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FRANKFURT (ODER)

Fast initial response features for EWMA Control Charts

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Usual EWMA scheme for the mean (with fixed control limits)

$$X_n \sim \mathcal{N}(\mu_n, \sigma^2 = 1), \quad \mu_n = \begin{cases} 0 & , n < m \\ \delta (= 1) & , n \geq m \end{cases}, \quad n = 1, 2, \dots$$

$$Z_0 = z_0 = 0, \quad Z_n \underset{n \geq 1}{=} (1 - \lambda) Z_{n-1} + \lambda X_n, \quad \lambda \in (0, 1],$$

$$E_m(Z_n) \underset{n < m}{=} 0 \quad (E_\infty(Z_n) = 0 \text{ for all } n, \text{ for } z_0 \neq 0 \text{ at least } \lim_{n \rightarrow \infty} E_\infty(Z_n) = 0),$$

$$\text{Var}_m(Z_n) = \frac{2}{2 - \lambda} (1 - [1 - \lambda]^{2n}) \rightarrow \frac{2}{2 - \lambda} \text{ for } n \rightarrow \infty,$$

$$L = \inf \left\{ n \in \mathbb{N} : |Z_n| > c \sqrt{\frac{2}{2 - \lambda}} \right\}.$$

Fast Initial Response (FIR)

Aim: Improve control chart performance at scheme start-up!

e. g.,

- following a process adjustment in quality control,
- after re-estimating the underlined model in monitoring financial data,
- one suspects a change before monitoring is started,
- weaken the influence of the initial setup (z_0) of the usual EWMA scheme,
- ...

Fast Initial Response – revisiting literature

- For the first time considered by Lucas/Crosier (1982) for CUSUM.
- Lucas/Saccucci (1990 ... 1988) Markov chain
5.1 FIR Feature ... A disadvantage of this approach is that it requires two separate EWMA's ... A detailed discussion of this approach will be given in a future article.
B.1 FIR Feature ... The FIR feature requires the simultaneous implementation of two one-sided EWMA ... We plan to implement the exact method for calculating FIR ARL's in our future work.
- Rhoads/Montgomery/Mastrangelo (1996), Monte-Carlo
Fast initial response scheme for exponentially weighted moving average control chart.
- Steiner (1999), Markov chain
EWMA control charts with time-varying control limits and fast initial response.
- Montgomery (2001): $\frac{1}{2}$ page in *Introduction to Statistical Quality Control, 4th ed.*

FIR as remedy for inertia

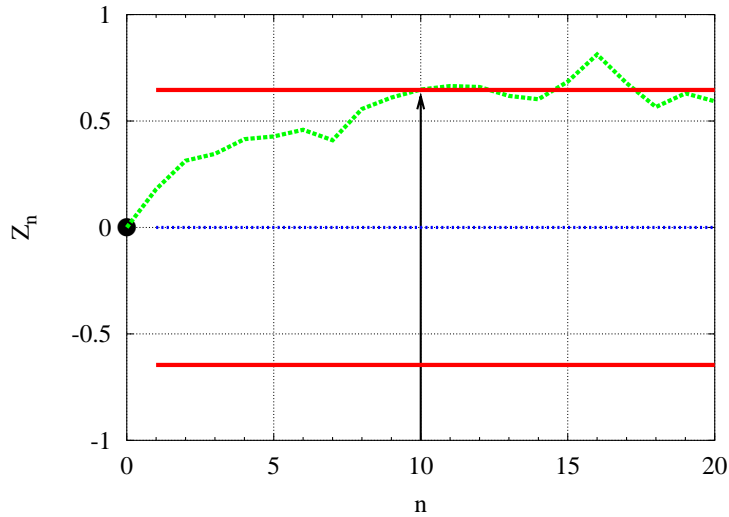
Problem: inert behavior of the usual EWMA scheme for early changes (small m), i. e. strong influence of the starting value z_0 at early time points if λ is small

$$\text{Var}_m(Z_n) = \frac{2}{2-\lambda} (1 - [1-\lambda]^{2n}) \ll \frac{2}{2-\lambda} = \lim_{n \rightarrow \infty} \text{Var}_m(Z_n)$$

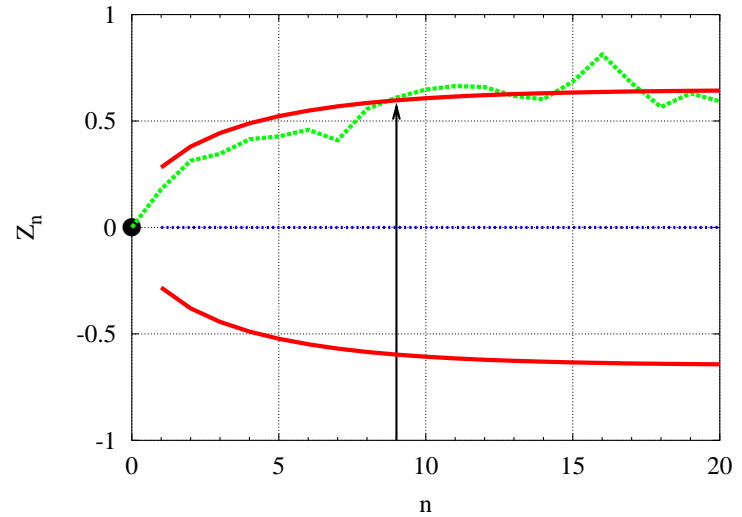
Motivating examples: $\lambda = 0.1$, $m = 1$, $E_\infty(L) = 500$

FIR EWMA examples I – $m = 1$

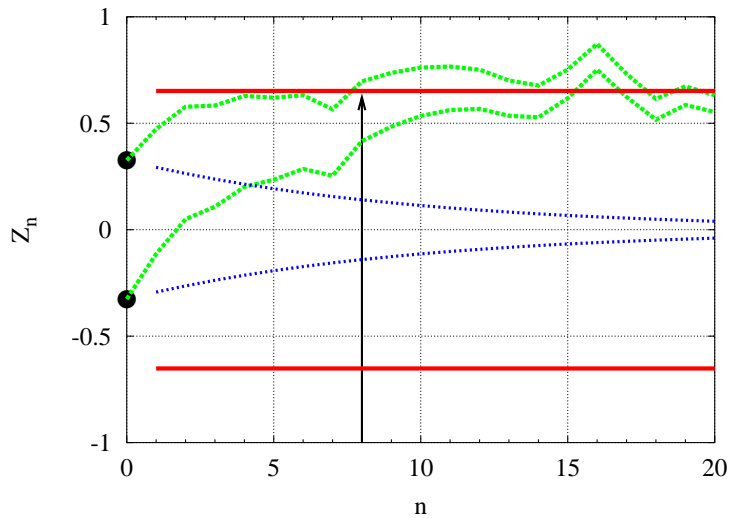
Roberts (1959), Wortham (1972), Ho (1978)



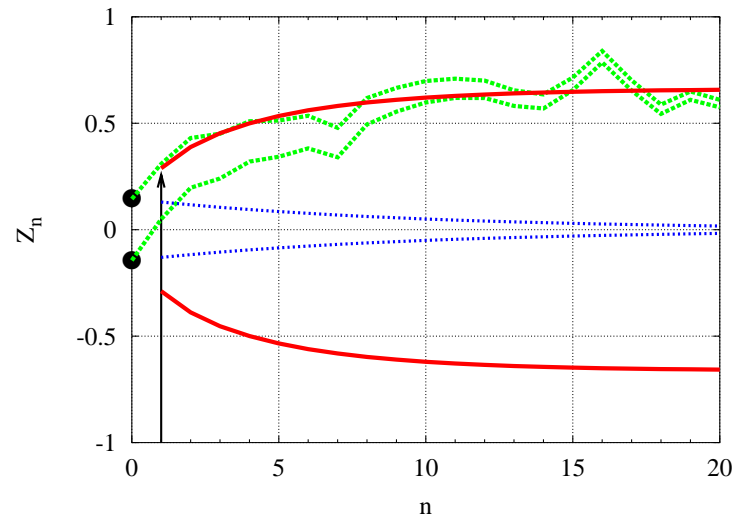
Roberts (1959), Chandrasekaran et al. (1995)



Lucas and Saccucci (1990)

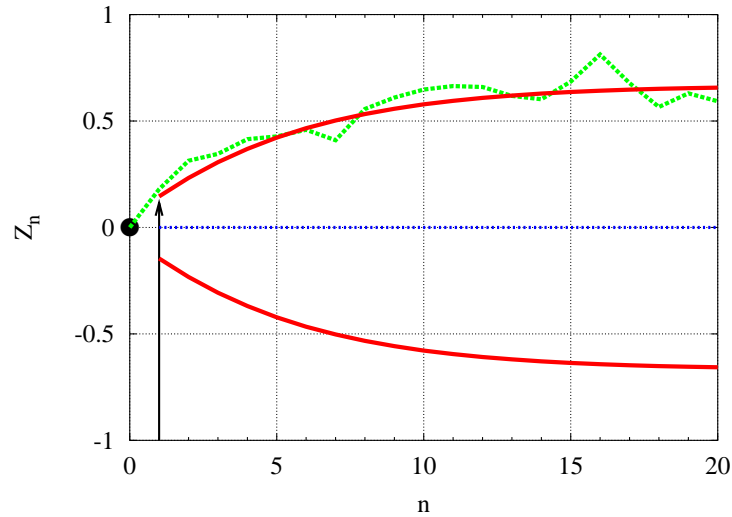


Rhoads et al. (1996)

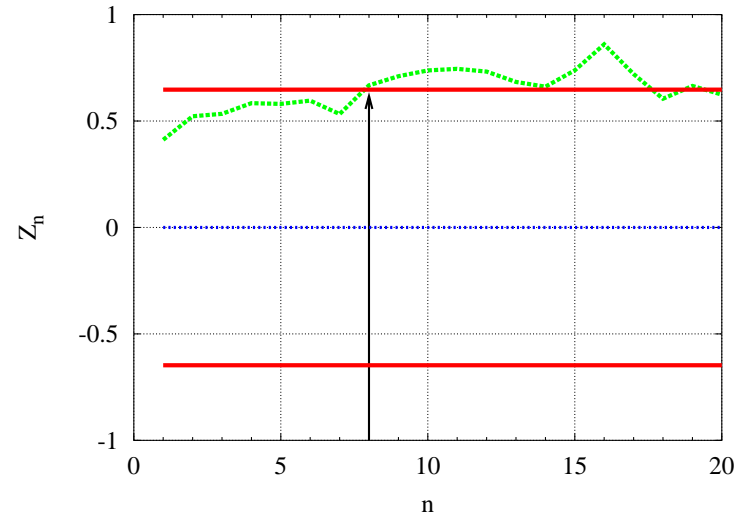


FIR EWMA examples II – $m = 1$

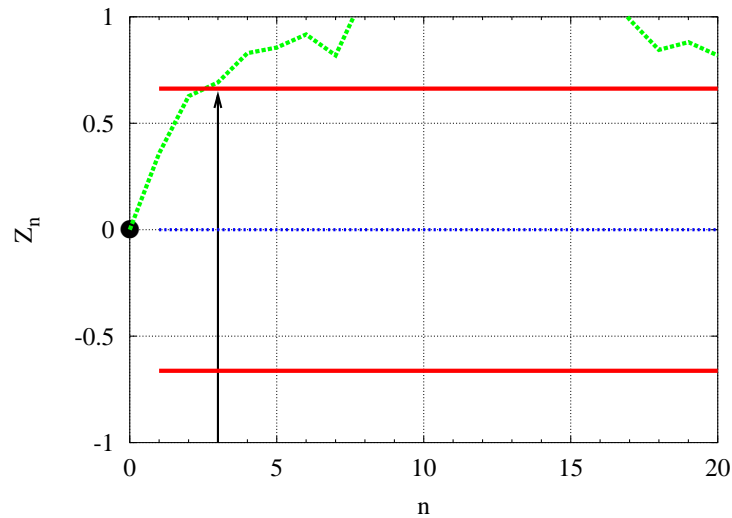
Steiner (1999)



(Knoth 2003)



Fink (1994)



?

?

Formulas

$$Z_0 = 0, \quad Z_n \underset{n \geq 1}{=} (1 - \lambda) Z_{n-1} + \lambda X_n, \quad c^* = c \sqrt{\frac{2}{2 - \lambda}}, \quad l_n = \sqrt{1 - (1 - \lambda)^{2n}}$$

raise alarm, if

$$(fcl) \quad |Z_n| > c^*,$$

$$(vacl) \quad |Z_n| > c^* l_n,$$

$$(fir) \quad \max \{ |Z_n^u|, |Z_n^l| \} > c^*, \quad Z_0^u = -Z_0^l = \frac{c^*}{2}, \quad \text{Lucas/Saccucci 1990}$$

$$(fvac) \quad \max \{ |Z_n^u|, |Z_n^l| \} > c^* l_n, \quad Z_0^u = -Z_0^l = l_1 \cdot \frac{c^*}{2}, \quad \text{Rhoads et al. 1996}$$

$$(fadj) \quad |Z_n| > c^* l_n \left(1 - (1 - f)^{1+a(n-1)} \right), \quad f = 0.5, \quad a = 0.3, \quad \text{Steiner 1999}$$

$$(stat) \quad |Z_n| > c^*, \quad Z_1 = X_1 \sqrt{2/(2 - \lambda)}, \quad Var_m(Z_n) = \frac{2}{2 - \lambda}, \quad (\text{Knoth 2003})$$

$$(switch) \quad |Z_n| > c^*, \quad \lambda = \begin{cases} \lambda_0 & , n \leq n_1 \\ \lambda_1 < \lambda_0 & , n > n_1 \end{cases}, \quad n_1 = 10, \lambda_1 = \lambda_0/2. \quad \text{Fink 1994}$$

FIR due to Lucas/Saccucci needs two charts?!

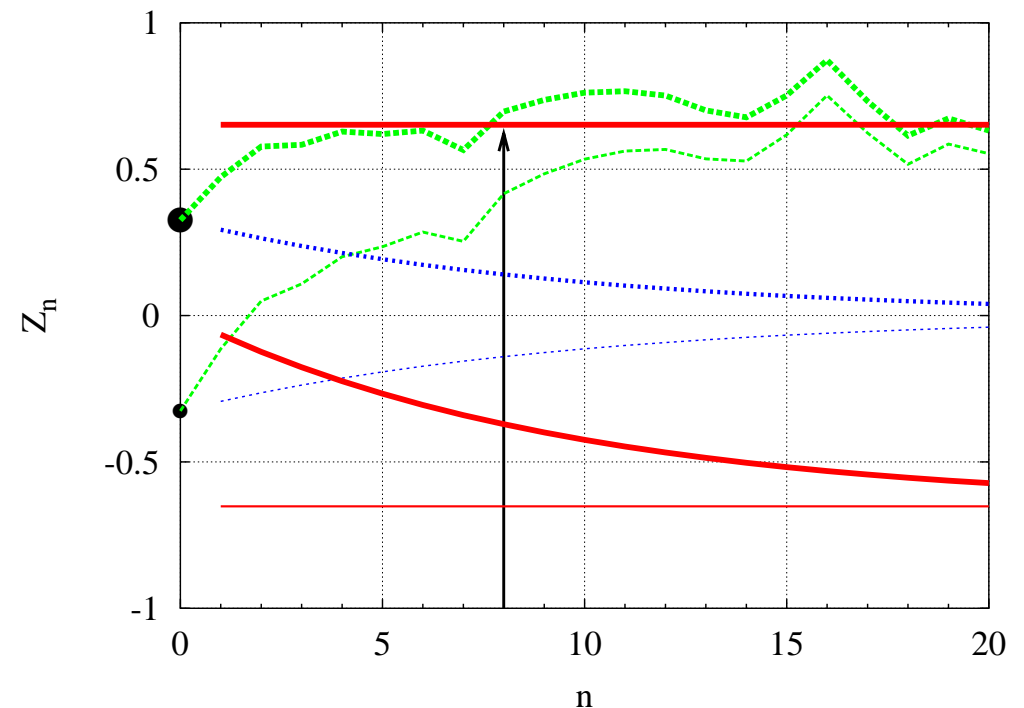
$$Z_0^u = -Z_0^l = \frac{c^*}{2},$$

$$\begin{aligned} Z_n^l &= Z_n^u - (Z_0^u - Z_0^l)(1 - \lambda)^n \\ &= Z_n^u - c^*(1 - \lambda)^n \end{aligned}$$

Therefore, ($L = \inf \{n \in \mathbb{N} : \dots\}$)

$$\begin{aligned} \max \{|Z_n^u|, |Z_n^l|\} > c^* &\Leftrightarrow \max \{Z_n^u, -Z_n^l\} > c^* \\ &\Leftrightarrow Z_n^u \notin [-(1 - (1 - \lambda)^n)c^*, c^*] \end{aligned}$$

Remark: Similar equivalence is valid for FIR due to Rhoads et al.



Computation of $E_m(L)$, critical value c , and ...

- For fixed limits (homogeneous) schemes (fcl, stat) a lot of methods available, for switch easy adaption possible, for fir at least the two-dimensional approach could be exploited (Lucas/Saccucci).
- For EWMA schemes with non-constant limits or varying transition probabilities literature offers:
 - modified Markov chain approaches, e. g., in Chandrasekaran et al. (1995) and Steiner (1999),
 - Monte-Carlo, e. g., in Rhoads et al. (1996),
 - density recursion method in Margavio, Conerly, Woodall, Drake (1995)

false alarm rate $r_n = P_\infty(L = n | L \geq n)$ for fcl, vac1 and a constant false alarm rate EWMA scheme

Density recursion I

A more general notation is $L = \inf \{n \in \mathbb{N} : Z_n \notin [a_n, b_n]\}$.

Consider the following recursion:

$$f_1(z; z_0) = \begin{cases} \frac{1}{\lambda} \phi_\mu \left(\frac{z - (1-\lambda)z_0}{\lambda} \right) & , a_1 \leq z \leq b_1 \\ 0 & , \text{otherwise} \end{cases} ,$$

$$f_n(z; z_0) = \begin{cases} \int_{a_{n-1}}^{b_{n-1}} f_{n-1}(\tilde{z}; z_0) \frac{1}{\lambda} \phi_\mu \left(\frac{z - (1-\lambda)\tilde{z}}{\lambda} \right) d\tilde{z} & , a_n \leq z \leq b_n \\ 0 & , \text{otherwise} \end{cases} .$$

Then $f_n^*(z; z_0) = f_n(z; z_0) / \int_{a_n}^{b_n} f_n(\tilde{z}; z_0) d\tilde{z}$

is a density of the EWMA statistic Z_n conditioned on $L > n$ and $Z_0 = z_0$,

where the denominator is equal to $P(L > n)$.

Density recursion II

Based on $P(L > n)$ one can obtain

$$\text{zero-state ARL} \quad E_{1(\infty)}(L) = \sum_{n=0}^{\infty} P_{1(\infty)}(L > n),$$

$$D_m = E_m(L - m + 1 | L \geq m) = \frac{\sum_{n=m}^{\infty} P_m(L > n)}{P_m(L > m - 1)},$$

$$\text{steady-state ARL} \quad D = \lim_{m \rightarrow \infty} D_m \approx D_{m_0}, \quad m_0 = 100,$$

$$P_m(L = n), \quad P_m(L = n + m | L \geq m), \quad L_\alpha, \quad r_n, \dots,$$

$$c \text{ as solution of } E_\infty(L) = A = 500.$$

Density recursion III

If $\lim_{n \rightarrow \infty} a_n = a$ and $\lim_{n \rightarrow \infty} b_n = b$,

then the tails of the stopping time L behave like those of a geometric distribution,

$$\text{i. e. } P_m(L > n + j) \approx \varrho^j P_m(L > n)$$

with ϱ as eigenvalue with largest magnitude of the limit transition kernel.

\rightsquigarrow use quadrature based methods until a certain n_0 and exploit the geometric tail,

$$\text{e. g., } E_\infty(L) \approx \sum_{i=0}^{n_0-1} P(L > n) + \frac{P(L > n_0)}{1 - \tilde{\varrho}}, \quad \tilde{\varrho} = \frac{P(L > n_0)}{P(L > n_0 - 1)}.$$

Remarks:

1. For homogeneous kernels Woodall (1983), Waldmann (1986) and Gan (1991) used the same ideas (exception D_m and D).
2. It is not clear, how in Margavio et al. (1995) the problem with the varying limits in the integration sequence is solved and why, e. g., Steiner (1999) did neither use nor even mention their results.

Density recursion IV

$$-a_n = b_n = c^* \quad \text{fcl, stat, switch,}$$

$$-a_n = b_n = c^* \sqrt{1 - (1 - \lambda)^{2n}} = c^* l_n \quad \text{vacl,}$$

$$a_n = -(1 - (1 - \lambda)^n) c^*, \quad b_n = c^* \quad \text{fir,}$$

$$a_n = -(1 - (1 - \lambda)^n) c^* l_n, \quad b_n = c^* l_n \quad \text{fvac1,}$$

$$-a_n = b_n = c^* l_n \left(1 - (1 - f)^{1+a(n-1)}\right) \quad \text{fadj.}$$

$$\lim_{n \rightarrow \infty} -a_n = \lim_{n \rightarrow \infty} b_n = c^* .$$

$$r_n = (b_n - a_n)/(2c^*), \quad d_n = (a_n + b_n)/2,$$

$$\begin{aligned} M_n(\tilde{z}, z) &= r_n M(d_{n-1} + r_{n-1}\tilde{z}, d_n + r_n z) \\ &= \frac{r_n}{\lambda} \phi_\mu \left(\frac{d_n + r_n z - (1 - \lambda)(d_{n-1} + r_{n-1}\tilde{z})}{\lambda} \right) . \end{aligned}$$

\rightsquigarrow smooth and positive (transition) kernel function on $[-c^*, c^*] \times [-c^*, c^*]$,

i. e. use abscissas independent on n and get simult. precise quadrature results

Critical values

in-control ARL, i. e. $E_{\infty}(L) = 500$, $\lambda = 0.1$

EWMA scheme	fcl	vacl	fir	fvacl	fadj	stat	switch
c	2.8143	2.8239	2.8415	2.8858	2.9131	2.8215	2.8879
$\widehat{E_{\infty}(L)}$, 10^9 rep.	499.99	500.02	500.01	499.92	500.02	500.01	499.94
se	.016	.016	.017	.017	.020	.016	.019
density rec.	499.99	500.04	499.99	499.93	500.04	499.99	499.97

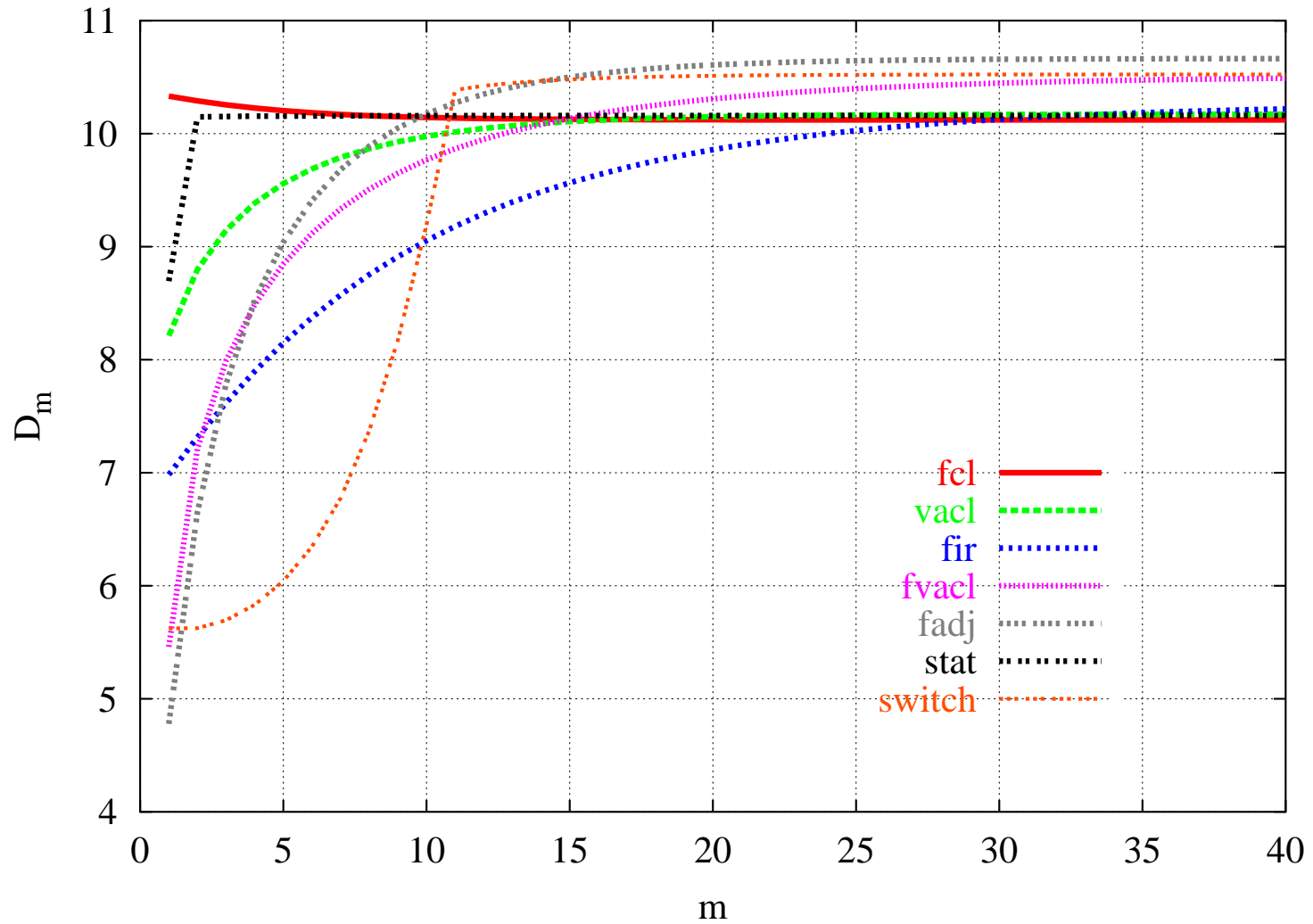
Zero-state and steady-state ARLs $E_1(L)$ and D

$\lambda = 0.1$, in-control ARL $E_\infty(L) = 500$, results as $\begin{pmatrix} E_1(L) \\ D \end{pmatrix}$

δ	EWMA scheme						
	fcl	vacl	fir	fvacl	fadj	stat	switch
0.5	<u>31.3</u>	28.8	24.8	22.9	21.6	29.3	<u>20.8</u>
	<u>30.6</u>	30.9	31.4	32.8	<u>33.6</u>	30.8	32.8
1.0	<u>10.3</u>	8.21	6.98	5.46	<u>4.78</u>	8.69	5.62
	<u>10.1</u>	10.2	10.3	10.5	<u>10.7</u>	10.2	10.5
1.5	<u>6.08</u>	4.17	3.90	2.52	<u>2.19</u>	4.56	3.36
	<u>5.99</u>	6.01	6.06	6.17	<u>6.24</u>	6.01	6.18
2.0	<u>4.36</u>	2.66	2.75	1.60	<u>1.45</u>	2.91	2.47
	<u>4.31</u>	4.32	4.35	4.42	<u>4.47</u>	4.32	4.43
3.0	<u>2.87</u>	1.51	1.81	1.09	<u>1.07</u>	1.57	1.68
	<u>2.85</u>	2.86	2.87	2.91	<u>2.94</u>	2.85	2.92

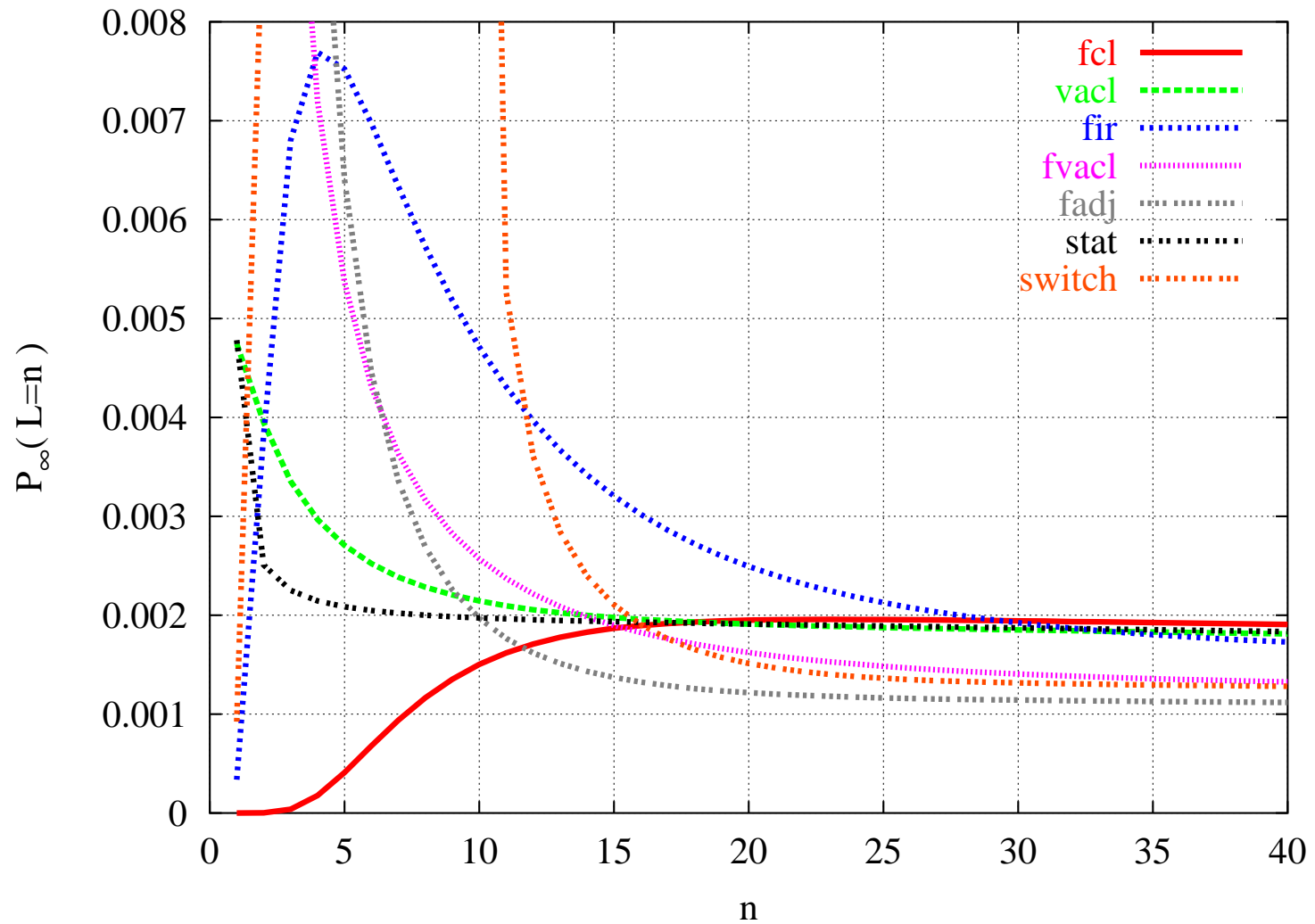
$$D_m = E_m(L - m + 1 | L \geq m)$$

$$(\lambda = 0.1, E_\infty(L) = 500, \delta = 1)$$



in-control PMF

$(\lambda = 0.1, E_{\infty}(L) = 500)$



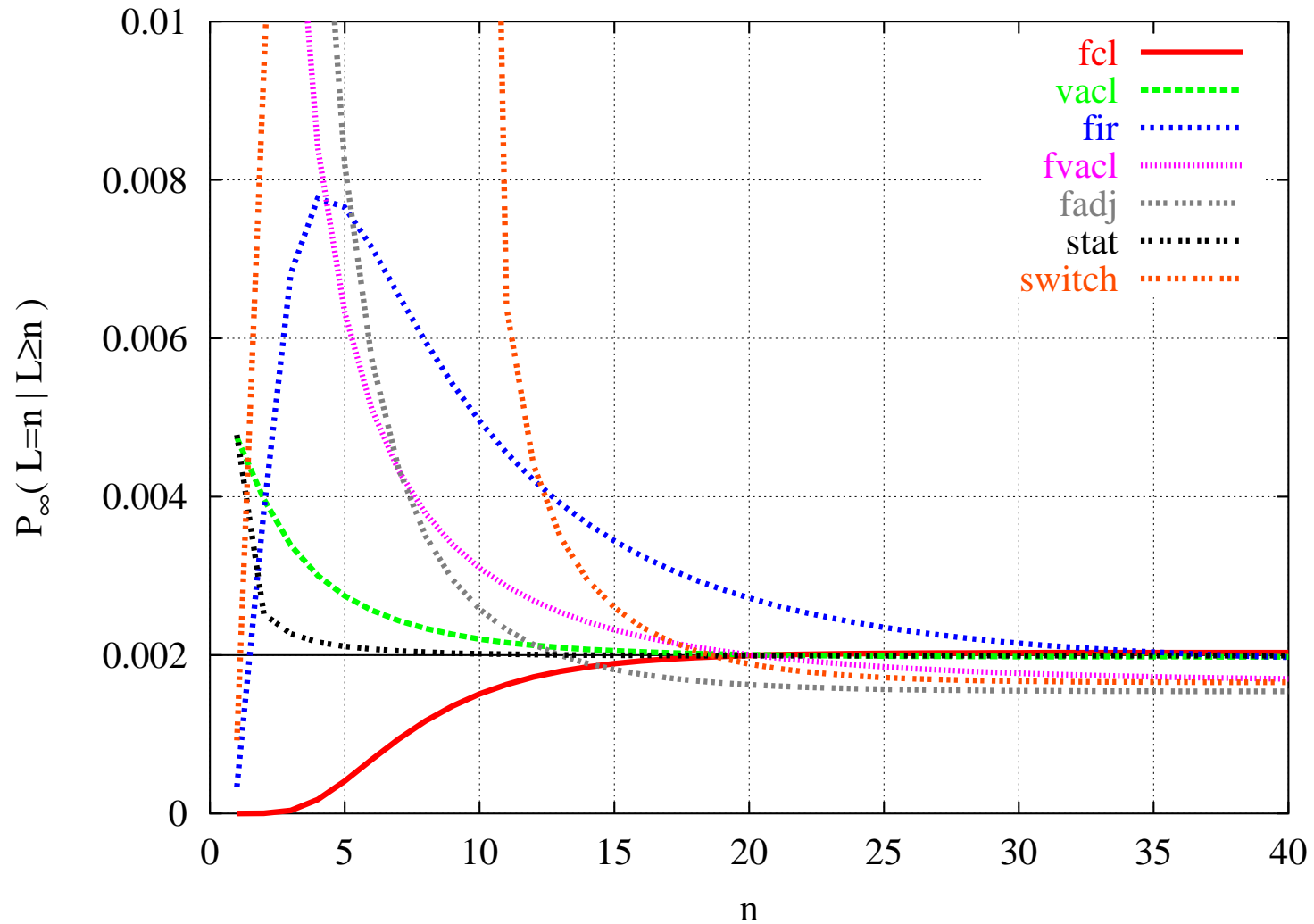
Probabilities for early false alarms

$\lambda = 0.1$, in-control ARL, i. e. $E_{\infty}(L) = 500$

EWMA scheme	fcl	vacl	fir	fvacl	fadj	stat	switch
$P_{\infty}(L = 1)$	0.0000	0.0047	0.0003	0.1125	0.1452	0.0048	0.0009
$P_{\infty}(L = 2)$	0.0000	0.0040	0.0038	0.0217	0.0435	0.0025	0.0094
$P_{\infty}(L = 3)$	0.0000	0.0034	0.0068	0.0109	0.0190	0.0022	0.0170
$P_{\infty}(L \leq 10)$	0.0063	0.0293	0.0551	0.1742	0.2391	0.0238	0.1761

false alarm rate $r_n = P_\infty(L = n | L \geq n)$

$(\lambda = 0.1, E_\infty(L) = 500)$



Conclusions

Regarding *fast initial response (FIR)*, one can distinguish three types of EWMA schemes:

feature	"true" FIR	"moderate" FIR	balanced
initial detection sensitivity	high	increased	low
long run performance	little deteriorated	slightly deteriorated	good
initial false alarm rate	high	moderate	low
	fvacl (Steiner) switch (Fink) fvacl (Rhoads et al.)	fir (Lucas/Sacc.)	stat (Knoth) fcl vacl