

POWER-ONE TESTS BASED ON SAMPLE SUMS¹

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This paper studies the properties of open-ended power-one tests of $H_0: \theta \leq \theta_0$ versus $H_1: \theta > \theta_0$ or of $H: \theta = \theta_0$ versus $K: \theta \neq \theta_0$ based on sample sums stopped at moving boundaries. The behavior of the expected sample size is analyzed and certain asymptotic results as $\theta \rightarrow \theta_0$ are obtained in the case of a location parameter and also in the case of an exponential family of distributions.

1. Introduction. Suppose X_1, X_2, \dots are i.i.d. random variables with distribution function F_θ , $\theta \in \Theta$, where Θ is an open subset of the real line. Let $\theta_0 \in \Theta$. We want to test the hypothesis $H_0: \theta \leq \theta_0$ versus $H_1: \theta > \theta_0$. Assume that the family of distributions F_θ is stochastically increasing, and that for all $\theta > \theta_0$, $E_\theta X_1$ exists and is $> \mu$ for some real number μ . Given $0 < \alpha < 1$, let $b(n)$ be a sequence of real numbers such that

$$(1.1) \quad \limsup_{n \rightarrow \infty} b(n)/n \leq \mu \quad \text{and} \\ P_\theta[S_n \geq b(n) \text{ for some } n \geq m] \leq \alpha,$$

where $S_n = X_1 + \dots + X_n$. Consider the following test: Stop sampling at stage

$$(1.2) \quad T = \inf \{n \geq m : S_n \geq b(n)\}$$

and reject H_0 . (We do not reject H_0 as long as we continue sampling.) For $\theta > \theta_0$, since $E_\theta X_1 > \mu$ and $\limsup_{n \rightarrow \infty} b(n)/n \leq \mu$, the strong law of large numbers implies that $P_\theta[\text{Reject } H_0] = P_\theta[T < \infty] = 1$, and in fact $E_\theta T < \infty$. Since the family of distribution functions F_θ is stochastically increasing, it follows from (1.1) that for $\theta \leq \theta_0$, $P_\theta[\text{Reject } H_0] = P_\theta[T < \infty] \leq \alpha$, i.e., the Type I error probability is $\leq \alpha$. Hence we have a level- α power-one test of H_0 versus H_1 . The expected sample size $E_\theta T$ is infinite for $\theta \leq \theta_0$ and is a finite nonincreasing function of θ for $\theta > \theta_0$. This monotone property of $E_\theta T$ follows easily from the fact that the family of distribution functions F_θ is stochastically increasing (cf. [8]). In Sections 2 and 3, we shall study the asymptotic behavior of $E_\theta T$ as $\theta \downarrow \theta_0$.

Similar ideas as above can be used to obtain level- α power-one tests of the two-sided hypothesis $H: \theta = \theta_0$ versus $K: \theta \neq \theta_0$. Here we assume that $E_{\theta_0} X_1 = \mu$ and that for $\theta \neq \theta_0$, $E_\theta X_1$ exists and is not equal to μ . Given $0 < \alpha < 1$, let

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$b_1(n) > b_2(n)$ be two sequences of real numbers satisfying the following condition:

$$(1.3) \quad \lim_{n \rightarrow \infty} b_i(n)/n = \mu \quad \text{for } i = 1, 2 \quad \text{and} \\ P_{\theta_0}[b_2(n) < S_n < b_1(n) \text{ for all } n \geq m] \geq 1 - \alpha.$$

We stop sampling at stage

$$(1.4) \quad N = \inf \{n \geq m : S_n \notin (b_2(n), b_1(n))\}$$

and reject H when we stop. For $\theta \neq \theta_0$, since $E_{\theta} X_1 \neq \mu$ and $\lim_{n \rightarrow \infty} b_i(n)/n = \mu$ ($i = 1, 2$), the strong law of large numbers implies that $P_{\theta}[N < \infty] = 1$, and in fact $E_{\theta} N < \infty$. By (1.3),

$$P_{\theta_0}[\text{Reject } H] = P_{\theta_0}[N < \infty] \leq \alpha,$$

and so the test has level α . The asymptotic behavior of $E_{\theta} N$ as $\theta \rightarrow \theta_0$ will be treated in Sections 2 and 3.

In [1], [2], [3], [4], [9], [10], [13], [14], [15], [16], by making use of sharp martingale inequalities, sequences $b(n)$ satisfying (1.1) and $b_1(n), b_2(n)$ satisfying (1.3) have been constructed for certain families of distributions. In the statistical literature, open-ended tests based on sample sums as described above for the unknown parameters of various parametric families have been considered by Darling and Robbins [1], [2], [3], [4], Fabian [5], Farrell [6], Robbins [14], Robbins and Siegmund [15], [17], [18], Pollak and Siegmund [13] and Lai and Siegmund [11]. The references [4], [11] and [13] contain some Monte Carlo studies of the expected sample sizes and the error probabilities of these tests.

In [18, page 429], Robbins and Siegmund have discussed the importance of finding the asymptotic behavior of the expected sample size of a power-one test at alternatives close to θ_0 for two-sided tests of $H: \theta = \theta_0$ versus $K: \theta \neq \theta_0$ or for one-sided tests of $H_0: \theta \leq \theta_0$ versus $H_1: \theta > \theta_0$. This kind of asymptotic behavior was first investigated by Farrell [6] who, in the case of a normalized exponential family of distributions, found the asymptotic mean (as $\theta \rightarrow \theta_0 = 0 = \mu$) of the two-sided stopping rule N as defined in (1.4) with $b_1(n) = -b_2(n) = b(n)$ and

$$(1.5) \quad b(n) = \sigma\{2n[\log_2(n + e) + c \log_3(n + e^e)]\}^{\frac{1}{2}},$$

where $\sigma^2 = E_0 X_1^2$, $c > \frac{3}{2}$ and \log_2 denotes $\log \log$, etc. For the special case where X_1, X_2, \dots are i.i.d. normal random variables, Farrell's result has been extended to the one-sided stopping rule T of (1.2) by Robbins and Siegmund [17] who, besides considering the iterated-logarithm boundary of (1.5), also consider the logarithmic case $b(n) \sim (n \log n)^{\frac{1}{2}}$ such that $b(n)$ is concave and increasing.

The above results of Farrell, Robbins and Siegmund will be extended to a general class of boundaries $b(n)$ in Section 3 where we study the asymptotic behavior (as $\theta \rightarrow \theta_0$) of the expected sample size of open-ended power-one tests of H_0 versus H_1 or of H versus K in the case of a general exponential family of distributions. These results are similar to the corresponding results in the case where θ is a location parameter. The location-parameter problem, which has

a simpler formulation, will be treated in Section 2 and we now briefly sketch the main results thereof. Since θ is a location parameter, we can write $X_i = Z_i + \theta$ so that the distribution of Z_i does not depend on θ . Assuming that $EZ_1 = 0$ and letting $\tilde{S}_n = Z_1 + \dots + Z_n$, the stopping rule of the power-one test of $H_0: \theta \leq 0$ versus $H_1: \theta > 0$ as given by (1.2) can be written as

$$(1.6) \quad T(\theta) = \inf \{n \geq m : \tilde{S}_n + n\theta \geq b(n)\}.$$

Suppose that $b(t)$ is a concave, increasing, positive continuous function on $[m, \infty)$ such that $\lim_{t \rightarrow \infty} b(t)/t = 0$ and $P[\tilde{S}_n < b(n) \text{ for all } n \geq m] > 0$. Then the equation $\theta t = b(t)$ has a unique root $t = g(\theta) > m$ for all sufficiently small positive θ , and under certain weak regularity conditions, we shall show in Section 2 that

$$(1.7) \quad \lim_{\theta \downarrow 0} ET(\theta)/g(\theta) = P[\tilde{S}_n < b(n) \text{ for all } n \geq m].$$

Since power-one tests of $H_0: \theta \leq 0$ versus $H_1: \theta > 0$ are of particular interest when detection of a small positive value of θ is important, it is desirable to know the expected number of observations required for such detection when θ is small, and (1.7) gives us a general asymptotic formula for evaluating this expected sample size.

The result (1.7) also suggests a new class of problems in the field of extended renewal theory and first passage times. For example, it would be interesting to compare (1.7) with corresponding results for lower-class boundaries. The particular lower-class boundary $b(n) = 0$ has been studied by Lan [12] who has shown that if X_1, X_2, \dots are i.i.d. with $EX_1 = 0$ and $EX_1^2 = \sigma^2 > 0$ and if $T^*(\theta) = \inf \{n \geq m : S_n + n\theta \geq 0\}$, then

$$(1.8) \quad \lim_{\theta \downarrow 0} \theta ET^*(\theta) = 2^{-1/2} \sigma \exp\{\sum_{n=1}^{\infty} n^{-1}(P[S_n < 0] - \frac{1}{2})\} < \infty.$$

In the case of an exponential family of distributions, as will be shown in Section 3, the analogue of (1.7) becomes

$$(1.9) \quad \lim_{\theta \downarrow 0} E_{\theta} T/g(\mu_{\theta}) = P_0[T = \infty],$$

where $\mu_{\theta} = E_{\theta} X_1$ and the family is normalized so that $\mu_0 = 0$. For the iterated-logarithm boundary defined by (1.5),

$$(1.10) \quad g(\mu_{\theta}) \sim \mu_{\theta}^{-2}(2\sigma^2 \log_2 \mu_{\theta}^{-1}) \quad \text{as } \theta \downarrow 0.$$

The above first-order asymptotic approximation does not involve the constant c of (1.5). In view of (1.9), (1.10) and the law of the iterated logarithm, one may suspect that power-one tests using iterated-logarithm boundaries as defined in (1.5) have asymptotically minimal (up to the first-order approximation) $E_{\theta} T$ as $\theta \downarrow 0$ among all tests of the form (1.2) having the same significance level. In fact, it has been shown by Farrell [6] that for any stopping rule τ such that $P_0[\tau = \infty] > 0$,

$$(1.11) \quad \limsup_{\theta \downarrow 0} \mu_{\theta}^2 E_{\theta} \tau / (2\sigma^2 \log_2 \mu_{\theta}^{-1}) \geq P_0[\tau = \infty].$$

While τ in (1.11) need not be of the form (1.2), it suffices, however, to restrict

ourselves to the class of stopping rules of the form (1.2). In fact, from considerations of sufficiency and monotonicity for the exponential family, it can be shown that given any stopping rule τ such that $P_0[\tau = \infty] > 0$ and $P_\theta[\tau < \infty] = 1$ for $\theta > 0$, there exists a sequence $\beta(n)$ of constants such that the stopping rule $T = \inf \{n \geq 1 : S_n \geq \beta(n)\}$ satisfies $P_0[T = \infty] = P_0[\tau = \infty]$ and $P_\theta[T > n] \leq P_\theta[\tau > n]$ for all $n = 1, 2, \dots$ and $\theta > 0$. In particular $E_\theta T \leq E_\theta \tau$ for all $\theta > 0$. (See [18, pages 424–425] and [6, pages 48–55]).

2. Power-one tests for a location parameter. The main results of this section are contained in Theorems 1, 2 and Corollary 1 and are illustrated in Example 1. Let Ψ be a known distribution function such that $\int_{-\infty}^{\infty} |x| d\Psi(x) < \infty$. Let $F_\theta(x) = \Psi(x - \theta)$, $\theta \in \Theta$, i.e., θ is a location parameter. The family F_θ then is stochastically increasing. Suppose we want to test $H_0: \theta \leq \theta_0$ versus $H_1: \theta > \theta_0$. It suffices just to consider the case $\theta_0 = 0$ and $\int_{-\infty}^{\infty} x d\Psi(x) = 0$. In this case, we use the stopping rule $T = \inf \{n \geq m : \sum_1^n X_i \geq b(n)\}$, where $b(n)$ is a sequence of positive numbers such that $\lim_{n \rightarrow \infty} b(n)/n = 0$ and $P_0[T = \infty] > 0$. Let Z_1, Z_2, \dots be i.i.d. with distribution function Ψ and let $T(\theta) = \inf \{n \geq m : \sum_1^n Z_i + n\theta \geq b(n)\}$. Obviously the distribution of $T(\theta)$ is the same as the distribution of T under P_θ . As indicated in Section 1, $E_\theta T = \infty$ if $\theta \leq 0$ and $E_\theta T$ is a finite nonincreasing function of θ for $\theta > 0$. From the definition of $T(\theta)$, it is easy to see that $E_\theta T$ is right continuous in θ for $\theta > 0$, and if $\Psi(x)$ is continuous, then $E_\theta T$ is continuous for $\theta > 0$. Considering the random variables $T(\theta)$, we observe that $T(0) = \infty$ implies that $\lim_{\theta \downarrow 0} T(\theta) = \infty$, and so $E_\theta T \geq ET(\theta)I_{[T(0)=\infty]} \rightarrow \infty$ as $\theta \downarrow 0$ since $P_0[T = \infty] > 0$. The following theorem studies the asymptotic behavior of $E_\theta T$ as $\theta \downarrow 0$.

THEOREM 1. *Suppose Z_1, Z_2, \dots are i.i.d. random variables such that $EZ_1 = 0$ and $E(Z_1^+)^{\nu} < \infty$ for some $\nu > 1$. Let $S_n = Z_1 + \dots + Z_n$. Let $b(t)$ be a positive continuous function on $[m, \infty)$ with $m \geq 1$ satisfying the following conditions:*

- (2.1) $b(t)$ is concave, increasing and $\lim_{t \rightarrow \infty} b(t)/t = 0$;
 - (2.2) For $\varepsilon_0 \leq \varepsilon < 1$ (where ε_0 is some positive number < 1), $\beta(\varepsilon) = \lim_{t \rightarrow \infty} b(\varepsilon t)/b(t)$ exists and is $> \varepsilon$, and $\lim_{\varepsilon \downarrow 1} \beta(\varepsilon) = 1$;
 - (2.3) $P[S_n < b(n)$ for all $n \geq m] > 0$;
 - (2.4) $\lim_{t \rightarrow \infty} t^{-1/\nu} b(t) = \infty$;
- For any $0 < \rho < 1$,
- (2.5) $\lim_{n \rightarrow \infty} P[S_i < b(i) - i\rho b(n)/n$ for all $n \geq i \geq m]$
 $= P[S_i < b(i)$ for all $i \geq m]$.

For each $\theta > 0$ such that $\theta m < b(m)$, let $t = g(\theta)$ be the root of $\theta t = b(t)$ with $t > m$, and let $T(\theta) = \inf \{n \geq m : S_n + n\theta \geq b(n)\}$. Then

(2.6) $\lim_{\theta \downarrow 0} ET(\theta)/g(\theta) = P[S_n < b(n)$ for all $n \geq m]$.

REMARKS. (i) If $b(t)$ is any continuous concave function on $[m, \infty)$ such that $\lim_{t \rightarrow \infty} b(t) = \infty$, then for $0 < \varepsilon < 1$, $(b(\varepsilon t) - b(m))/(\varepsilon t - m) \geq (b(t) - b(m))/(t - m)$ and so $\liminf_{t \rightarrow \infty} b(\varepsilon t)/b(t) \geq \varepsilon$. Condition (2.2) in Theorem 1 requires that $\lim_{t \rightarrow \infty} b(\varepsilon t)/b(t)$ actually exists and is $> \varepsilon$. If there exists $\alpha \in (0, 1)$ such that $b(t) = t^\alpha U(t) + O(t^\alpha)$, where $U(t)$ is a slowly varying function such that $\lim_{t \rightarrow \infty} U(t) = \infty$, or if $b(t) = ct^\alpha + o(t^\alpha)$, where $c > 0$, then (2.2) holds.

(ii) Suppose Z_1, Z_2, \dots are i.i.d. with $EZ_1 = 0, 0 < EZ_1^2 < \infty$. Let $b(t)$ be a positive continuous function on $[m, \infty)$ ($m \geq 1$) satisfying conditions (2.1), (2.2), (2.3) of Theorem 1. Suppose further that $b(t)$ satisfies

$$(2.7) \quad \lim_{t \rightarrow \infty} b(t)/(t \log_2 t)^{\frac{1}{2}} = \infty .$$

Then condition (2.5) holds. To see this, we note that for $0 < \rho < 1$ and $i_0 (> m)$ sufficiently large, if $n \geq i \geq i_0$, then

$$\begin{aligned} b(i) - b(m) &\geq \{(i - m)/(n - m)\}(b(n) - b(m)) \\ &\geq \frac{1}{2}(1 + \rho)ib(n)/n, \end{aligned}$$

and so $i\rho b(n)/n \leq 2\rho b(i)/(1 + \rho)$ for $n \geq i \geq i_0$. Therefore in view of (2.7), the law of the iterated logarithm implies that given $\varepsilon > 0$, we can choose $n_0 \geq i_0$ such that for $n > n_0$,

$$\begin{aligned} P[S_i \geq b(i) - i\rho b(n)/n \text{ for some } n \geq i \geq n_0] \\ \leq P[S_i \geq (1 - \rho)b(i)/(1 + \rho) \text{ for some } i \geq n_0] < \varepsilon. \end{aligned}$$

Since $\lim_{n \rightarrow \infty} b(n)/n = 0$, it is now clear that (2.5) holds. Noting that $EZ_1^2 < \infty$, we can set $\nu = 2$ in Theorem 1 so that (2.7) also implies (2.4).

(iii) The growth condition (2.7) in (ii) rules out the most delicate (such as given by (1.5)) among the upper-class boundaries. To show that condition (2.5) still holds for these boundaries, we shall use a more delicate argument by considering the Wiener process and using the Skorohod embedding. This argument, which will be given in detail after the proof of Theorem 1, yields the following corollary of Theorem 1.

COROLLARY 1. Suppose Z_1, Z_2, \dots are i.i.d. nondegenerate random variables with $EZ_1 = 0$ and $E\{Z_1 \log_2(|Z_1| + e)\}^2 < \infty$. Let $b(t)$ be a positive continuous function on $[m, \infty)$ belonging to the upper class so that (2.3) holds. Assume that $b(t)$ satisfies the regularity conditions (2.1), (2.2) and

$$(2.8) \quad t^{-\frac{1}{2}}b(t) \text{ is ultimately nondecreasing.}$$

Define $T(\theta)$ and $g(\theta)$ as in Theorem 1. Then (2.6) holds.

We now proceed to prove Theorem 1. First we establish the following lemma:

LEMMA 1. Let $b(t)$ be a positive continuous function on $[m, \infty)$ ($m \geq 1$) such that $\lim_{t \rightarrow \infty} b(t) = \infty$ and conditions (2.1), (2.2) of Theorem 1 are satisfied. Let the function $g(\theta)$ be as defined in Theorem 1. Then

- (a) $g(\theta)$ is a strictly decreasing function and $\lim_{\theta \downarrow 0} g(\theta) = \infty$.
- (b) Let $f: (0, a) \rightarrow [m, \infty)$ where a is any positive number. Then

$$\begin{aligned} \liminf_{\theta \downarrow 0} b(f(\theta))/(\theta f(\theta)) \geq 1 &\implies \limsup_{\theta \downarrow 0} f(\theta)/g(\theta) \leq 1, \\ \limsup_{\theta \downarrow 0} b(f(\theta))/(\theta f(\theta)) \leq 1 &\implies \liminf_{\theta \downarrow 0} f(\theta)/g(\theta) \geq 1, \\ b(f(\theta)) \sim \theta f(\theta) \text{ as } \theta \downarrow 0 &\implies f(\theta) \sim g(\theta) \text{ as } \theta \downarrow 0. \end{aligned}$$

(c) Take any $\theta^* > 0$ such that $\theta^*m < b(m)$. Then there exists $\phi: (0, \theta^*) \rightarrow [m, \infty)$ such that $\lim_{\theta \downarrow 0} \phi(\theta) = \infty$, $\lim_{\theta \downarrow 0} \phi(\theta)/g(\theta) = 0$ and $\lim_{\theta \downarrow 0} b(\phi(\theta))/b(g(\theta)) = 0$.

(d) Let $b^*(t)$ be any positive, continuous, strictly increasing and concave function on $[m^*, \infty)$ with $m^* \geq m$ such that $b^*(t) \sim b(t)$ as $t \rightarrow \infty$. For all $\theta > 0$ such that $\theta m^* < b^*(m^*)$, let $t = g^*(\theta)$ be the root of $\theta t = b^*(t)$ with $t > m^*$. Then $g^*(\theta) \sim g(\theta)$ as $\theta \downarrow 0$.

(e) There exists a function $\gamma: [\rho_1, 1) \rightarrow (0, 1)$ where ρ_1 is some positive number < 1 such that $\limsup_{\theta \downarrow 0} \gamma(\rho)g(\rho\theta)/g(\theta) \leq 1$ for all $\rho \in [\rho_1, 1)$ and $\lim_{\rho \uparrow 1} \gamma(\rho) = 1$.

PROOF. (a) is obvious. We shall only prove the second implication of (b), since the first implication can be proved similarly and the third implication follows from the first two implications. Suppose that $\limsup_{\theta \downarrow 0} b(f(\theta))/(\theta f(\theta)) \leq 1$. Then $\lim_{\theta \downarrow 0} f(\theta) = \infty$ and so $\lim_{\theta \downarrow 0} b(f(\theta)) = \infty = \lim_{\theta \downarrow 0} \theta f(\theta)$. Assume that $\liminf_{\theta \downarrow 0} f(\theta)/g(\theta) < 1$. Then there exist a positive sequence $\theta_n \downarrow 0$ and $1 > \varepsilon > \varepsilon_0$ such that $m < f(\theta_n) < \varepsilon g(\theta_n)$. Using (2.2) and the concavity of $b(t)$, we obtain that

$$\begin{aligned} 1 &< \lim_{n \rightarrow \infty} \{b(\varepsilon g(\theta_n)) - b(m)\} / \{\varepsilon b(g(\theta_n)) - m\theta_n\} \\ &= \lim_{n \rightarrow \infty} \{b(\varepsilon g(\theta_n)) - b(m)\} / \{\theta_n(\varepsilon g(\theta_n) - m)\} \\ &\leq \liminf_{n \rightarrow \infty} \{b(f(\theta_n)) - b(m)\} / \{\theta_n(f(\theta_n) - m)\}, \end{aligned}$$

contradicting the assumption that $1 \geq \limsup_{\theta \downarrow 0} b(f(\theta))/(\theta f(\theta))$.

To prove (c), define $\pi(x) = \inf \{y \geq m: b(y) \geq (b(x))^{\frac{1}{2}}\}$ for $x > m$. Then $\lim_{x \rightarrow \infty} \pi(x) = \infty$ and $b(\pi(x)) = (b(x))^{\frac{1}{2}}$ for all large x . By the concavity of b , $\lim_{x \rightarrow \infty} \pi(x)/x = 0$. Let $\phi(\theta) = \pi(g(\theta))$. Then ϕ satisfies the desired conclusions. To prove (d), we observe that as $\theta \downarrow 0$, $g^*(\theta) \rightarrow \infty$ and $b(g^*(\theta)) \sim b^*(g^*(\theta)) = \theta g^*(\theta)$, and so $g^*(\theta) \sim g(\theta)$ by (b).

We now prove (e). Let $\rho_0 = \varepsilon_0/\beta(\varepsilon_0) < 1$. For $\rho \in [\rho_0, 1)$, let $\tilde{\gamma}(\rho) = \sup \{\varepsilon: \varepsilon_0 \leq \varepsilon < 1 \text{ and } \rho\beta(\varepsilon) \geq \varepsilon\}$. Since $\lim_{\varepsilon \uparrow 1} \varepsilon/\beta(\varepsilon) = 1 > \rho$, $\tilde{\gamma}(\rho) < 1$. It is easy to see that $\lim_{\rho \uparrow 1} \tilde{\gamma}(\rho) = 1$. Therefore we can choose $\rho_1 \in [\rho_0, 1)$ such that for all $\rho \in [\rho_1, 1)$, $2\tilde{\gamma}(\rho) - 1 > \varepsilon_0$. For each $\rho \in [\rho_1, 1)$, we can choose $\gamma(\rho) \in (2\tilde{\gamma}(\rho) - 1, \tilde{\gamma}(\rho)]$ such that $\rho\beta(\gamma(\rho)) \geq \gamma(\rho)$. Obviously $\lim_{\rho \uparrow 1} \gamma(\rho) = 1$. By (2.2), as $\theta \downarrow 0$,

$$b(\gamma(\rho)g(\rho\theta)) \sim \beta(\gamma(\rho))b(g(\rho\theta)) = \rho\theta\beta(\gamma(\rho))g(\rho\theta) \geq \theta\gamma(\rho)g(\rho\theta),$$

and so $\limsup_{\theta \downarrow 0} \gamma(\rho)g(\rho\theta)/g(\theta) \leq 1$ by (b). \square

PROOF OF THEOREM 1. Set $P^* = P[S_i < b(i) \text{ for all } i \geq m]$. We shall first show that

$$(2.9) \quad \liminf_{\theta \downarrow 0} ET(\theta)/g(\theta) \geq P^*.$$

For $n > m$, define $c_n = (b(n) - b(m))/(n - m)$. Then c_n is nonincreasing and $\lim_{n \rightarrow \infty} c_n = 0$. Let ρ_1 and γ be given by Lemma 1(e). Take any $\rho \in [\rho_1, 1)$ and any sequence θ_n such that $\rho c_{n+1} \leq \theta_n \leq \rho c_n$. Obviously

$$P[T(\theta_n) > n] = P[S_i + i\theta_n < b(i) \text{ for all } n \geq i \geq m] \\ \leq P[S_i < b(i) \text{ for all } n \geq i \geq m] = P^* + o(1) \quad \text{as } n \rightarrow \infty.$$

The key to our proof of (2.9) lies in the fact that the reverse inequality also holds as $n \rightarrow \infty$ so that

$$(2.10) \quad \lim_{n \rightarrow \infty} P[T(\theta_n) > n] = P^*.$$

To see this, we take $\rho' \in (\rho, 1)$ and note that for all large n ,

$$P[T(\theta_n) > n] \geq P[S_i < b(i) - i\rho'b(n)/n \text{ for all } n \geq i \geq m] \\ \rightarrow P^* \quad \text{as } n \rightarrow \infty \text{ by condition (2.5).}$$

Since $(n + 1 - m)c_{n+1} = b(n + 1) - b(m)$, Lemma 1(d) implies that $g(c_{n+1}) \sim n$. Using this and (2.10), we obtain that

$$(2.11) \quad ET(\theta_n) \geq nP[T(\theta_n) > n] \sim g(c_{n+1})P^* \geq (1 + o(1))\gamma(\rho)g(\theta_n)P^*.$$

To see the last inequality above, we note that $c_{n+1} \leq \theta_n/\rho$ implies that $g(c_{n+1}) \geq g(\theta_n/\rho)$ and apply Lemma 1(e). Therefore we have proved that for any $\rho \in [\rho_1, 1)$ and any sequence θ_n such that $\rho c_{n+1} \leq \theta_n \leq \rho c_n$, (2.11) holds. This implies that $\liminf_{\theta \downarrow 0} ET(\theta)/g(\theta) \geq \gamma(\rho)P^*$. Since $\lim_{\rho \uparrow 1} \gamma(\rho) = 1$, we have established (2.9).

For $\theta > 0$, $S_{T(\theta)} + \theta T(\theta) \leq b(T(\theta)) + X_{T(\theta)}^+ + \theta + |S_m| + m\theta$. Therefore applying Wald's lemma, we have

$$(2.12) \quad \theta ET(\theta) \leq Eb(T(\theta)) + E^{1/\nu}(X_{T(\theta)}^+)^{\nu} + \theta + m(E|X_1| + \theta).$$

Now $E(X_{T(\theta)}^+)^{\nu} \leq E \sum_{i=1}^{T(\theta)} (X_i^+)^{\nu} = (ET(\theta))E(X_1^+)^{\nu}$. From condition (2.4), it follows that $\lim_{\theta \downarrow 0} b(g(\theta))/(g(\theta))^{1/\nu} = \infty$, and so $\lim_{\theta \downarrow 0} \theta(g(\theta))^{1-1/\nu} = \infty$, implying that $\lim_{\theta \downarrow 0} \theta(ET(\theta))^{1-1/\nu} = \infty$ in view of (2.9). Hence $\lim_{\theta \downarrow 0} \theta ET(\theta) = \infty$ and $E^{1/\nu}(X_{T(\theta)}^+)^{\nu} = O(E^{1/\nu}T(\theta)) = o(\theta ET(\theta))$. Putting this in (2.12), we obtain

$$(2.13) \quad \theta ET(\theta)(1 + o(1)) \leq Eb(T(\theta)).$$

Let ϕ be the function constructed in Lemma 1(c). We note that

$$(2.14) \quad Eb(T(\theta)) \leq E[b(T(\theta)) | T(\theta) > \phi(\theta)]P[T(\theta) > \phi(\theta)] \\ + b(\phi(\theta))P[T(\theta) \leq \phi(\theta)] \\ \leq b(E[T(\theta) | T(\theta) > \phi(\theta)])P[T(\theta) > \phi(\theta)] + o(b(g(\theta))).$$

The last relation above follows from Jensen's inequality (since b is concave) and Lemma 1(c). Now $b(g(\theta)) = \theta g(\theta) = O(\theta ET(\theta))$ by (2.9). Hence letting $f(\theta) = E[T(\theta) | T(\theta) > \phi(\theta)]$, we obtain that

$$(2.15) \quad 1 \leq \liminf_{\theta \downarrow 0} b(f(\theta))P[T(\theta) > \phi(\theta)]/(\theta ET(\theta)), \quad \text{by (2.13) and (2.14),} \\ \leq \liminf_{\theta \downarrow 0} b(f(\theta))/(\theta f(\theta)), \quad \text{since } ET(\theta) \geq f(\theta)P[T(\theta) > \phi(\theta)].$$

By Lemma 1(b), (2.15) implies that

$$(2.16) \quad \limsup_{\theta \downarrow 0} f(\theta)/g(\theta) \leq 1 .$$

Since $\lim_{\theta \downarrow 0} \phi(\theta)/g(\theta) = 0$ by Lemma 1(c), we obtain using (2.16) that

$$(2.17) \quad \begin{aligned} ET(\theta) &\leq \phi(\theta) + f(\theta)P[T(\theta) > \phi(\theta)] \\ &\leq o(g(\theta)) + (1 + o(1))g(\theta)P[T(0) > \phi(\theta)] . \end{aligned}$$

Noting that $\lim_{\theta \downarrow 0} P[T(0) > \phi(\theta)] = P[T(0) = \infty] = P^*$, the desired conclusion (2.6) follows from (2.9) and (2.17). \square

PROOF OF COROLLARY 1. Without loss of generality, we shall assume that $EZ_1^2 = 1$. Since b belongs to the upper class and (2.8) holds, $\lim_{t \rightarrow \infty} t^{-\frac{1}{2}}b(t) = \infty$ by the Feller-Kolmogorov integral test (cf. [7, page 132]), i.e., (2.4) holds if we set $\nu = 2$ in Theorem 1. Therefore it remains to show that condition (2.5) is also satisfied. For $0 < \rho < 1$,

$$(2.18) \quad \begin{aligned} P[S_i \geq b(i) - i\rho b(n)/n \text{ for some } n \geq i \geq m] \\ \geq P[S_i \geq b(i) \text{ for some } n \geq i \geq m] \\ \rightarrow P[S_i \geq b(i) \text{ for some } i \geq m] \quad \text{as } n \rightarrow \infty . \end{aligned}$$

For each n , let $k(n)$ be a positive integer $\geq m$ such that

$$(2.19) \quad \lim_{n \rightarrow \infty} k(n) = \infty \quad \text{and} \quad \lim_{n \rightarrow \infty} k(n)b(n)/n = 0 .$$

Given $\varepsilon > 0$, since $b(t) - 1$ obviously also belongs to the upper class, we can choose n_0 such that $P[S_i \geq b(i) - 1 \text{ for } i \geq n_0] \leq \varepsilon$ and $\rho b(n)k(n)/n \leq 1$ for $n \geq n_0$. Then for $n \geq n_0$,

$$(2.20) \quad \begin{aligned} P[S_i \geq b(i) - i\rho b(n)/n \text{ for some } m \leq i \leq k(n)] \\ \leq P[S_i \geq b(i) - \rho n_0 b(n)/n \text{ for some } m \leq i \leq n_0] + \varepsilon \\ = P[S_i \geq b(i) \text{ for some } m \leq i \leq n_0] + o(1) + \varepsilon \quad (\text{as } n \rightarrow \infty) \\ \leq P[S_i \geq b(i) \text{ for some } i \geq m] + \varepsilon + o(1) . \end{aligned}$$

In view of (2.18) and (2.20), it suffices to show that

$$(2.21) \quad \lim_{n \rightarrow \infty} P[S_i \geq b(i) - i\rho b(n)/n \text{ for some } k(n) \leq i \leq n] = 0 .$$

Since $EZ_1 = 0$, $EZ_1^2 = 1$ and $E\{Z_1 \log_2(|Z_1| + e)\}^2 < \infty$, by the Skorohod embedding theorem (cf. [7, page 128]), we have (by redefining the random variables on a new probability space if necessary)

$$(2.22) \quad \lim_{n \rightarrow \infty} |S_n - S_n^*|/\{n^{\frac{1}{2}}(\log_2 n)^{-\frac{1}{2}}\} = 0 \quad \text{a.s.},$$

where $S_n^* = Y_1 + \dots + Y_n$ and Y_1, Y_2, \dots are i.i.d. standard normal random variables. Take any $\lambda > 0$ and let $\beta(t) = b(t) - \lambda t^{\frac{1}{2}}(\log_2 t)^{-\frac{1}{2}}$. A standard argument involving the Feller-Kolmogorov test shows that $\beta(t)$ also belongs to the upper class. Therefore by Lemma 2 below, letting $\rho < \rho' < 1$,

$$(2.23) \quad \begin{aligned} \lim_{n \rightarrow \infty} P[S_i^* + i\rho'(b(n) - b(m))/(n - m) \geq \beta(i) \\ \text{for some } k(n) \leq i \leq n] = 0 . \end{aligned}$$

From (2.22) and (2.23), the desired conclusion (2.21) follows. \square

LEMMA 2. Suppose that under P_θ , X_1, X_2, \dots are i.i.d. $N(\theta, 1)$ random variables. Set $S_n = X_1 + \dots + X_n$. Let $\beta(n)$ be a positive sequence such that $\lim_{n \rightarrow \infty} \beta(n)/n = 0$, $\lim_{n \rightarrow \infty} n^{-\frac{1}{2}}\beta(n) = \infty$ and $P_0[S_n < \beta(n) \text{ for all large } n] = 1$. Let $c(n)$ be a positive nonincreasing sequence such that $\lim_{n \rightarrow \infty} c(n) = 0$ and $\lim_{n \rightarrow \infty} nc(n)/\beta(n) = 1$. Suppose $k(n) \leq n$ is a sequence of positive integers such that $\lim_{n \rightarrow \infty} k(n) = \infty$. Then for any $\rho \in (0, 1)$,

$$(2.24) \quad \lim_{n \rightarrow \infty} P_{\rho c(n)}[S_i \geq \beta(i) \text{ for some } k(n) \leq i \leq n] = 0.$$

PROOF. We shall use the ideas of Farrell’s proof of Lemma 4 in [6]. Let $m(n)$ be a sequence of nonnegative integers such that

$$(2.25) \quad \lim_{n \rightarrow \infty} m(n) = \infty, \quad \lim_{n \rightarrow \infty} c(m(n))/c(n) = \infty \quad \text{and} \\ \lim_{n \rightarrow \infty} c(n)(m(n))^{\frac{1}{2}} = \infty.$$

Such a sequence can be constructed as follows. Set $d_n = n^{\frac{1}{2}}c(n)$. Then $d_n \rightarrow \infty$ since $nc(n) \sim \beta(n)$. Let α_n be a decreasing sequence of positive numbers such that $\lim_{n \rightarrow \infty} \alpha_n = 0$ and $\lim_{n \rightarrow \infty} \alpha_n d_n = \infty$. Define $m(n) = \sup\{m : c(m)/c(n) \geq \alpha_n d_m\}$ ($\sup \emptyset = 0$). Then $m(n)$ satisfies (2.5).

Let $T_n = \inf\{i \geq k(n) : S_i \geq \beta(i)\}$. We now make use of a standard likelihood-ratio argument in sequential analysis. If $k(n) \leq m(n)$, we have

$$(2.26) \quad P_\theta[T_n \leq m(n)] = \sum_{i=k(n)}^{m(n)} \int_{[T_n=i]} e^{\theta S_i - i\theta^2/2} dP_0.$$

Hence

$$(2.27) \quad \frac{d^2}{d\theta^2} P_\theta[T_n \leq m(n)] = \sum_{i=k(n)}^{m(n)} \int_{[T_n=i]} \{(S_i - i\theta)^2 - i\} e^{\theta S_i - i\theta^2/2} dP_0.$$

Choose n_1 such that for $i \geq n_1$, $\beta(i) > \rho ic(i) + i^{\frac{1}{2}}$. Therefore if $i > n_1$, $\theta \leq \rho c(i)$ and $x \geq \beta(i)$, then $(x - i\theta)^2 - i = (x - i\theta - i^{\frac{1}{2}})(x - i\theta + i^{\frac{1}{2}}) > 0$. Since $S_i \geq \beta(i)$ on the event $[T_n = i]$ and $c(i) \geq c(j)$ if $i \leq j$, it then follows from (2.27) that we can choose n_2 such that for $n \geq n_2$, $P_\theta[T_n \leq m(n)]$ is a convex function of θ in the interval $[0, \rho c(m(n))]$. Therefore by (2.25), for all large n , $c(n) < c(m(n))$ and

$$(2.28) \quad P_{\rho c(n)}[T_n \leq m(n)] \\ \leq (c(n)/c(m(n)))P_{\rho c(m(n))}[T_n \leq m(n)] \\ + \{1 - c(n)/c(m(n))\}P_0[S_i \geq \beta(i) \text{ for some } k(n) \leq i \leq m(n)] \\ \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Clearly $m(n) < n$ for large n . Again by using likelihood ratios, we obtain

$$(2.29) \quad P_{\rho c(n)}[m(n) \leq T_n \leq n] \\ = \sum_{i=m(n)}^n \int_{[T_n=i]} \exp\left\{- (1 - \rho)c(n)S_i + \frac{i}{2}(1 - \rho^2)c^2(n)\right\} dP_{c(n)} \\ \leq \max_{m(n) \leq i \leq n} \exp\left\{- (1 - \rho)c(n)\left(\beta(i) - \frac{i}{2}(1 + \rho)c(n)\right)\right\},$$

since $S_i \geq \beta(i)$ on the event $[T_n = i]$. Let $1 > \varepsilon > (1 + \rho)/2$. Then for all large i , $\beta(i)/i > \varepsilon c(i) \geq \varepsilon c(n)$ if $n \geq i$. Hence letting $\delta = (1 - \rho)\{\varepsilon - \frac{1}{2}(1 + \rho)\}$, we have for all large n and $m(n) \leq i \leq n$,

$$(2.30) \quad \exp\left\{-(1 - \rho)c(n)\left(\beta(i) - \frac{i}{2}(1 + \rho)c(n)\right)\right\} \leq \exp\{-\delta c^2(n)m(n)\} \rightarrow 0$$

in view of (2.25). From (2.28), (2.29) and (2.30), the desired conclusion (2.24) follows. \square

EXAMPLE 1. Let Z_1, Z_2, \dots be i.i.d. with $EZ_1 = 0$, $0 < EZ_1^2 < \infty$ and let $S_n = Z_1 + \dots + Z_n$. Let $b(t)$ be a concave, increasing, positive continuous function on $[m, \infty)$ such that $P[S_n < b(n)$ for all $n \geq m] = P^* > 0$. Define $T(\theta)$ as in Theorem 1.

(a) If $b(t) \sim (t \log t)^\frac{1}{2}$, then by Theorem 1 (see Remarks (i) and (ii)),

$$\lim_{\theta \downarrow 0} \theta^2 ET(\theta)/(2|\log \theta|) = P^* .$$

(b) Let $\frac{1}{2} < \alpha < 1$. If $b(t) \sim t^\alpha$, then again by Theorem 1,

$$\lim_{\theta \downarrow 0} \theta^{1/(1-\alpha)} ET(\theta) = P^* .$$

The analogue of Theorem 1 for stopping rules of two-sided power-one tests also holds. More specifically, we have the following theorem.

THEOREM 2. Suppose Z_1, Z_2, \dots are i.i.d. random variables such that $EZ_1 = 0$ and $E|Z_1|^\nu < \infty$ for some $\nu > 1$. Let $S_n = Z_1 + \dots + Z_n$. For $i = 1, 2$, let $b_i(t)$ be a positive continuous function on $[m, \infty)$ with $m \geq 1$ satisfying the following conditions:

$$(2.31) \quad b_i(t) \text{ is concave, increasing and } \lim_{t \rightarrow \infty} b_i(t)/t = 0 ;$$

$$(2.32) \quad \text{For } \varepsilon_0 \leq \varepsilon < 1 \text{ (where } \varepsilon_0 \text{ is some positive number } < 1), \\ \beta_i(\varepsilon) = \lim_{t \rightarrow \infty} b_i(\varepsilon t)/b_i(t) \text{ exists and is } > \varepsilon, \text{ and } \lim_{\varepsilon \uparrow 1} \beta_i(\varepsilon) = 1 ;$$

$$(2.33) \quad P[-b_2(n) < S_n < b_1(n) \text{ for all } n \geq m] > 0 ;$$

$$(2.34) \quad \lim_{t \rightarrow \infty} t^{-1/\nu} b_i(t) = \infty ;$$

For any $0 < \rho < 1$,

$$(2.35) \quad \lim_{n \rightarrow \infty} P[-b_2(k) < S_k < b_1(k) - k\rho b_1(n)/n \text{ for all } n \geq k \geq m] \\ = \lim_{n \rightarrow \infty} P[-b_2(k) + k\rho b_2(n)/n < S_k < b_1(k) \text{ for all } n \geq k \geq m] \\ = P[-b_2(k) < S_k < b_1(k) \text{ for all } k \geq m] .$$

$$(2.36) \quad P[S_n = b_1(n)] = P[S_n = -b_2(n)] = 0 \quad \text{for all } n \geq m .$$

For each $\theta > 0$ such that $\theta m < b_1(m)$, let $t = g_1(\theta)$ be the root of $\theta t = b_1(t)$ with $t > m$, while for each $\theta < 0$ such that $\theta m > -b_2(m)$, let $t = g_2(\theta)$ be the root of $\theta t = -b_2(t)$ with $t > m$. For any real θ , let

$$N(\theta) = \inf \{n \geq m : S_n + n\theta \geq b_1(n) \text{ or } S_n + n\theta \leq -b_2(n)\} .$$

Then $EN(\theta) < \infty$ if $\theta \neq 0$, $EN(0) = \infty$, $\lim_{\theta \rightarrow \infty} EN(\theta) = m$, and

$$\begin{aligned} \lim_{\theta \rightarrow 0+} EN(\theta)/g_1(\theta) &= \lim_{\theta \rightarrow 0-} EN(\theta)/g_2(\theta) \\ &= P[-b_2(n) < S_n < b_1(n) \text{ for all } n \geq m]. \end{aligned}$$

PROOF. We shall only restrict ourselves to the asymptotic behavior as $\theta \rightarrow 0+$ since the argument for $\theta \rightarrow 0-$ is similar. Set $p = P[-b_2(i) < S_i < b_1(i) \text{ for all } i \geq m]$. For $n > m$, define $c_n = (b_1(n) - b_1(m))/(n - m)$. Let ρ_1 and γ be given by Lemma 1(e) with $g = g_1$. Take any $\rho \in [\rho_1, 1)$ and any sequence θ_n such that $\rho c_{n+1} \leq \theta_n \leq \rho c_n$. We shall first show that

$$(2.37) \quad \lim_{n \rightarrow \infty} P[N(\theta_n) > n] = p.$$

By condition (2.35), taking $\rho' \in (\rho, 1)$, we obtain that for all large n ,

$$\begin{aligned} P[N(\theta_n) > n] &\geq P[-b_2(i) < S_i < b_1(i) - i\rho'b_1(n)/n \text{ for all } n \geq i \geq m] \\ &\rightarrow p \quad \text{as } n \rightarrow \infty. \end{aligned}$$

On the other hand, in view of (2.33), given $\varepsilon > 0$, we can choose $n_0 > m$ such that $P[S_i \leq -b_2(i) \text{ for some } i \geq n_0] < \varepsilon$, and so for $n \geq n_0$,

$$\begin{aligned} P[N(\theta_n) > n] &\leq P[-b_2(i) - i\rho c_n < S_i < b_1(i) \text{ for all } n \geq i \geq m] \\ &\leq P[-b_2(i) < S_i < b_1(i) \text{ for all } n \geq i \geq m] \\ &\quad + P[-b_2(i) - i\rho c_n < S_i \leq -b_2(i) \text{ for some } n_0 \geq i \geq m] + \varepsilon \\ &\rightarrow p + \varepsilon \quad \text{as } n \rightarrow \infty \text{ by condition (2.36)}. \end{aligned}$$

Since ε is arbitrary, we have established (2.37). Making use of (2.37), we can then prove as in Theorem 1 that

$$(2.38) \quad \liminf_{\theta \downarrow 0} EN(\theta)/g_1(\theta) \geq p.$$

Let $\psi(\theta)$ be as constructed in Lemma 1(c) with $b = b_1$ and $g = g_1$. Set $f(\theta) = E[N(\theta) | N(\theta) > \psi(\theta)]$. Noting that for $\theta > 0$, $S_{N(\theta)} + \theta N(\theta) \leq b_1(N(\theta)) + X_{N(\theta)}^+ + \theta + |S_m| + m\theta$, we can apply Wald's lemma and Jensen's inequality as in the proof of Theorem 1 to show that

$$(2.39) \quad \limsup_{\theta \downarrow 0} f(\theta)/g_1(\theta) \leq 1.$$

Define $N_+(\theta) = \inf \{n \geq m : S_n + n\theta \geq b_1(n)\}$, $N_-(\theta) = \inf \{n \geq m : S_n + n\theta \leq -b_2(n)\}$. Then by (2.39) and Lemma 1(c), we have for $\theta > 0$,

$$(2.40) \quad \begin{aligned} EN(\theta) &\leq \psi(\theta) + f(\theta)P[N(\theta) > \psi(\theta)] \\ &\leq o(g_1(\theta)) + (1 + o(1))g_1(\theta)P[N_+(\theta) > \psi(\theta), N_-(\theta) > \psi(\theta)]. \end{aligned}$$

Observe that

$$\begin{aligned} 0 &\leq P[N_+(\theta) > \psi(\theta), N_-(\theta) > \psi(\theta)] - P[N_+(\theta) > \psi(\theta), N_-(\theta) > \psi(\theta)] \\ &\leq P[S_i > -b_2(i) - i\theta \text{ for all } m \leq i \leq \psi(\theta) \\ &\quad \text{and } S_i \leq -b_2(i) \text{ for some } m \leq i \leq \psi(\theta)] \\ &\rightarrow 0 \quad \text{as } \theta \downarrow 0 \text{ by (2.33) and (2.36)}. \end{aligned}$$

Since $\lim_{\theta \downarrow 0} P[N_+(\theta) > \phi(\theta), N_-(\theta) > \phi(\theta)] = p$, it follows from (2.40) that

$$(2.41) \quad \limsup_{\theta \downarrow 0} EN(\theta)/g_1(\theta) \leq p. \quad \square$$

Suppose Z_1, Z_2, \dots are i.i.d. with $EZ_1 = 0$ and $0 < EZ_1^2 < \infty$. For $i = 1, 2$, let $b_i(t)$ be a positive continuous function on $[m, \infty)$ satisfying conditions (2.31), (2.32), (2.33) and (2.36) of Theorem 2. Suppose further that

$$(2.42) \quad \lim_{t \rightarrow \infty} b_i(t)/(t \log_2 t)^{\frac{1}{2}} = \infty \quad \text{for } i = 1, 2.$$

Then condition (2.34) holds with $\nu = 2$ and condition (2.35) is also satisfied. (See Remark (ii) to Theorem 1.) Alternatively, instead of assuming (2.42), we can obtain the same conclusion by assuming that

$$(2.43) \quad E\{Z_1 \log_2 (|Z_1| + e)\}^2 < \infty \quad \text{and} \quad t^{-\frac{1}{2}}b_i(t) \quad \text{is ultimately nondecreasing for } i = 1, 2.$$

(See the proof of Corollary 1.) Hence Theorem 2 gives the analogue of Corollary 1 for $N(\theta)$.

3. The case of an exponential family. Suppose X_1, X_2, \dots are i.i.d. random variables having a common density $\exp(\theta x - h(\theta))$ with respect to some nondegenerate measure π on the real line, where $\theta \in \Theta$ is an unknown natural parameter of the exponential family and Θ is an open interval of the real line. Let $\mu_\theta = E_\theta X_1 = h'(\theta)$. For a given $\theta_0 \in \Theta$, we want to test the hypothesis $H: \theta = \theta_0$ versus $K: \theta \neq \theta_0$. By considering $X_i^* = X_i - h'(\theta_0)$ and $\theta^* = \theta - \theta_0$ if necessary, we may assume without loss of generality that $\theta_0 = 0$ and $\mu_0 = 0$. The following theorem studies the asymptotic behavior of the expected sample size as $\theta \rightarrow 0$ for power-one tests of H versus K based on sample sums and suitable stopping boundaries $b_1(n)$ and $-b_2(n)$.

THEOREM 3. *Suppose that under $P_\theta, X_1, X_2, \dots$ are i.i.d. with a common density $\exp(\theta x - h(\theta))$ with respect to a fixed nondegenerate measure π on the real line such that $\mu_0 = 0$, where $\mu_\theta = E_\theta X_1, \theta \in \Theta, \Theta$ being an open interval (possibly infinite) containing 0. Let $S_n = X_1 + \dots + X_n$. For $i = 1, 2$, let $b_i(t)$ be a positive continuous function of $[m, \infty)$ with $m \geq 1$ satisfying conditions (2.31), (2.32) of Theorem 2 and the following two conditions:*

$$(3.1) \quad \lim_{t \rightarrow \infty} t^{-\frac{1}{2}}b_i(t) = \infty \quad \text{for } i = 1, 2;$$

$$(3.2) \quad P_0[-b_2(n) < S_n < b_1(n) \text{ for all } n \geq m] > 0.$$

Let $N = \inf \{n \geq m : S_n \geq b_1(n) \text{ or } S_n \leq -b_2(n)\}$. Define $g_1(\theta)$ for $\theta > 0$ and $g_2(\theta)$ for $\theta < 0$ as in Theorem 2. Then $E_\theta N < \infty$ if $\theta \neq 0$ and

$$(3.3) \quad \lim_{\theta \rightarrow 0+} E_\theta N/g_1(\mu_\theta) = \lim_{\theta \rightarrow 0-} E_\theta N/g_2(\mu_\theta) = P_0[N = \infty].$$

PROOF. As shown in [6, page 40], $P_\theta[N < \infty, S_N \geq b_1(N)]$ is a nondecreasing and left continuous function of θ , while $P_\theta[N < \infty, S_N \leq -b_2(N)]$ is a non-increasing and right continuous function of θ . This fact will be used in the

proof. We shall only consider the case $\theta \rightarrow 0+$ in (3.3) as the proof for $\theta \rightarrow 0-$ is exactly analogous. Let $\sigma^2 = E_0 X_1^2$. As is well known, $\mu_\theta \sim \sigma^2 \theta$ as $\theta \rightarrow 0$.

For $n > m$, define $c_n = (b_1(n) - b_1(m))/\{\sigma^2(n - m)\}$. Let ρ_1 and γ be given by Lemma 1(e) with $g = g_1$. Take any $\rho \in [\rho_1, 1)$ and any sequence θ_n such that $\rho c_{n+1} \leq \theta_n \leq \rho c_n$. Like Farrell's proof of Lemma 4 of [6] (see also our proof of Lemma 2 above), it can be shown that

$$(3.4) \quad \lim_{n \rightarrow \infty} P_{\theta_n}[N < n, S_N \geq b_1(N)] = P_0[N < \infty, S_N \geq b_1(N)].$$

(This result corresponds to relation (41) of [6].) By the right continuity of $P_\theta[N < \infty, S_N \leq -b_2(N)]$ as a function of θ , we obtain that

$$(3.5) \quad \lim_{n \rightarrow \infty} P_{\theta_n}[N < \infty, S_N \leq -b_2(N)] = P_0[N < \infty, S_N \leq -b_2(N)].$$

From (3.4) and (3.5), it follows that

$$(3.6) \quad \begin{aligned} \lim_{n \rightarrow \infty} P_{\theta_n}[N \geq n, S_N \geq b_1(N)] \\ = 1 - P_0[N < \infty, S_N \geq b_1(N)] - P_0[N < \infty, S_N \leq -b_2(N)] \\ = P_0[N = \infty]. \end{aligned}$$

Since $(n + 1 - m)\sigma^2 c_{n+1} = b_1(n + 1) - b_1(m)$, Lemma 1(d) implies that $g_1(\sigma^2 c_{n+1}) \sim n$. Using this and (3.6), we obtain that

$$(3.7) \quad \begin{aligned} E_{\theta_n} N \geq n P_{\theta_n}[N \geq n, S_N \geq b_1(N)] \sim g_1(\sigma^2 c_{n+1}) P_0[N = \infty] \\ \geq g_1(\sigma^2 \theta_n / \rho) P_0[N = \infty] \geq (1 + o(1)) \gamma(\rho) g_1(\sigma^2 \theta_n) P_0[N = \infty]. \end{aligned}$$

Since $\lim_{\theta \rightarrow 0} \mu_\theta / \theta = \sigma^2$, the monotonicity of g_1 and Lemma 1(e) imply that

$$(3.8) \quad g_1(\sigma^2 \theta) \sim g_1(\mu_\theta) \quad \text{as } \theta \rightarrow 0.$$

From (3.7) and (3.8), it then follows as in the proof of Theorem 1 that

$$(3.9) \quad \liminf_{\theta \downarrow 0} E_\theta N / g_1(\mu_\theta) \geq P_0[N = \infty].$$

We note that $E_\theta |X_N| \leq E_\theta (\sum_{i=1}^N X_i^2) \sim \sigma E_\theta \frac{1}{2} N$ as $\theta \downarrow 0$ by Wald's lemma. Since $\lim_{\theta \downarrow 0} (g_1(\theta))^{-\frac{1}{2}} b_1(g_1(\theta)) = \infty$ by (3.1), $\lim_{\theta \downarrow 0} \theta^2 g_1(\theta) = \infty$. Therefore $\lim_{\theta \downarrow 0} \mu_\theta E_\theta \frac{1}{2} N = \infty$ in view of (3.9). Hence $\lim_{\theta \downarrow 0} \mu_\theta E_\theta N = \infty$ and $E_\theta |X_N| = o(\mu_\theta E_\theta N)$. Therefore using Wald's lemma, we obtain that

$$(3.10) \quad \begin{aligned} \mu_\theta E_\theta N = E_\theta S_N \leq E_\theta b_1(N) + E_\theta |X_N| + m E_\theta |X_1| \\ = E_\theta b_1(N) + o(\mu_\theta E_\theta N) \end{aligned}$$

as $\theta \downarrow 0$. For $\theta > 0$, let $\psi^*(\theta)$ be the greatest integer $\leq |\log \theta|$. Since μ_θ is a strictly increasing function of θ , $E_\theta[N | N > \psi^*(\theta)]$ can be written as a function of μ_θ , say $E_\theta[N | N > \psi^*(\theta)] = f(\mu_\theta)$. By Jensen's inequality,

$$(3.11) \quad E_\theta b_1(N) \leq b_1(f(\mu_\theta)) P_\theta[N > \psi^*(\theta)] + b_1(\psi^*(\theta)).$$

Since $b_1(\psi^*(\theta)) = o(\psi^*(\theta))$ and $\mu_\theta g_1(\mu_\theta) = b_1(g_1(\mu_\theta)) \geq b_1(\mu_\theta^{-2}) > \theta^{-1} > \psi^*(\theta)$ for all small positive θ , we obtain that $b_1(\psi^*(\theta)) = o(\mu_\theta E_\theta N)$ as $\theta \downarrow 0$ in view of (3.9). Therefore from (3.10) and (3.11), it follows as in the proof of Theorem 1 that $\liminf_{\theta \downarrow 0} b_1(f(\mu_\theta)) / (\mu_\theta f(\mu_\theta)) \geq 1$. By Lemma 1(b), this implies that

$$(3.12) \quad \limsup_{\theta \downarrow 0} f(\mu_\theta) / g_1(\mu_\theta) \leq 1.$$

Therefore as $\theta \downarrow 0$,

$$(3.13) \quad \begin{aligned} E_\theta N &\leq \phi^*(\theta) + f(\mu_\theta)P_\theta[N > \phi^*(\theta)] \\ &\leq o(g_1(\mu_\theta)) + (1 + o(1))g_1(\mu_\theta)P_\theta[N \geq \phi^*(\theta)]. \end{aligned}$$

In view of (3.9) and (3.13), it remains to prove that

$$(3.14) \quad \lim_{\theta \downarrow 0} P_\theta[N \geq \phi^*(\theta)] = P_0[N = \infty].$$

We note that for $0 < \theta < 1$, $\theta \leq \exp(-\phi^*(\theta)) = o(c_{\phi^*(\theta)})$. Therefore given any $\rho \in (0, 1)$, $\theta < \rho c_{\phi^*(\theta)}$ for all small positive θ . Since for every fixed k , $P_\theta[N < k, S_N \geq b_1(N)]$ is nondecreasing in θ , it then follows that as $\theta \downarrow 0$,

$$(3.15) \quad \begin{aligned} P_{\rho c_{\phi^*(\theta)}}[N < \phi^*(\theta), S_N \geq b_1(N)] &\geq P_\theta[N < \phi^*(\theta), S_N \geq b_1(N)] \\ &\geq P_0[N < \phi^*(\theta), S_N \geq b_1(N)] \\ &\rightarrow P_0[N < \infty, S_N \geq b_1(N)]. \end{aligned}$$

Now $\lim_{\theta \downarrow 0} P_{\rho c_{\phi^*(\theta)}}[N < \phi^*(\theta), S_N \geq b_1(N)] = P_0[N < \infty, S_N \geq b_1(N)]$ by (3.4), and so (3.15) implies that

$$(3.16) \quad \lim_{\theta \downarrow 0} P_\theta[N < \phi^*(\theta), S_N \geq b_1(N)] = P_0[N < \infty, S_N \geq b_1(N)].$$

Given any $k \geq m$, we have for all small positive θ ,

$$(3.17) \quad \begin{aligned} P_0[N < \infty, S_N \leq -b_2(N)] &\geq P_\theta[N < \infty, S_N \leq -b_2(N)] \\ &\geq P_\theta[N < \phi^*(\theta), S_N \leq -b_2(N)] \\ &\geq P_\theta[N \leq k, S_N \leq -b_2(N)] \\ &\rightarrow P_0[N \leq k, S_N \leq -b_2(N)] \quad \text{as } \theta \downarrow 0. \end{aligned}$$

Since (3.17) holds for all $k \geq m$, it follows that

$$(3.18) \quad \lim_{\theta \downarrow 0} P_\theta[N < \phi^*(\theta), S_N \leq -b_2(N)] = P_0[N < \infty, S_N \leq -b_2(N)].$$

From (3.16) and (3.18), (3.14) follows immediately. \square

EXAMPLE 2. Let X_1, X_2, \dots be the same as in Theorem 3 and let $\sigma^2 = E_0 X_1^2$, $b_1(t) = b_2(t) = \sigma\{2t[\log_2(t + e) + c \log_3(t + e^e)]^{\frac{1}{2}}\}$, where $c > \frac{3}{2}$. Choose $m \geq 1$ such that b_1, b_2 are concave on $[m, \infty)$ and $P_0[-b_2(n) < S_n < b_1(n) \text{ for all } n \geq m] > 0$. Then for $i = 1, 2$, $0 < \varepsilon < 1$, $\lim_{t \rightarrow \infty} b_i(\varepsilon t)/b_i(t) = \varepsilon^{\frac{1}{2}}$. Solving $b_1(t) = \theta t$ for t in terms of θ , we obtain that $t = g_1(\theta) \sim \theta^{-2}(2\sigma^2 \log_2 \theta^{-1})$ as $\theta \rightarrow 0+$. Likewise $g_2(\theta) \sim \theta^{-2}(2\sigma^2 \log_2 |\theta|^{-1})$ as $\theta \rightarrow 0-$. Hence defining N as in Theorem 3, we obtain by the theorem that

$$(3.19) \quad \lim_{\theta \rightarrow 0} \mu_\theta^2 E_\theta N / (2\sigma^2 \log |\log |\mu_\theta||) = P_0[N = \infty],$$

which was obtained by Farrell in [6]

Using a similar argument as the proof of Theorem 3, we can prove the following analogue of Theorem 3 for power-one tests of the one-sided hypothesis $H_0: \theta > 0$ based on sample sums and upper-class boundaries.

THEOREM 4. Let $X_1, X_2, \dots, S_n, \Theta, P_\theta, \mu_\theta$ be as in Theorem 3 ($\mu_0 = 0$). Let $b(t)$ be a positive continuous function on $[m, \infty)$ with $m \geq 1$ such that $\lim_{t \rightarrow \infty} t^{-1}b(t) = \infty$ and $P_0[S_n < b(n) \text{ for all } n \geq m] > 0$. Assume that $b(t)$ also satisfies the regularity conditions (2.1) and (2.2) of Theorem 1 and define $g(\theta)$ as in Theorem 1. Let $T = \inf \{n \geq m : S_n \geq b(n)\}$. Then

$$(3.20) \quad \lim_{\theta \downarrow 0} E_\theta T / g(\mu_\theta) = P_0[T = \infty].$$

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