

# A LOWER BOUND ON THE ARL TO DETECTION OF A CHANGE WITH A PROBABILITY CONSTRAINT ON FALSE ALARM<sup>1</sup>

BY BENJAMIN YAKIR

*Hebrew University of Jerusalem*

An inequality that relates the probability of false alarm of a change-point detection policy to its average run length to detection is proved. By means of this inequality, a lower bound on the rate of detection, when the change occurs after a long delay, is derived.

**1. Introduction.** Let  $f_\theta(x) = \exp\{\theta x - \Psi(\theta)\}$  be the density, with respect to some  $\sigma$ -finite measure, of a one-parameter exponential family. Let  $(a, b)$  be an open interval of real numbers on which  $\Psi$  is finite, and let  $\theta_0$  and  $\theta$  be known,  $a < \theta_0 < \theta < b$ . We will assume, as in [2, page 1268], that  $\theta_0 = 0$ .

Consider the following simple formulation of a change-point detection with a probability bound on false alarm: Let  $X_1, X_2, \dots$  be an infinite sequence of random variables, and let  $1 \leq \nu \leq \infty$ , an (extended) integer, be an unknown parameter: the change-point of the sequence. With each  $1 \leq \nu \leq \infty$  and  $a < \omega \leq 0$ , a probability measure on the sequence of observations is associated by which the observations are independent. Under the probability measure  $\mathbb{P}_{\omega, \theta}^{(\nu)}$ , the marginal density of the first  $\nu - 1$  observations is  $f_\omega$  whereas the density of the observations that follow is  $f_\theta$ . Under the probability measure  $\mathbb{P}_{\omega, \theta}^{(\infty)}$  ( $= \mathbb{P}_\omega$ ) the observations are i.i.d. with density  $f_\omega$ , and under  $\mathbb{P}_{\omega, \theta}^{(1)}$  ( $= \mathbb{P}_\theta$ ) they are i.i.d. with density  $f_\theta$ . Let  $Z_i^{\omega, \theta} = \log[f_\theta(X_i)/f_\omega(X_i)]$  be the log-likelihood ratio of an observation, and set  $Z_i^\theta = Z_i^{0, \theta}$ . Denote by  $I(\omega, \theta)$  the  $\mathbb{P}_\theta$ -expectation of  $Z_i^{\omega, \theta}$  [ $I(\theta) = I(0, \theta)$ ].

A detection policy is a stopping time, defined on the sequence of observations. For the most part of this paper, the stopping times that will be considered are those that satisfy the probability constraint

$$(1) \quad \mathbb{P}_0(N < \infty) \leq \alpha,$$

for a given  $0 < \alpha < 1$ . Among such policies one seeks the policy that minimizes the average run length (ARL) to detection:

$$(2) \quad \mathbb{E}_{0, \theta}^{(\nu)}(N - \nu + 1 | N \geq \nu).$$

However, which among the policies that satisfy (1) is best depends on the value of  $\nu$  in (2). It is well known, for example, that when  $\nu = 1$  the power-1

---

Received August 1994; revised April 1995.

<sup>1</sup>This work supported in part by NIH Grant T32 ES07271.

AMS 1991 subject classifications. Primary 62L10; secondary 62N10.

Key words and phrases. Quality control, control charts, change-point detection, sequential tests.

SPRT minimizes (2) among all stopping times that satisfy (1). Nevertheless, the performance of the SPRT is poor when large values of  $\nu$  are considered.

The main result in this article is an inequality that relates the probability of a false alarm in (1) to the ARL to detection, given in (2). As a corollary of this inequality (Corollary 2), we are able to confirm part of a conjecture of Pollak and Siegmund [2, Section 6], regarding the rate of divergence of (2), considered as a function of  $\nu$ , as  $\nu \rightarrow \infty$ .

At the end of this paper we will look at stopping times that satisfy the stronger constraint

$$(3) \quad \mathbb{P}_\omega(N < \infty) \leq \alpha,$$

for all  $a < \omega \leq 0$ . The conjecture in [2] states that, for any stopping rule satisfying (3),

$$(4) \quad \limsup_{\nu \rightarrow \infty} \mathbb{E}_{\omega, \theta}^{(\nu)}(N - \nu + 1 | N \geq \nu) / \log \nu \geq 1/I(0, \theta).$$

It will be shown, via a counterexample, that the conjecture as stated is not true.

**2. The main inequality.** The main result is the following theorem.

**THEOREM 1.** *Assume that  $Z_i^\theta$ , the log-likelihood ratio of an observation, is nonlattice under  $\mathbb{P}_0$  and has a finite  $\mathbb{P}_0$ -expectation  $I(\theta)$ . Then there exists a finite constant  $c$  such that*

$$(5) \quad \sum_{\nu=1}^{\infty} \exp\{-I(\theta)\mathbb{E}_{0, \theta}^{(\nu)}(N - \nu + 1 | N \geq \nu)\} \leq \frac{\alpha \exp[c/(1 - \alpha)]}{1 - \alpha}.$$

The constant  $c$  does not depend on  $\alpha$ ,  $\nu$  or  $N$ . However, it does depend on  $\theta$ .

**PROOF.** Define, for each  $n = 1, 2, \dots$ ,

$$R_n^\theta = \sum_{\nu=1}^n \exp\left\{\sum_{i=\nu}^n Z_i^\theta\right\}.$$

These statistics are known as the Shiriyayev–Roberts statistics. It is easy to check that, for any stopping time  $N$ ,

$$\begin{aligned} \mathbb{P}_0(N < \infty) &= \sum_{j=1}^{\infty} \mathbb{E}_0[\mathbb{1}\{N = j\}] \\ &= \sum_{j=1}^{\infty} \sum_{\nu=1}^j \mathbb{E}_{0, \theta}^{(\nu)}\left[\frac{1}{R_j^\theta} \mathbb{1}\{N = j\}\right] \\ &= \sum_{\nu=1}^{\infty} \mathbb{E}_{0, \theta}^{(\nu)}\left[\frac{1}{R_N^\theta} \mathbb{1}\{N \geq \nu\}\right] \\ &\geq \sum_{\nu=1}^{\infty} \exp\{-\mathbb{E}_{0, \theta}^{(\nu)}[\log R_N^\theta | N \geq \nu]\} \mathbb{P}_{0, \theta}^{(\nu)}(N \geq \nu). \end{aligned}$$

In particular, if  $N$  satisfies (1), then  $\mathbb{P}_{0,\theta}^{(\nu)}(N \geq \nu) \geq 1 - \alpha$ , hence

$$(6) \quad \sum_{\nu=1}^{\infty} \exp\{-\mathbb{E}_{0,\theta}^{(\nu)}[\log R_N^\theta | N \geq \nu]\} \leq \frac{\alpha}{1 - \alpha}.$$

For each  $n \geq \nu$ , the statistic  $R_n^\theta$  can be represented as a product of two terms:

$$\begin{aligned} R_n^\theta &= \exp\left\{\sum_{i=\nu}^n Z_i^\theta\right\} \left(\sum_{j=1}^{\nu-1} \exp\left\{\sum_{i=j}^{\nu-1} Z_i^\theta\right\} + 1 + \sum_{j=\nu+1}^n \exp\left\{-\sum_{i=\nu}^{j-1} Z_i^\theta\right\}\right) \\ &= \exp\left\{\sum_{i=\nu}^n Z_i^\theta\right\} \times W^\theta(\nu, n). \end{aligned}$$

It follows, since  $Z_\nu^\theta, Z_{\nu+1}^\theta, \dots$  are i.i.d. under the measure  $\mathbb{P}_{0,\theta}^{(\nu)}(\cdot | N \geq \nu)$ , that

$$(7) \quad \begin{aligned} &\mathbb{E}_{0,\theta}^{(\nu)}[\log R_N^\theta | N \geq \nu] \\ &= I(\theta) \mathbb{E}_{0,\theta}^{(\nu)}(N - \nu + 1 | N \geq \nu) + \mathbb{E}_{0,\theta}^{(\nu)}(\log W^\theta(\nu, N) | N \geq \nu) \\ &\leq I(\theta) \mathbb{E}_{0,\theta}^{(\nu)}(N - \nu + 1 | N \geq \nu) + \frac{\mathbb{E}_{0,\theta}^{(\nu)}(\log W^\theta(\nu, N))}{1 - \alpha}. \end{aligned}$$

(Notice that  $\log W^\theta$  is positive when  $N \geq \nu$ , and its definition can be extended to be zero when  $N < \nu$ .)

However, the  $\mathbb{P}_{0,\theta}^{(\nu)}$ -distribution of the random variable  $W^\theta(\nu, N)$  is stochastically dominated, uniformly in  $\nu$  and  $N$ , by the distribution of  $W_\infty^\theta + 1 + W_1^\theta$ , where  $W_\infty^\theta$  and  $W_1^\theta$  are independent of each other, the distribution of  $W_1^\theta$  is the  $\mathbb{P}_\theta$ -distribution of  $\sum_{j=1}^\infty \exp\{-\sum_{i=1}^j Z_i^\theta\}$  and the distribution of  $W_\infty^\theta$  is the  $\mathbb{P}_0$ -distribution of  $\sum_{j=1}^\infty \exp\{\sum_{i=1}^j Z_i^\theta\}$ . To assert this claim, notice that the  $\mathbb{P}_{0,\theta}^{(\nu)}$ -distribution of  $\sum_{j=1}^{\nu-1} \exp\{\sum_{i=j}^{\nu-1} Z_i^\theta\}$  is identical to the  $\mathbb{P}_0$ -distribution of  $\sum_{j=1}^{\nu-1} \exp\{\sum_{k=1}^j Z_k^\theta\}$  which is  $\mathbb{P}_0$ -almost surely smaller than  $\sum_{j=1}^\infty \exp\{\sum_{i=1}^j Z_i^\theta\}$ . Likewise,  $\sum_{j=\nu+1}^N \exp\{-\sum_{i=\nu}^{j-1} Z_i^\theta\}$  is  $\mathbb{P}_{0,\theta}^{(\nu)}$ -almost surely smaller than  $\sum_{j=\nu}^\infty \exp\{-\sum_{i=\nu}^j Z_i^\theta\}$ .

From Theorem 4 in [1] it follows that the second term on the right-hand side of (7) is bounded, uniformly in  $\nu$  and  $N$ . Plugging (7) into (6) completes the proof.  $\square$

**COROLLARY 1.** *Assume that  $Z_i^\theta$  is nonlattice under  $\mathbb{P}_0$  and has a finite second moment under  $\mathbb{P}_\theta$ . Let  $a_1, a_2, \dots$  be a sequence of positive numbers. Then the upper limit*

$$(8) \quad \limsup_{\nu \rightarrow \infty} \left[ \mathbb{E}_{0,\theta}^{(\nu)}(N - \nu + 1 | N \geq \nu) + \frac{\log a_\nu}{I(\theta)} \right]$$

*is infinite for every stopping time  $N$  that satisfies (1) iff  $\sum_{\nu=1}^\infty a_\nu = \infty$ .*

PROOF. Assume first that  $\sum_{\nu=1}^{\infty} a_{\nu} < \infty$ . Let

$$(9) \quad \alpha_{\nu} = \frac{\alpha a_{\nu}}{\sum_{k=1}^{\infty} a_k}, \quad 1 \leq \nu < \infty.$$

The stopping time

$$N = \inf \left\{ n : \sum_{\nu=1}^n \alpha_{\nu} \exp \left\{ \sum_{i=\nu}^n Z_i^{\theta} \right\} + \sum_{\nu=n+1}^{\infty} \alpha_{\nu} \geq 1 \right\},$$

which is similar to a policy suggested in [2], would produce a finite upper limit in (8). The last claim follows from the relation

$$N \leq \inf \left\{ n : \sum_{i=\nu}^n Z_i^{\theta} \geq -\log \alpha_{\nu} \right\}, \quad \nu = 1, 2, \dots,$$

Wald's lemma and from the fact that the expectation of the overshoot is bounded.

Assume next that the sum of the sequence  $\{a_{\nu}\}$  diverges. Inequality (5) can be rewritten in the form

$$\sum_{\nu=1}^{\infty} a_{\nu} \exp \left\{ -I(\theta) \left[ \mathbb{E}_{0,\theta}^{(\nu)}(N - \nu + 1 | N \geq \nu) + \frac{\log a_{\nu}}{I(\theta)} \right] \right\} \leq \frac{\alpha \exp[c/(1 - \alpha)]}{1 - \alpha}.$$

It is easy to see that a finite upper limit in (8) would produce a contradiction.  $\square$

COROLLARY 2. Under the assumptions of Corollary 1, relation (59) in [2] holds for  $\omega = 0$ ; that is,

$$\limsup_{\nu \rightarrow \infty} \mathbb{E}_{0,\theta}^{(\nu)}(N - \nu + 1 | N \geq \nu) / \log \nu \geq 1/I(0, \theta),$$

for any stopping rule that satisfies (3) [and hence (1)].

PROOF. The proof follows immediately from the proof of Corollary 1 with  $\alpha_{\nu} = 1/\nu$ .  $\square$

**3. A counterexample.** Corollary 2 states that the conjecture of Pollak and Siegmund is true for  $\omega = 0$ . It is false, however, for  $a < \omega < 0$ . For any such  $\omega$  one can find a stopping rule  $\sigma$ , satisfying (3), for which (4) is false. Indeed, fix  $a < \omega < 0$ . Consider, for any integer  $m$  and negative real  $\eta$ ,  $a < \eta \leq 0$ , the auxiliary stopping rule

$$N(m, \eta) = \inf \left\{ n \geq m + 1 : \sum_{\nu=m+1}^n \alpha_{\nu} \exp \left\{ \sum_{i=\nu}^n Z_i^{\eta, \theta} \right\} + \sum_{\nu=n+1}^{\infty} \alpha_{\nu} \geq 2 \right\},$$

where  $\alpha_{\nu}$  is defined in (9) with  $a_{\nu} = 1/(\nu \log^2 \nu)$ . Note that, for any  $a < \zeta \leq \eta$ ,

$$\mathbb{P}_{\zeta}(N(m, \eta) < \infty) \leq \frac{\alpha}{2}$$

and that, for any  $\nu > m$ ,

$$\mathbb{E}_{\xi, \theta}^{(\nu)}(N(m, \eta) - \nu + 1 | N(m, \eta) \geq \nu) \leq \frac{-\log(\alpha_\nu/2) + \text{const.}}{I(\eta, \theta)}.$$

The constant is a bound on the expected overshoot. It depends only on  $\eta$  and  $\theta$ .

Given  $0 < \varepsilon < -\omega$ , let  $\delta_m$  be a test of  $H_0: \eta \geq \omega + \varepsilon$  versus  $H_1: \eta < \omega + \varepsilon$ , based on the first  $m$  observations, with significance level  $\alpha/2$ . Let  $m = m(\omega, \varepsilon)$  be a sample size needed to assure that the power of the test, at  $\eta = \omega$ , is at least  $1 - \varepsilon$ .

Consider the stopping rule

$$\sigma = \sigma(\omega, \varepsilon) = N(m, (\omega + \varepsilon)\delta_m).$$

The stopping rule  $\sigma$  satisfies (3), since  $\mathbb{P}_\eta(\sigma < \infty)$  is less than  $\alpha/2$  for  $\eta \leq \omega + \varepsilon$  and less than  $\alpha$  for  $\omega + \varepsilon < \eta \leq 0$ . This stopping rule is a random mixture of the two stopping times  $N(m, 0)$  and  $N(m, \omega + \varepsilon)$ . The mixture is based on the first  $m$  observations. It can be shown that

$$\mathbb{P}_{\omega, \theta}^{(\nu)}(\delta_m = 0 | \sigma \geq \nu) < \frac{\varepsilon}{(1 - \alpha/2)(1 - \varepsilon)}.$$

As a result it follows that the ARL to detection, for all  $\nu > m$ , satisfies the relation

$$\begin{aligned} & \mathbb{E}_{\omega, \theta}^{(\nu)}(\sigma - \nu + 1 | \sigma \geq \nu) \\ & \leq \frac{-\log \alpha_\nu/2 + \text{const.}}{I(\omega + \varepsilon, \theta)} + \frac{\varepsilon}{(1 - \alpha/2)(1 - \varepsilon)} \frac{-\log \alpha_\nu/2 + \text{const.}}{I(0, \theta)}. \end{aligned}$$

A contradiction to (4) can be derived from the above inequality since  $I(\omega + \varepsilon, \theta) > I(0, \theta)$  and  $\varepsilon$  can be chosen arbitrarily small.

### REFERENCES

[1] KESTEN, H. (1973). Random difference equations and renewal theory for products of random matrices. *Acta Math.* **131** 207-248.  
 [2] POLLAK, M. and SIEGMUND, D. (1975). Approximations to the expected sample size of certain sequential tests. *Ann. Statist.* **3** 1267-1282.

DEPARTMENT OF STATISTICS  
 THE HEBREW UNIVERSITY OF JERUSALEM  
 JERUSALEM 91905  
 ISRAEL