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A NOTE ON A SEQUENTIAL PROBABILITY RATIO TEST

por

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Summary. This is a type of problem that lies outside the scope of the exponential family. If the Z_1 are real valued, with density $\frac{1}{\sigma} \, \exp \, \left[- \, \frac{z - \mu}{\sigma} \right] h(z - \mu)$ (here h(z) = 1 or 0, according as z>0 or $z\leqslant 0$), and where one value of σ is tested against another, it is shown that

$$\ell_n R_n = b + \sum_{i=1}^{n} (Z_i - U_n - a),$$

where $U_n = \min_{1 \le i \le n} Z_i$, a is a positive constant. Using this expression it is proved that for every non-degenerate distribution of the Z_i , P(N > n) is exponentially bounded, which, of course, implies termination with probability 1.

§1. Introducción. In Abu-Salih [1], the following model was discussed. Z, Z_1, Z_2, \ldots is a sequence of independent identically distributed (iid) m-vectors, with k-parameter exponential distribution P. G^* is a group of transformations of the form $Z_n \mapsto C(Z_n + b)$, where $C \in G$, G is a Lie group of m × m nonsingular matrices, dim $G \geqslant 1$, and G is closed in the general linear group GL(m,R); b is an m-vector of reals, and the total ity of vectors b form an invariant subspace under G.

Let P have the density

(1.1)
$$P_{\Theta}^{Z}(x) = B(\Theta)h(z) \exp \left(\sum_{j=1}^{k} \Theta_{j}S_{j}(z)\right)$$

with respect to Lebesgue measure on the m-dimensional Euclidian space E^m , and where $\theta = (\theta_1, \dots, \theta_k)$ belongs to the natural parameter space Ω , and $S = (S_1, \dots, S_k)$ is a continuously differentiable mapping of E^m into E^k .

Let $U=(U_1,U_2,\ldots)$ be a maximal invariant under G^* in the sample space, and $\gamma=\gamma(\theta)$ a maximal invariant in Ω . For given $\theta^1,\theta^2\in\Omega$ such that $\gamma(\theta^1)\neq\gamma(\theta^2)$, write $U^n=(U_1,U_2,\ldots,U_n)$, and let P^* be its density under $\gamma(\theta^1)$, i=1,2, with respect to some σ -finite

measure. Let

$$(1.2) rn = P'2n / P'1n$$

and

(1.3)
$$R_n = r_n(U^n)$$
;

then R_n is the probability ratio at the n^{th} stage of sampling based on the maximal invariant U. A sequential probability ratio test (SPRT) based on $\{R_n\}$ continues sampling as long as $B < R_n < A$ (B and A are two fixed stopping bounds), stops and accepts θ^1 (resp. θ^2) the first time that $R_n < B$ (resp. $R_n > A$). A SPRT based on $\{R_n\}$ will be called in invariant SPRT.

The limiting behavior of R_n is studies in [1] under the assumption that the actual distribution belongs to certain family \mathcal{F}_{\bullet} , and it is proved that there are three cases:

(i)
$$\lim_{n\to\infty} R_n = \infty$$
, a.e.P,

(ii)
$$\lim_{n\to\infty} R_n = 0$$
, a.e.P,

(iii)
$$\lim_{n\to\infty} \sup_{n\to\infty} R_n = \infty$$
, $\lim_{n\to\infty} \inf_{n\to\infty} R_n = 0$, $\lim_{n\to\infty} \inf_{n\to\infty} R_n = 0$

each one corresponding to a subfamily of \mathcal{F} . This

establishes termination with probability 1 of the (SPRT) based on $\{R_n\}$.

The results obtained above form an extension of those of Wijsman in [2] and [3] in which the underlying model was assumed to be multivariate normal. Our methods of proof are closely modeled on those in [2] and [3].

\$2. Sequential probability ratio test based on negative exponential distribution with location parameter. It is of interest to consider a model similar to the exponential one, except for a location parameter in the function h(z) of (1.1). We were unable to reduce this model to the one we have summarized in the introduction. Yet, we have worked a simple example for which we obtained an exponential bound on P(N > n) for any non-degenarate distribution P.

Let z, z_1, z_2, \dots be iid random variables with density p_{Θ}^z with respect to Lebesgue measure. Assume

(2.1)
$$p_{\theta}^{Z}(z) = \frac{1}{\sigma} h(z-\mu) \exp(-\frac{1}{\sigma}(z-\mu))$$

where $\theta=(\mu,\sigma)$ and h(x)=1 if x>0, h(x)=0 if $x\leqslant 0$. $\Omega=\{\theta=(\mu,\sigma); -\infty<\mu<\infty,\sigma>0\}$ is the parameter space. The joint density of 142

$$(z_1, \dots, z_n) \quad \text{is given by}$$

$$(2.2) \quad P_{\Theta}^{z_1, \dots, z_n}(z_1, \dots, z_n)$$

$$= \frac{1}{\sigma^n} \left[\prod_{i=1}^n h(z_i - \mu) \right] \exp \left[-\frac{1}{\sigma} \sum_{i=1}^n (z_i - \mu) \right].$$

Test about σ , e.g. H_o : $\sigma = \sigma_1$ vs H_1 : $\sigma = \sigma_2$, where $\sigma_1 > \sigma_2$. Consider the group of translations G acting on the sample space as follows:

g:
$$Z_i \rightarrow Z_i + a$$
 for $i = 1, 2, ...$

where $-\infty$ < a < ∞ and g \in G. It is clear that G leaves the model invariant.

Using (2.11) in [1] ((3.3) in [2]) we get

$$\frac{\int \frac{1}{\sigma_{2}^{b}} \exp \left[-\frac{1}{\sigma_{2}} \sum_{i=1}^{n} (z_{i} + a - \mu)\right] \prod_{i=1}^{n} h(z_{i} + a - \mu) da}{\int \frac{1}{\sigma_{1}^{n}} \exp \left[-\frac{1}{\sigma_{1}} \sum_{i=1}^{n} (z_{i} + a - \mu)\right] \prod_{i=1}^{n} h(z_{i} + a - \mu) da} \\
= \frac{\frac{1}{\sigma_{1}^{n}} \frac{\sigma_{2}}{\sigma_{1}^{n}} \exp \left[-\frac{1}{\sigma_{2}} \sum_{i=1}^{n} (z_{i} - \mu)\right] \exp \left[-\frac{n}{\sigma_{2}} (\mu - u_{n})\right]}{\frac{1}{\sigma_{1}^{n}} \frac{\sigma_{1}^{n}}{\sigma_{1}^{n}} \exp \left[-\frac{1}{\sigma_{1}} \sum_{i=1}^{n} (z_{i} - \mu)\right] \exp \left[-\frac{n}{\sigma_{1}} (\mu - u_{n})\right]}$$

(Where
$$u_n = \min_{1 \le i \le n} z_i$$
)

=
$$(\frac{\sigma_1}{\sigma_2})^{n-1} \exp[(\frac{1}{\sigma_1} - \frac{1}{\sigma_2}) \sum_{n=1}^{n} (z_i - u_n)].$$

But $R_n = r_n(Z_1, \dots, Z_n)$, hence from (2.3):

(2.4)
$$\ln R_n = \ln \left(\frac{\sigma_1}{\sigma_2}\right)^{-1} + n \ln \frac{\sigma_1}{\sigma_2}$$

+ $\left(\frac{1}{\sigma_1} - \frac{1}{\sigma_2}\right) \sum_{i=1}^{n} \left(z_i - v_n\right)$

The SPRT mentioned above will continue sampling as long as $\ln B < \ln R_n < \ln A$ and, from (2.4), it continues sampling if

(2.5)
$$\ln B + \ln \frac{\sigma_1}{\sigma_2}$$

$$< n \ln \frac{\sigma_1}{\sigma_2} + (\frac{1}{\sigma_1} - \frac{1}{\sigma_2}) \sum_{i=1}^{n} (Z_i - U_n)$$

$$< \ln A + \ln \frac{\sigma_1}{\sigma_2} ,$$

where $U_n = \min_{1 \le i \le n} Z_i$. Let

(2.6)
$$A_{1} = (\ln B + \ln \frac{\sigma_{1}}{\sigma_{2}}) / (\frac{1}{\sigma_{1}} - \frac{1}{\sigma_{2}})$$

$$A_{2} = (\ln A + \ln \frac{\sigma_{1}}{\sigma_{2}}) / \frac{1}{\sigma_{1}} - \frac{1}{\sigma_{2}}$$

$$A_{3} = - \ln \frac{\sigma_{1}}{\sigma_{2}} / (\frac{1}{\sigma_{1}} - \frac{1}{\sigma_{2}})$$

$$= (\ln \frac{1}{\sigma_1} - \ln \frac{1}{\sigma_2}) / (\frac{1}{\sigma_1} - \frac{1}{\sigma_2})$$

(a is positive since numerator and denominator have the same sing). Using (2.5) and (2.6) we continue sampling as long as:

(2.7)
$$A_1 < \sum_{i=1}^{n} (Z_i - U_n - a) < A_2$$

From now on, we drop the assumption that Z, Z_1 , Z_2 , ... have the density (2.1) and instead consider Z_1, Z_2, \ldots iid random variables with distribution P and denote the distribution of (Z_1, \ldots, Z_n) also by P. The only restriction we impose on P is that it be non-degenerate. Under this conditions, we like to establish termination, with probability 1, of the SPRT based on (2.7) and find exponential bounds on P(N > n). Let

(2.8)
$$E_n = \{A_1 < \sum_{i=1}^n (Z_i - U_n - a) < A_2\}$$
.

and let us compute $P(E_{n+1}|E_1,...,E_n)$. Given E_n with $\sum_{i=1}^{n} (A_i - U_n - a) = d_n$, suppose $Z_{n+1} \ge U_n$; then

$$\sum_{i=1}^{n+1} (Z_i - U_{n+1} - a) = d_n + Z_{n+1} - U_n - a.$$

Hence, given E_n and $Z_{n+1} \geqslant U_n$, E_{n+1} implies $Z_{n+1} - U_n - a < D = A_2 - A_1$, and, in particular

(2.9)
$$U_n \leq Z_{n+1} \leq U_n + D + a$$
.

On the other hand, given E_n and $Z_{n+1} \subset U_n$, then E_{n+1}

implies that
$$\sum_{n+1}^{n+1} (Z_i - U_{n+1} - a) = d_n + n(U_n - Z_{n+1}) - a$$
 $i=1$

lies between A_1 and A_2 hence

$$A_1 - d_n < n(U_n - Z_{n+1}) - a < A_2 - d_n$$

which implies

(2.10)
$$U_n - \frac{D}{n} - \frac{a}{n} < Z_{n+1} < U_n + \frac{D}{n} - \frac{a}{n}$$
.

Comparing (2.9) and (2.10) we have: $given E_n$, then E_{n+1} implies

$$(2.11) U_n - \frac{L}{n} < Z_{n+1} < U_n + L ,$$

where L = D + a. Furthermore, given E_1, \dots, E_n , then E_{n+1} implies (2.11), and so

(2.12)
$$P(E_{n+1}|E_1,...,E_n)$$

 $\leq P(U_n - \frac{L}{n} < Z_{n+1} < U_n + L|E_1,...,E_n)$
 $= E[I_{F_{n+1}}|E_1,...,E_n]$
 $= E[E[I_{F_{n+1}}|U_n,E_1,...,E_n]|E_1,...,E_n]$

with probability 1, where I_{F} is the indicator n+1function of F_{n+1} and $F_{n+1} = \{U_n - \frac{L}{n} < Z_{n+1} < U_n + L\}$. 146

Suppose that the support of Z is not contained in an interval of length 2L, and define $\rho = \sup_{-\infty < x < \infty} P(x-L < Z < x+L); \quad \text{then} \quad \rho < 1. \quad \text{Fursuperson}$ thermore, since Z_{n+1} is independent of Z_1, \dots, Z_n and hence independent of E_1, \dots, E_n and U_n , we have

$$E[I_{F_{n+1}} | E_1, \dots, E_n, U_n = u_n]$$

$$= P(u_n - \frac{L}{n} < Z_{n+1} < u_n + L | E_1, \dots, E_n, U_n = u_n)$$

$$= P(u_n - \frac{L}{n} < Z_{n+1} < u_n + L) \le \rho < 1;$$

and therefore,

$$E[E[I_{F_{n+1}}|E_1,...,E_n,U_n]|E_1,...,E_n] \le \rho < 1$$
;

hence (2.12) becomes:

(2.13)
$$P(E_{n+1}|E_1,...E_n) \le \rho < 1$$
 for $n = 1,2,...$

But

$$P(N>n) = P(B < R_m < A, 1 \le m \le n)$$

$$= P(E_1E_2...E_n)$$

$$= P(E_1)P(E_2|E_1)P(E_3|E_1E_2)...P(E_n|E_1...E_{n-1})$$

$$\le P(E_1) \rho^{n-1} < c \rho^n,$$

with c > 0, $\rho < 1$.

Therefore, we have established an exponential bound on P(N>n), for P with support not contain-

ed in an interval of length 2L.

The case of bounded support is considered in the rest of the paper. Without loss of generality, we assume that the support of P is (0,b). We may do this because we are studying $\sum_{i=1}^{n} (Z_i - U_i - a)$ which is invariant under i=1 translations.

CASE 1: $a \in (0,b)$. Let δ be a positive constant such that $a + 2\delta < b$. Hence

(2.14)
$$P(Z > a + 2\delta) = p > 0.$$

Let

$$m = \left[\frac{D}{\delta}\right] + 1 \qquad (D = |A_2 - A_1|)$$

$$v_k = \min_{1 \le i \le mk} Z_i$$

(2.15)
$$E_k = \{A_1 < \sum_{i=1}^{mk} (Z_i - V_k - a) < A_2\}$$

(so E_k is the same as E_{mk} from (2.8)

$$B_k = \{Z_i > a+2\delta, i = m(k-1)+1,...,mk\}$$

$$A_{k} = \{v_{k} \leq \delta \}.$$

Since $\delta > 0$ then

(2.16)
$$P(Z > \delta) = q_1 < 1$$
.

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Let
$$\Delta_{k} = \sum_{i=1}^{mk} (Z_{i} - V_{k} - a) - \sum_{i=1}^{m(k-1)} (Z_{i} - V_{k-1} - a) ;$$
then
$$\Delta_{k} = \sum_{i=m(k-1)+1}^{mk} (Z_{i} - a) - mV_{k} + m(k-1)(V_{k-1} - V_{k})$$

$$\geq \sum_{i=m(k-1)+1}^{mk} (Z_{i} - a) - mV_{k} ,$$

since $V_k \leq V_{k-1}$. Also we have

(2.17) i) Given
$$E_{k-1}$$
, then E_k implies $|\Delta_k| < D$.
ii) Given A_k , then B_{k+j} implies $\Delta_{k+j} > 2m\delta$
 $-m\delta = m\delta > D$, which implies $|\Delta_{k+j}| > D$ for $j = 1, 2, \ldots$

Therefore, given E_{k+j} and A_k we have for any j = 1, 2, ...

(2.18)
$$B_{k+j+1}$$
 implies \tilde{E}_{k+j+1} (~ denotes complementation)

(2.19)
$$E_{k+j+1}$$
 implies B_{k+j+1} .

Now,

(2.20)
$$P(E_1 E_2 ... E_{2k})$$

 $\leq P(E_k E_{k+1} ... E_{2k})$
 $= P(E_k ... E_{2k} A_k) + P(E_k ... E_{2k} A_k)$

$$\leq P(E_k \dots E_{2k}^{A_k}) + P(\tilde{A}_k)$$

$$= P(E_k \dots E_{2k}^{A_k}) + q_1^{mk},$$

by (2.16). And

(2.21)
$$P(E_k ... E_{2k} A_k) = P(E_k ... E_{2k} | A_k) P(A_k)$$

 $\leq P(E_k ... E_{2k} | A_k)$,

because $0 \le P(A_k) \le 1$ for k = 1, 2... Also

(2.22)
$$P(E_{k}...E_{2k}|A_{k})$$

= $P(E_{k}|A_{k})P(E_{k+1}|E_{k},A_{k})...P(E_{2k}|E_{k},E_{k+1},...$
 $...,E_{2k-1},A_{k})$
 $\leq P(E_{k}|A_{k})P(\tilde{B}_{k+1}|E_{k},A_{k})...P(\tilde{B}_{2k}|E_{k},...$
 $...,E_{2k-1},A_{k})$

by (2.19). But, since B_{k+j} is independent of A_k and E_{k+j-i} for $i=1,\ldots,j$ then

(2.23)
$$P(\tilde{B}_{k+j}|A_k, E_{k+j-i}, i = 1,...,j)$$

= $P(\tilde{B}_{k+j}) = 1 - p^m = p_1$

where p is given by (2.14) and so $p_1 < 1$.

Using (2.23), (2.22), and (2.21), then (2.20) becomes: $P(E_1E_2...E_{2k}) \le c_1p_1^k + q_1^{mk} \le c_2p_2^{2k}$, 150

where $\rho_2^2 = \max(p_1, q_1^m) < 1$, and c_1, c_2 are easily determined positive constants. From (2.7)

$$P(N>n) = P(A_1 < \sum_{i=1}^{\nu} (Z_i - U_{\nu} - a) < A_2, \text{ for } \nu = 1, ..., n)$$

$$\leq P(A_1 < \sum_{i=1}^{\nu} (Z_i - U_{\nu} - a) < A_2, \text{ for } \nu = m, 2m, ...$$

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(where k is the largest integer such that 2mk≤n)

$$\leq c_2 \rho_2^{2k} < c\rho^n$$

where $\rho = \rho_2^{1/m} < 1$, and c > 0. Hence,

(2.24)
$$P(N > n) < c\rho_i^n, c > 0, \rho < 1$$

CASE 2: a = b.

$$\sum_{i=1}^{n} (Z_{i} - U_{n} - a) = \sum_{i=1}^{n} (Z_{i} - U_{n} - b) \leq \sum_{i=1}^{n} (Z_{i} - b)$$

with probability 1.

Let N^* be the smallest integer ℓ for which $\sum_{i=1}^{\ell} (Z_i - b) \le A_1$, then $N \le N^*$. But Z < b with probability 1, therefore we need to consider $A_1 < 0$ only.

Let δ be a constant, $0 < \delta < b$, and $m = \left[\frac{d}{\delta}\right] + 1$, where $d = |A_1|$. Let

(2.25)
$$E_k = \{A_1 < \sum_{i=1}^{mk} (Z_i - b) \le 0\}$$

 $B_k = \{Z_i - b < -\delta, i = m(k-1)+1, ..., mk\}.$

Given E_{k-1} , then E_k implies

$$(2.26) \quad |\Delta_{k}| = |\sum_{i=1}^{mk} (Z_{i} - b) - \sum_{i=1}^{m(k-1)} (Z_{i} - b)|$$

$$= |\sum_{i=m(k-1)+1}^{mk} (Z_{i} - b)| < d.$$

But, B implies

(2.27)
$$|\Delta_k| = |\sum_{m(k-1)+1}^{mk} (Z_i - b)| > m\delta \ge d$$
,

which implies $\widetilde{E_{k-1}^E}_k$, and therefore $E_{k-1}^E_k$ implies \widetilde{B}_k . Hence, for k = 2, 3, ...

(2.28)
$$P(E_k | E_{k-1}, ..., E_1) \le P(\tilde{B}_k | E_{k-1}, ..., E_1)$$

= $P(\tilde{B}_k) = 1 - (P(Z < b - \delta))^m = q < 1,$

and therefore,

(2.29)
$$P(E_1, ..., E_k) = P(E_1)P(E_2|E_1)....P(E_k|E_{k-1}, ..., E_1) \le c q^k$$
,

with c > 0, q < 1.

(2.30)
$$P(N^*>n) = P(A_1 < \sum_{i=1}^{\nu} (Z_i-b) \le 0, \nu = 1,...,n)$$

 $\le P(A_1 < \sum_{i=1}^{m\ell} (Z_i-b) \le 0, \ell = 1,2,...$
 $...,[n/m])$
 $= P(E_1...E_k) \le c_1 q^k < c \rho^n$,

where c > 0 (properly defined), and $\rho^n = q^{[n/m]} < 1$.

But $N \leq N^*$, hence

(2.31) $P(N>n) \leq P(N^*>n) \leq c\rho^n$, with c > 0, $\rho < 1$.

CASE 3: a > b. We observe that $Z_i - U_n - a < b - a < 0$ with probability 1, and hence we have to terminate sampling with Probability 1 after at most $[-A_1/a-b] + 1$ steps, when A_1 is negative. When A_1 is positive, we have termination after the first observation. This completes the proof.

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