

EUROPA-UNIVERSITÄT VIADRINA



FRANKFURT (O)DER

CUSUMs and EMWAs – Favorites and Falsities

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Outline

1. Introduction
2. Connections between worst case and average scenarios
3. Roberts' ARL study
4. Lucas'/Saccucci's worst case ARLs
- (5. EWMA resembles the CUSUM for small λ)

1. Introduction

Change-point-model

sequence X_1, X_2, \dots with cdf $\{F_{\theta_{(i)}}\}_{i=1,2,\dots}$ and unknown integer $m =$ **change point** with

$$\theta_{(i)} = \begin{cases} \theta_0 & , i < m \\ \theta_1 & , i \geq m \end{cases} .$$

Control chart = detection, monitoring, or surveillance scheme

- transformation $\{X_i\}_{i=1,2,\dots,n} \rightarrow Z_n$ and
- stopping time $L = \inf \{n \in \mathbb{N} : Z_n \notin \mathcal{O}\}$, $\mathcal{O} = (-\infty, ucl], [lcl, ucl]$, or $[lcl, \infty)$.

Standard model

- independent $\{X_i\}$,
- $X_i \sim \mathcal{N}(\mu_{(i)}, \sigma^2 = 1)$,
- $\mu_{(i)} = \begin{cases} \mu_0 = 0 & , i < m \\ \mu_1 & , i \geq m \end{cases}$,
- $i < m$ – in control ..., $i \geq m$ – out of control state
- $E_m(\cdot)$ – expectation for change point at m
- Average Run Length – $ARL = E_\infty(L)$ and $ARL = E_1(L)$, respectively
 \rightsquigarrow denote ARL with \mathcal{L}_μ

One-sided Schemes

- CUSUM – Page (1954)

$$Z_n = \max \{0, Z_{n-1} + X_n - k\}, \quad Z_0 = z_0,$$

$$L = \inf \{n \in \mathbb{N} : Z_n > h\} \quad (k = (\mu_0 + \mu_1)/2)$$

- EWMA – Roberts (1959) (reflecting barrier – Waldmann (1986), Gan (1993))

$$Z_n = \max \{z_{\text{reflect}}^*, (1 - \lambda) Z_{n-1} + \lambda X_n\}, \quad Z_0 = z_0,$$

$$L = \inf \left\{ n \in \mathbb{N} : Z_n > c \sqrt{\lambda/(2 - \lambda)} \right\}, \quad z_{\text{reflect}}^* = z_r \sqrt{\lambda/(2 - \lambda)}$$

- GRSR – Girshick/Rubin (1952), Širjajev (1963/76), Roberts (1966)

$$Z_n = (1 + Z_{n-1}) \exp(X_n - k), \quad Z_0 = z_0,$$

$$L = \inf \{n \in \mathbb{N} : Z_n > g\} \quad (\exp[(\mu_1 - \mu_0) X_n - (\mu_1^2 - \mu_0^2)/2])$$

Two-sided Schemes

- CUSUM: coupling of 1-sided schemes
- Crosier-CUSUM (1986)

$$Z_n = \begin{cases} 0 & , C_n \leq k \\ (Z_{n-1} + X_n) \cdot \left(1 - \frac{k}{C_n}\right) & , C_n > k \end{cases}, \quad C_n \stackrel{n \geq 1}{=} |Z_{n-1} + X_n|,$$

$$L = \inf \{n \in \mathbb{N} : |Z_n| > h\}$$

- EWMA – Roberts (1959)

$$Z_n = (1 - \lambda) Z_{n-1} + \lambda X_n, \quad Z_0 = z_0, \quad L = \inf \left\{ n \in \mathbb{N} : |Z_n| > c \sqrt{\lambda/(2 - \lambda)} \right\}$$

- GRSR: Pollak/Siegmund (1985) – $Z_n = (Z_n^+ + Z_n^-)/2$
performs like coupling of 1-sided schemes $Z_n = \max\{Z_n^+, Z_n^-\} \rightsquigarrow$ coupling
(PS: $\mathcal{L}_1 = 11.142$, $E(D_*) = 9.644$, coupling: $\mathcal{L}_1 = 11.142$, $E(D_*) = 9.630$, both: $\mathcal{L}_0 = 500$)

Statements about performance

- Lorden (1971), Moustakides (1986), Ritov (1990): CUSUM is optimal
- Pollak (1987): GRSR is asymptotically optimal
- Lucas/Saccucci (1990): *Our comparisons showed that the ARL's for the EWMA are usually smaller than the ARL's of CUSUM up to a value of the shift near the one that the scheme was designed to detect.*
- different authors: little practical difference between ...

2. Worst case and average scenarios

Markov-type schemes

$$\mathcal{F}_{m-1} = \sigma(X_1, \dots, X_{m-1}), \quad \mathcal{F}_{m-1}^* = \sigma(\mathcal{F}_{m-1} \cap \{L > m - 1\}),$$

$$D_*^{(m)} = E_m(L - m + 1 \mid \mathcal{F}_{m-1}^*) = \mathcal{L}_{\mu_1}(Z_{m-1}), \quad Z_{m-1} \sim f_{m-1}^*(\cdot)$$

$$E_m([L - m + 1]^+ \mid \mathcal{F}_{m-1}) = D_*^{(m)} \cdot I_{\{L > m-1\}} + 0 \cdot I_{\{L \leq m-1\}},$$

$$\mathcal{D}_{\text{Lorden}} = \sup_{m \geq 1} \text{ess sup} (D_*^{(m)}), \quad \mathcal{D}_{\text{Pollak/Siegmund}} = \sup_{m \geq 1} E(D_*^{(m)}),$$

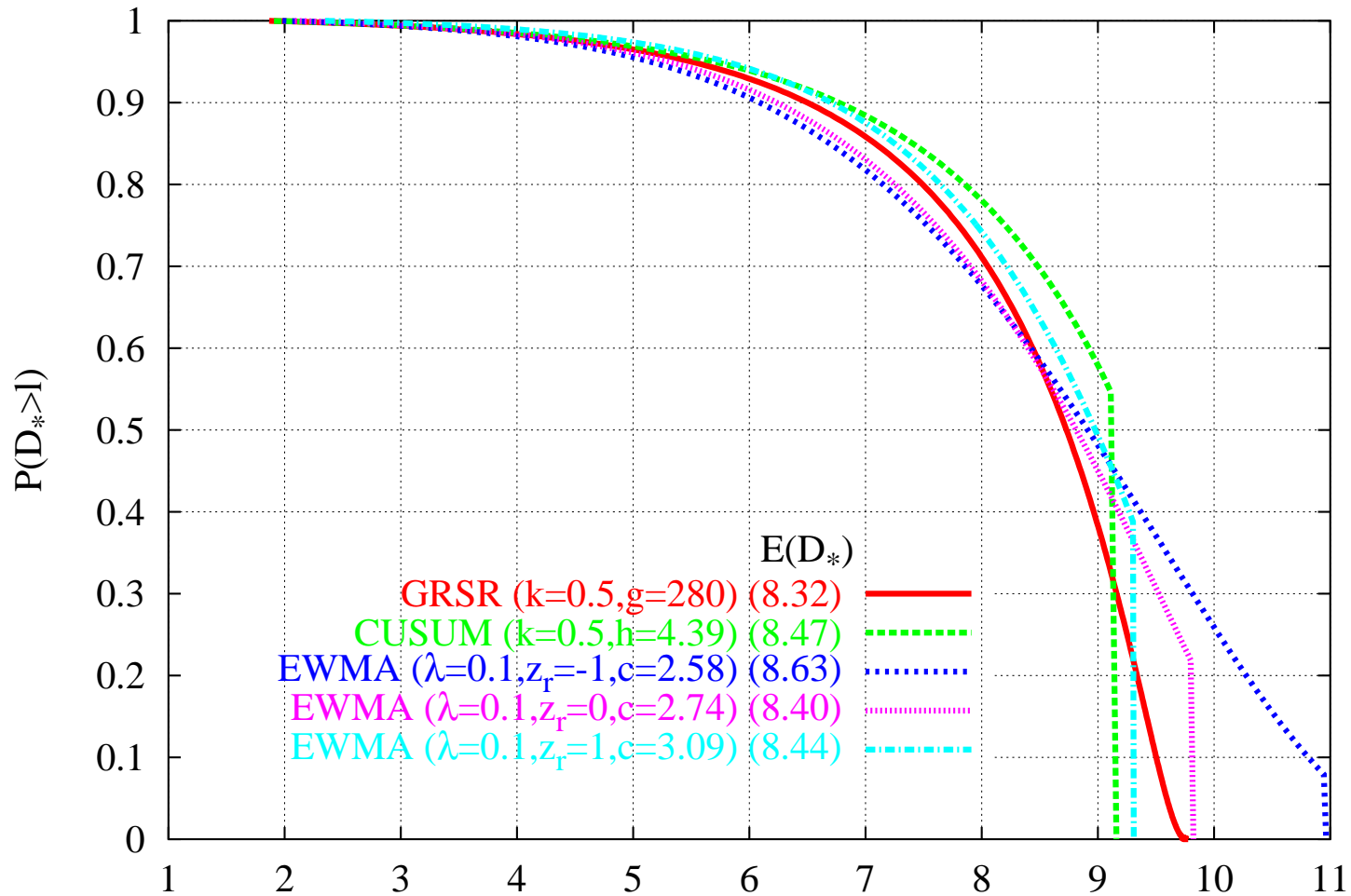
$$D_* = \lim_{m \rightarrow \infty} D_*^{(m)}, \quad D_*^{(m)} \approx D_* \text{ for } m \geq m_* \text{ (small)}$$

$$E(D_*) = \int_{\mathcal{O}} \psi_{\mu_0}(z) \mathcal{L}_{\mu_1}(z) d\mathcal{M}(z),$$

$$\text{whereby } \psi_{\mu_0}(z) = \lim_{m \rightarrow \infty} f_{m-1}^*(z) \quad \text{with} \quad f_{m-1}^*(z) := \frac{f_{m-1}(z)}{\int_{\mathcal{O}} f_{m-1}(z) d\mathcal{M}(z)}$$

One-sided Schemes – survival function $P(D_* > l)$

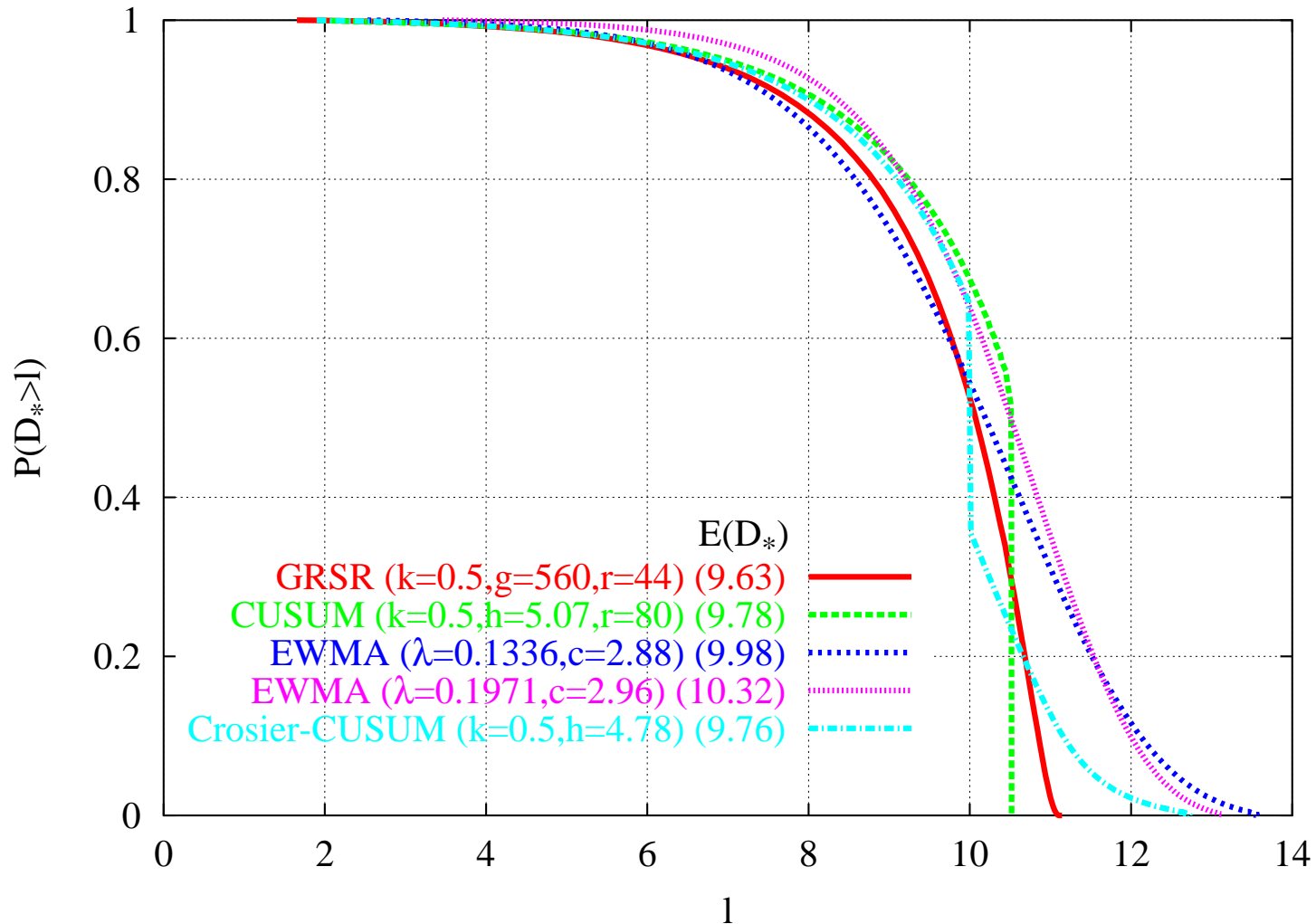
$$E_\infty(L) = 500, \mu_1 = 1$$



ess sup (D_*) : GRSR – 9.78, CUSUM – 9.16, EWMA₋₁ – 10.96 (9.13), EWMA₀ – 9.82, EWMA₁ – 9.31

Two-sided Schemes – survival function $P(D_* > l)$

$$E_\infty(L) = 500, \mu_1 = 1$$



$\text{ess sup}(D_*)$: GRSR – 11.14, CUSUM – 10.52, EWMA₁ – 13.54 (10.19), EWMA₂ – 13.14 (10.52), Crosier – 12.73 (10.01)

3. Roberts did a study based on the ARL criterion

(one-sided schemes)

chart/scheme	design ($\mathcal{L}_0 = 740$)	criterion	computation method
Shewhart	$c = 3$	$\mathcal{L}_\mu = E(D_*)$	exact ($\mathcal{L}_0 = 741.43$)
MA	$n = 8, c = 2.79$	$E(D_*^{(9)})$	Monte Carlo, 25 000 repetitions
EWMA (GMA)	$\lambda = 0.25, c = 2.87$	$E(D_*^{(9)})$	Monte Carlo, 25 000 repetitions
CUSUM	$k = 0.47, h = 5$	\mathcal{L}_μ	"translation" of two-sided results of Ewan/Kemp (1960)
GRSR	$k = 0.5, g = 390$	$E(D_*^{(9)})$	Monte Carlo, 50 000 repetitions

Remarks of Roberts

- **CUSUM:** *These curves all assume $T = 0$, and are upper bounds for those for $T = 8$, say, though the differences are small except for $C_{26.8}(0)$, ...*

$$(T = m - 1; C_{26.8}(0) - h = 26.8, k = 0)$$

- **MA/EWMA:** *By allowing the shifts to occur after eight observations with $\mu_i = \mu_0$ (i. e., $T = 8$), complications due to changing limits were averted.*

$$(\text{EWMA: } \sqrt{\text{Var}(Z_9)/[\lambda/(2 - \lambda)]} = \sqrt{1 - (1 - \lambda)^{2.9}} \underset{\lambda=0.25}{=} 0.997 \approx 1)$$

Roberts – original results and recent values

scheme	Δ								
	0	0.5	1.0	1.5	2.0	2.5	3.0	3.5	
$n = 8$	740	40	10.2	6.0	4.6	3.8	3.3	2.9	Roberts
MA	737	39	10.0	5.8	4.4	3.7	3.1	2.8	$E(D_*^{(9)})/E(D_*)$
$c = 2.79$	736	39	8.9	4.2	2.6	1.9	1.5	1.2	$E_1(L)$
$\lambda = 0.25$	740	40	10.1	5.1	3.5	2.7	2.2	1.9	Roberts
EWMA	686	39	9.8	5.0	3.4	2.6	2.1	1.8	$E(D_*^{(9)})/E(D_*)$
$c = 2.87$	681/689	37/39	8.4/10.0	3.9/5.1	2.5/3.4	1.9/2.6	1.5/2.2	1.3/1.9	$E_1(L)$
$k = 0.47$	740	34	10.0	5.8	4.3	3.5	2.9	2.5	Roberts
CUSUM	(719)/718	32	9.2/9.1	5.1	3.6	2.8	2.4/2.3	2.0	$E(D_*^{(9)})/E(D_*)$
$h = 5$	724	34	9.9	5.6	3.9	3.1	2.5	2.2	$E_1(L)$
$k = 0.5$	740	32	9.2	5.2	3.7	2.9	2.4	2.1	Roberts
GRSR	(690)/689	30	9.1/8.9	5.2/5.1	3.7/3.6	2.9/2.8	2.4	2.1	$E(D_*^{(9)})/E(D_*)$
$g = 390$	697	33	10.4	6.1	4.4	3.5	2.9	2.5	$E_1(L)$

4. Wrong worst case ARLs in Lucas/Saccucci 1990

(two-sided schemes)

<i>shift</i>	<i>Zero state</i>			<i>worst case</i>		CUSUM* $k = 0.4, h = 6$
	EWMA ₁	EWMA ₂	CUSUM	EWMA ₁	EWMA ₂	
0.00	465	465	465	310 (?)	315 (?)	470
0.25	116 ₁₁₄	118 ₁₁₇	139 ₁₃₇	97.6 (120)	100 (122)	120 ₁₁₆
0.50	33.3 _{32.6}	33.8 _{33.2}	38.0 _{36.5}	34.2 (37.6)	34.7 (38.0)	33.6 _{31.6}
0.75	16.0 _{15.6}	16.1 _{15.8}	17.0 _{16.0}	19.1 (19.8)	19.1 (19.8)	16.5 _{15.2}
1.00	10.1 _{9.85}	10.0 _{9.85}	10.4 _{9.65}	13.3 (13.4)	13.2 (13.3)	10.6 _{9.66}
1.50	5.71 _{5.62}	5.67 _{5.58}	5.75 _{5.29}	8.43	8.32	6.19 _{5.59}

EWMA₁ – $\lambda = 0.133, c = 2.856,$

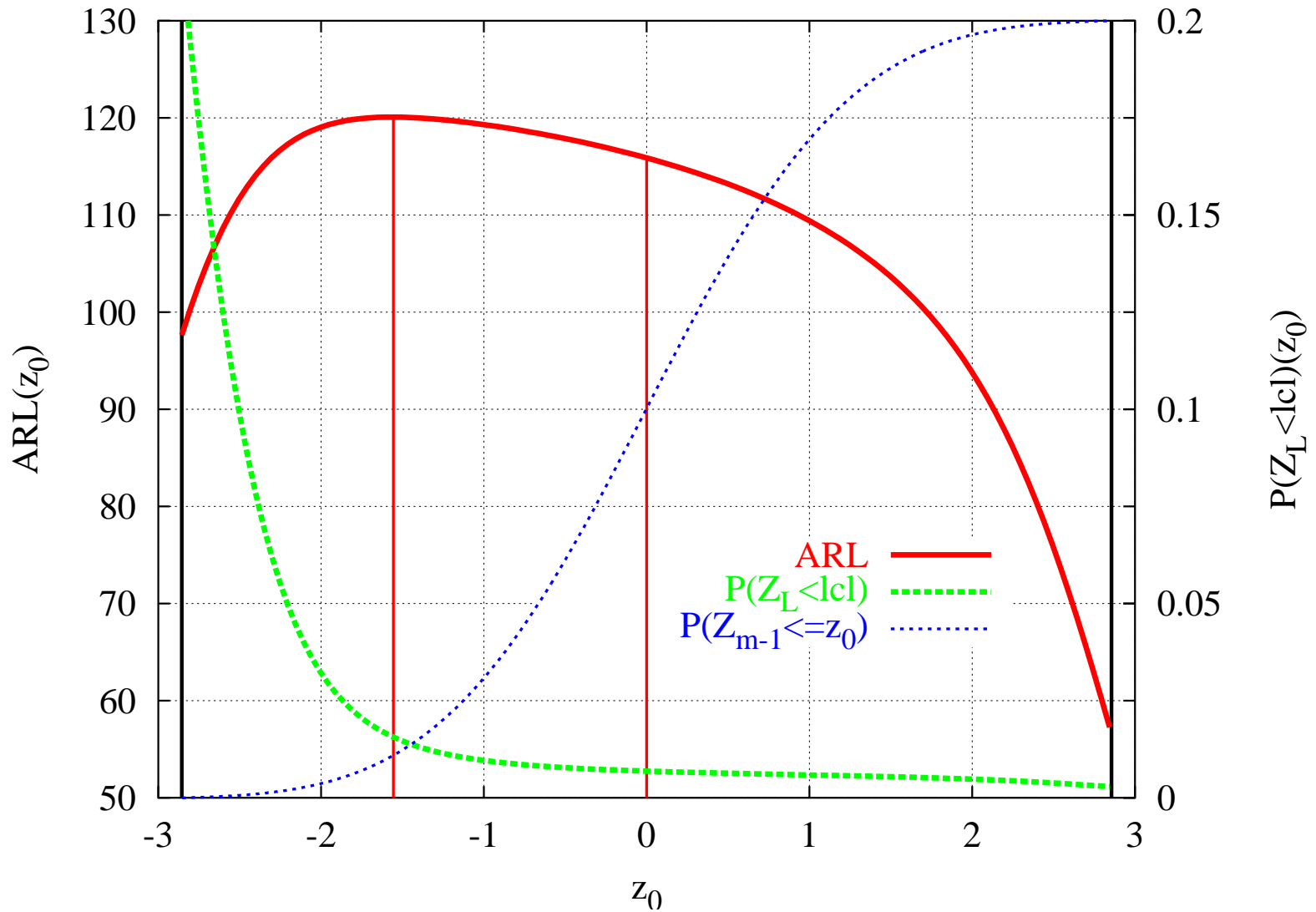
EWMA₂ – $\lambda = 0.139, c = 2.866,$

CUSUM – $k = 0.5, h = 5$

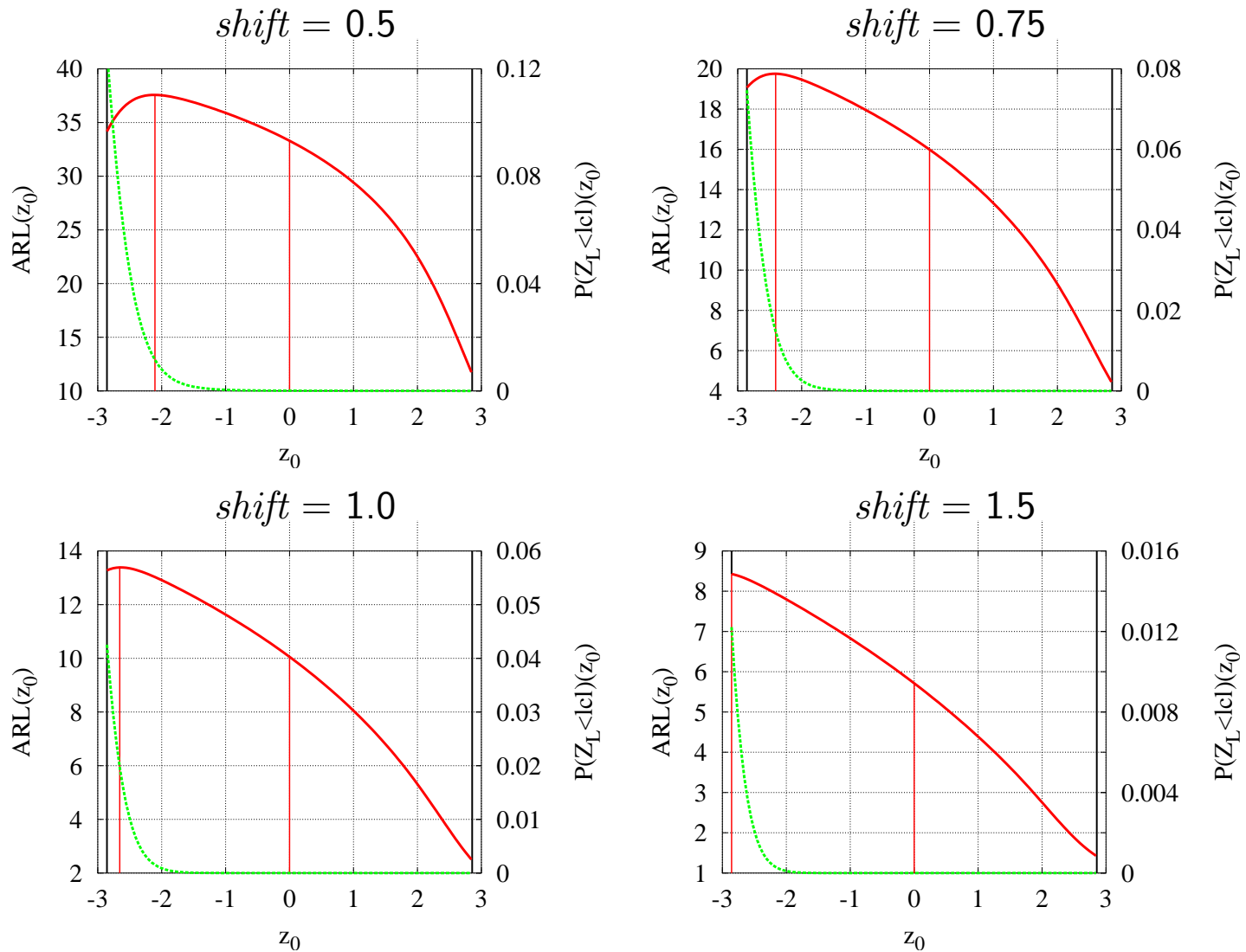
tiny values – $E(D_*)$

Worst case EWMA ARLs I

shift = 0.25

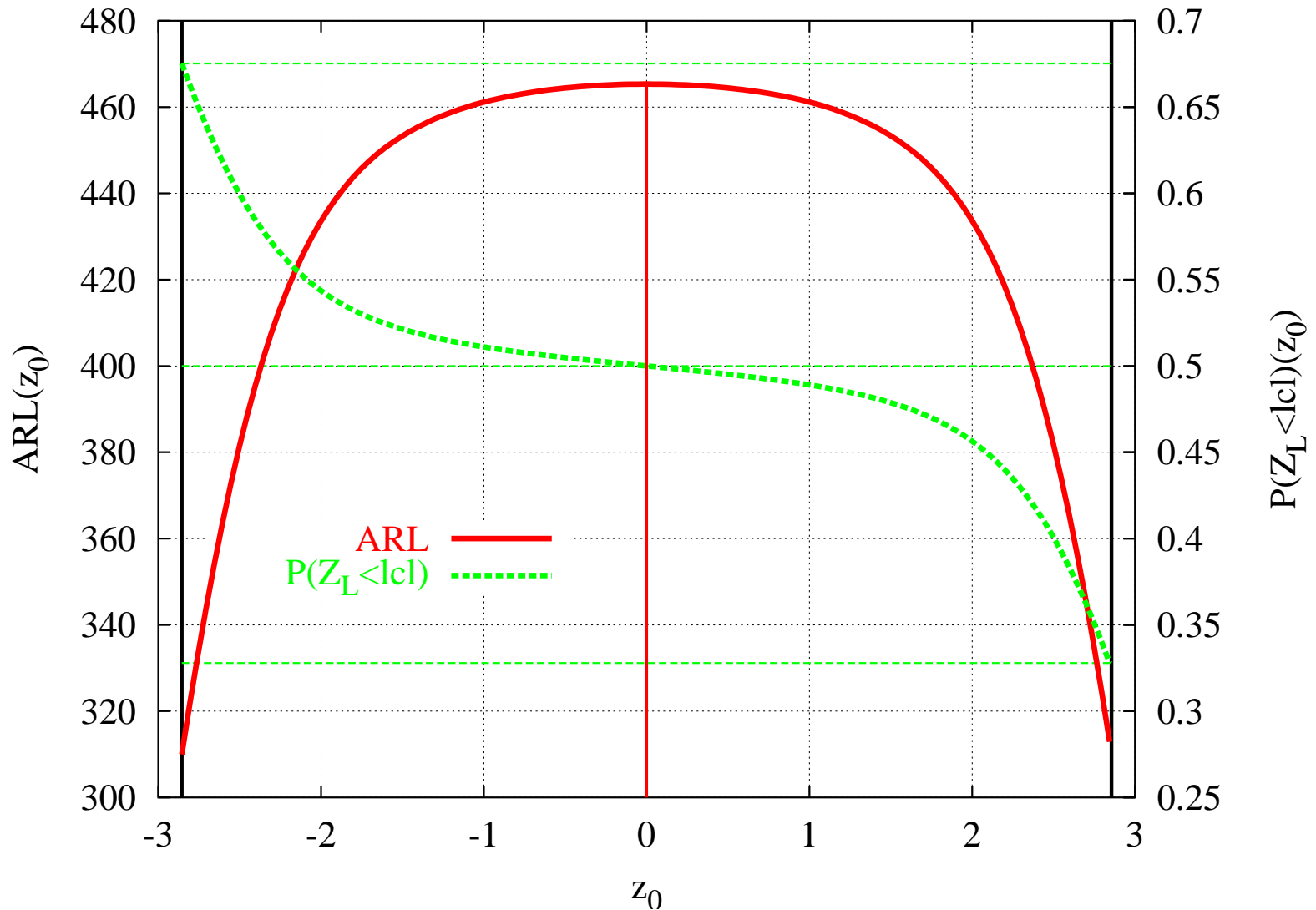


Worst case EWMA ARLs II



Worst case EWMA ARLs III

shift = 0.0 – in control



5. EWMA resembles the CUSUM for small λ

SAS Online-Manual (SAS 8.0, February 2000)

EWMAARL(δ, r, k)

r is the weight factor for the current subgroup mean in the EWMA, where $0 < r \leq 1$. If $r = 1$, the EWMAARL function returns the average run length for a Shewhart chart for means. Refer to Wadsworth and others (1986). If $r \leq 0.05$, $k \geq 3$, and $\delta < 0.10$, the algorithm used is unstable. However, note that the EWMA behaves like a cusum when $r \rightarrow 0$, and in this case the CUSUMARL function is applicable.

HUNTER, J. S. (1986) *The exponentially weighted moving average*, JQT **18**, 203-210.

Thus, as $\lambda \rightarrow 0$, the EWMA takes on the appearance of the CUSUM. The EWMA control chart for values of $0 < \lambda < 1$ stands between the Shewhart and CUSUM control charts in its use of the historical data.

p. 207

EWMA resembles the CUSUM for small λ ?

YES

NO

CROWDER, S. V. (1987)

WOODALL, W. H. & MARAGAH, H. D. (1990)

MITTAG, H. J. (1993)

RINNE, H. & MITTAG, H. J. (1995)

\approx **NO**

GOHOUT, W. (1996)

CROWDER, S. V. (1989)

LEDOLTER, J. (1999)

HUNTER, J. S. (1990)

STEMANN, D. (1997)

SAS (2000)

no comment

STEMANN, D. & WEIHS, C. (2001)

many others

Hunter (1990)

HUNTER, J. S. (1990) *Discussion of Lucas & Saccucci*, *Technometrics*, **32**, p. 21-22.

Each observation in the CUSUM then takes weight $1/n_i$, where i is the index number of the trial. Of course, n_i varies randomly from trial to trial.

Consider $E(n_i) = ASN$ and $\sqrt{Var(n_i)} = SD(SN)$ and

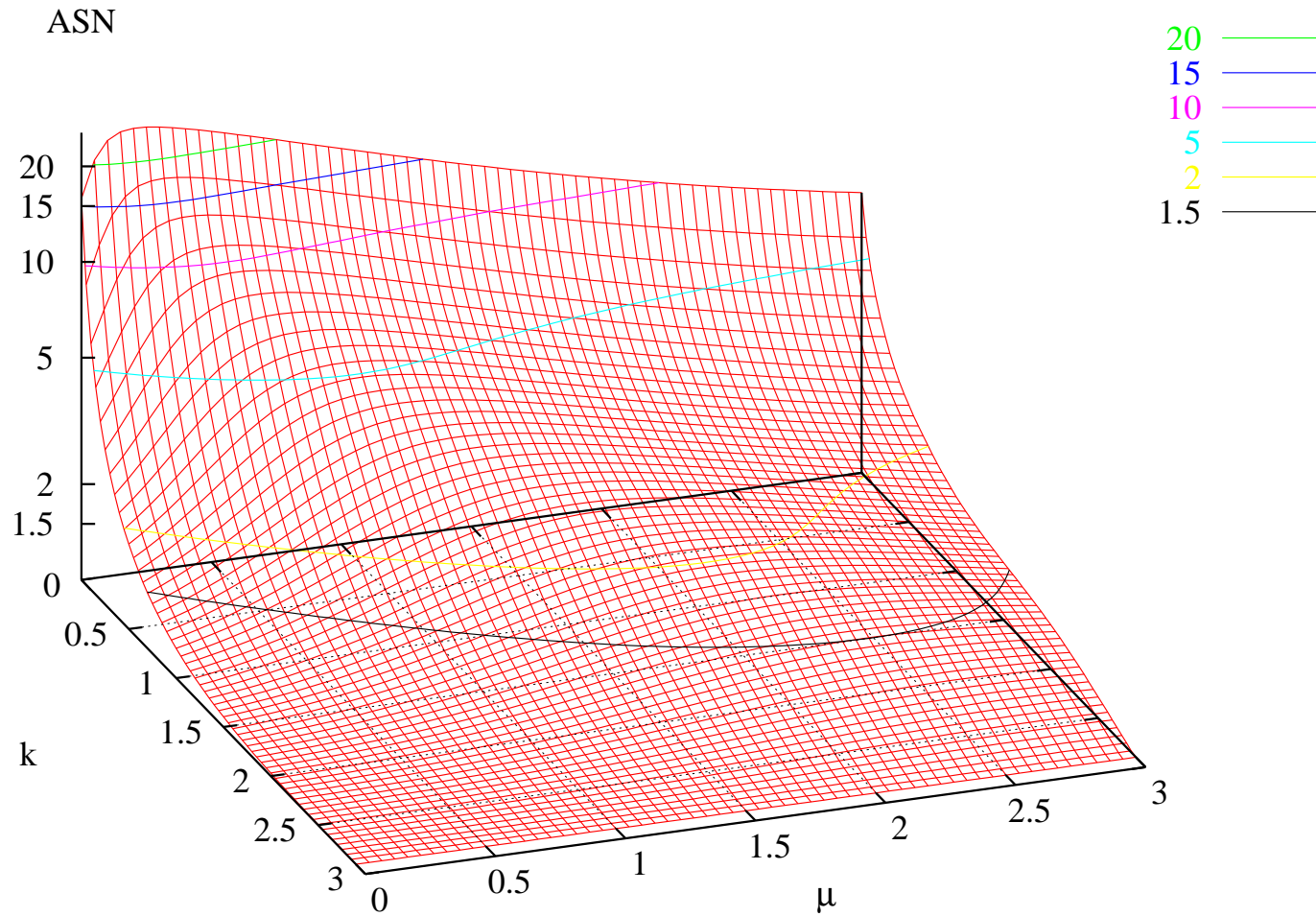
remember that $\mathcal{L}_\mu(z) = ASN_\mu(z) + OC_\mu(z) \frac{ASN_\mu(0)}{1-OC_\mu(0)}$

Example: one-sided CUSUM with $E_\infty(L) = 500$

\rightsquigarrow "natural" domain of reference value k — $\left[0, 2.8782 = 1/(1 - \Phi^{-1}(1/500)) \right]$

Average Sample Number of (one-sided) CUSUM trials

$$E_{\infty}(L) = 500$$



SD of the Sample Number of (one-sided) CUSUM trials

$$E_{\infty}(L) = 500$$

