \\ F(k_1 - a_2) + w_1 f(a_1 + k_1 - a_2) & w_2 f(k_1) & \dots & w_m f(a_m + k_1 - a_2) \\ \vdots & \vdots & & \vdots \\ F(k_1 - a_m) + w_1 f(a_1 + k_1 - a_m) & w_2 f(a_2 + k_1 - a_m) & \dots & w_m \cdot f(k_1) \end{pmatrix}$$

cont'd

2.30

7

(b) cont'd =

$$R = \begin{pmatrix} (1-e^{-1.25}) + .5e^{-1.5} & .5e^{-2} & .5e^{-2.5} & .5e^{-3} & .5e^{-3.5} & .5e^{-4} & .5e^{-4.5} & .5e^{-5} \\ (1-e^{-.75}) + .5e^{-1} & .5e^{-1.5} & .5e^{-2} & .5e^{-2.5} & .5e^{-3} & .5e^{-3.5} & .5e^{-4} & .5e^{-4.5} \\ (1-e^{-.25}) + .5e^{-.5} & .5e^{-1} & .5e^{-1.5} & .5e^{-2} & .5e^{-2.5} & .5e^{-3} & .5e^{-3.5} & .5e^{-4} \\ 0 & .5e^{-.5} & .5e^{-1} & .5e^{-1.5} & .5e^{-2} & .5e^{-2.5} & .5e^{-3} & .5e^{-3.5} \\ 0 & 0 & .5e^{-.5} & .5e^{-1} & .5e^{-1.5} & .5e^{-2} & .5e^{-2.5} & .5e^{-3} \\ 0 & 0 & 0 & .5e^{-.5} & .5e^{-1} & .5e^{-1.5} & .5e^{-2} & .5e^{-2.5} \\ 0 & 0 & 0 & 0 & .5e^{-.5} & .5e^{-1} & .5e^{-1.5} & .5e^{-2} \\ 0 & 0 & 0 & 0 & 0 & .5e^{-.5} & .5e^{-1} & .5e^{-1.5} \end{pmatrix}$$

(c). $m=8$, $k_1=1.5$, $h_1=4.0$

$$f^*(x) = \text{density of } Q_1 - k_1 = Q_1 - 1.5 \\ = \begin{cases} e^{-(1.5+x)} & , x \geq -1.5 \\ 0 & , \text{ow.} \end{cases}$$

replace continuous dsn $f^*(x)$ by discrete dsn concentrating on multiples of $h_1/m = 4/8 = .5$ from $-h_1 = -4$ to $h_1 = 4$

$$P_{-8} = P_{-7} = P_{-6} = P_{-5} = P_{-4} = 0$$

$$P_{-3} = \int_{-1.5}^{-1.25} e^{-(1.5+x)} dx = -e^{-(1.5+x)} \Big|_{-1.5}^{-1.25} = 1 - e^{-.25} = .2212$$

$$P_{-2} = e^{-.25} - e^{-.75} = .3064$$

$$P_{-1} = e^{-.75} - e^{-1.25} = .1859$$

$$P_0 = e^{-1.25} - e^{-1.75} = .1127$$

$$P_1 = e^{-1.75} - e^{-2.25} = .0084$$

$$P_2 = e^{-2.25} - e^{-2.75} = .0415$$

$$P_3 = e^{-2.75} - e^{-3.25} = .0252$$

$$P_4 = e^{-3.25} - e^{-3.75} = .0153$$

$$P_5 = e^{-3.75} - e^{-4.25} = .0093$$

$$P_6 = e^{-4.25} - e^{-4.75} = .0056$$

$$P_7 = e^{-4.75} - e^{-5.25} = .0034$$

$$P_8 = \int_{3.75}^{\infty} e^{-(1.5+x)} dx = -e^{-(1.5+x)} \Big|_{3.75}^{\infty} = .0052$$

2.30

8

(c). cont'd

States = "no alarm to date & current high side CUSUM is $\hat{i}(.5)$, $\hat{i} = 0, \dots, 7$ "
and "alarm"

$$R = \begin{pmatrix} .8262 & .0684 & .0415 & .0252 & .0153 & .0093 & .0056 & .0034 \\ .7135 & .1127 & .0684 & .0415 & .0252 & .0153 & .0093 & .0056 \\ .5276 & .1859 & .1127 & .0684 & .0415 & .0252 & .0153 & .0093 \\ .2212 & .3064 & .1859 & .1127 & .0684 & .0415 & .0252 & .0153 \\ 0 & .2212 & .3064 & .1859 & .1127 & .0684 & .0415 & .0252 \\ 0 & 0 & .2212 & .3064 & .1859 & .1127 & .0684 & .0415 \\ 0 & 0 & 0 & .2212 & .3064 & .1859 & .1127 & .0684 \\ 0 & 0 & 0 & 0 & .2212 & .3064 & .1859 & .1127 \end{pmatrix}$$

This is quite similar to the R matrix in (b).

2.32

9

$$(a). \text{Var } X = E(X-EX)^2 = EX^2 - (EX)^2$$

$$E(X+1)^2 = EX^2 + 2EX + 1$$

$$(b). \text{Suppose } L_1 = E(RL)$$

$L_2 = E(AP)$ where $AP = \#$ additional points for a signal if there is no alarm yet & current point is between 2 & 3- σ limits.

$$M_1 = E(RL^2)$$

$$M_2 = E(AP^2)$$

Then, from section 2.2 of the notes,

$$M_1 = 1^2 \cdot q_1 + E(1+AP)^2 q_2 + E(1+RL)^2 (1-q_1-q_2)$$

$$= q_1 + (M_2 + 2L_2 + 1)q_2 + (M_1 + 2L_1 + 1)(1-q_1-q_2)$$

$$M_2 = 1 \cdot (q_1 + q_2) + E(1+RL)^2 (1-q_1-q_2)$$

$$= (q_1 + q_2) + (M_1 + 2L_1 + 1)(1-q_1-q_2)$$

$$(c). \text{Var}(RL) = ERL^2 - (ERL)^2$$

$$= M_1 - L_1^2$$