

# ON THE ASYMPTOTIC FORMULA FOR THE PROBABILITY OF A TYPE I ERROR OF MIXTURE TYPE POWER ONE TESTS<sup>1</sup>

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Let  $X_1, X_2, \dots$  be iid with density  $f_y$  with respect to a sigma finite measure  $\mu$ , where  $\{f_y\}_{y \in \Omega}$ ,  $\Omega \subseteq R$ , is an exponential family. Let  $F$  be a probability measure on  $\Omega$  and let  $\theta_0 \in \Omega$ . Define

$$T(B, F) = \min \left\{ n \left| \int_{\Omega} \frac{f_y(X_1) \cdots f_y(X_n)}{f_{\theta_0}(X_1) \cdots f_{\theta_0}(X_n)} dF(y) \geq B \right. \right\},$$

$T(B, F) = \infty$  if no such  $n$  exists. Previous studies have found that if  $F$  has a positive and continuous density with respect to Lebesgue measure on  $\Omega$ , then

$$BP_{\theta_0}(T(B, F) < \infty) \rightarrow_{B \rightarrow \infty} \int_{\Omega} \int_0^{\infty} \exp\{-x\} dH_{\theta}(x) dF(\theta),$$

where  $H_{\theta}$  are certain measures arising in a renewal-theoretic context.

Here we show that in a nonlattice context, this convergence holds for general probability measures  $F$ . We also show that the convergence is uniform for all probability measures  $F$  whose support is contained in an arbitrary interval  $[a, b]$  interior to  $\Omega$ , if the distribution of  $X_1$  is strongly nonlattice for all  $y \in \Omega$ .

**1. Introduction and summary.** Let  $\Omega$  be an open interval on the real line and let  $\{f_y\}_{y \in \Omega}$  be the densities of a one-parameter exponential family with natural state space  $\Omega$  with respect to a sigma-finite measure  $\mu$ . Denote:

$$f_y(x) = \exp\{yx - \psi(y)\}, \quad -\infty < x < \infty, \quad y \in \Omega.$$

Without loss of generality, assume that  $0 = \psi(0) = \psi'(0)$ . Let  $X_1, X_2, \dots$  be a sequence of iid random variables, let  $P_{\theta}$  be the probability measure (on  $R^{\infty}$ ) under which  $X_i$  have density  $f_{\theta}$  with respect to  $\mu$ , and let  $E_{\theta}$  denote expectation under  $P_{\theta}$ . Let  $F$  be a probability measure over  $\Omega$  with  $F(\{0\}) = 0$ . Denote:

$$L(n, y) = \prod_{i=1}^n [f_y(X_i)/f_{\theta_0}(X_i)],$$

$$L(n, F) = \int_{\Omega} L(n, y) dF(y),$$

$$T(B, F) = \min\{n | L(n, F) \geq B\} \\ = \infty \quad \text{if no such } n \text{ exists.}$$

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The statistical test which stops at  $T(B, F)$  and rejects  $H_0: \theta = 0$  in favor of  $H_1: \theta \neq 0$  has power one for certain values of  $\theta$  [cf. Robbins (1970)]. It is known that [Lai and Siegmund (1977) and Woodroffe (1982), Section 6.1] the significance level of this test is

$$(1) \quad P_0(T(B, F) < \infty) = \int_{\Omega} E_{\theta} [1/L(T(B, F), F)] dF(\theta).$$

Let  $S_n^y = y \sum_{i=1}^n X_i - n\psi(y)$ , let  $\tau = \min\{n | S_n^y \geq A\}$ ,  $\tau = \infty$  if no such  $n$  exists and let  $\rho_y = S_{\tau}^y - A$  on  $\{\tau < \infty\}$ . If the  $P_0$ -distribution of  $yX_i - \psi(y)$  is non-lattice, it follows from standard renewal theory [cf. Feller (1971)] that under  $P_y$ ,  $\rho_y$  has (as  $A \rightarrow \infty$ ) a limiting distribution  $H_y$ . If  $F$  has a positive continuous density with respect to Lebesgue measure on  $\Omega$ , then by Lai and Siegmund (1977)

$$(2) \quad BP_0(T(B, F) < \infty) \rightarrow_{B \rightarrow \infty} \int_{\Omega} \int_0^{\infty} \exp\{-x\} dH_{\theta}(x) dF(\theta)$$

[see also Woodroffe (1982), Section 6.2]. The method involved in the proof of (2) is nonlinear renewal theory [developed by Woodroffe (1976) and Lai and Siegmund (1977); see Woodroffe (1982) for a survey]. Formula (2) yields an approximation for the significance level of the test associated with  $T(B, F)$ . The approximation is remarkably good, even for low values of  $B$  [Lai and Siegmund (1977)].

Let  $\delta_{\theta}$  denote the probability measure degenerate at  $\theta$ . For testing  $H_0: \theta = 0$  against an alternative  $H_1: \theta = y$  with a power one test,  $T(B, \delta_y)$  is optimal in the sense that it has least  $P_y$ -expected sample size among all power one tests with significance level  $\alpha \leq P_0(T(B, \delta_y) < \infty)$ . If the alternative  $H_1$  is not simple,  $T(B, \delta_y)$  may not yield a test of power one at every point in the alternative and is not efficient for values of  $\theta$  other than  $y$ . One can maintain power one and asymptotic ( $B \rightarrow \infty$ ) efficiency at every point in the alternative by employing a rule  $T(B, F)$  with  $F$  having a positive continuous density for all points in the alternative [see Pollak and Siegmund (1975) and Pollak (1978)]. The asymptotic efficiency of  $T(B, F)$  is manifest when  $B$  is large. When  $B$  is not large  $T(B^*, \delta_y)$  will have a significantly smaller  $P_y$ -expected sample size than  $T(B, F)$  [where  $B^*$  is such that  $T(B^*, \delta_y)$  and  $T(B, F)$  yield tests with power one at  $\theta = y$  having the same level of significance]. For reasons of continuity,  $E_{\theta}T(B^*, \delta_y)$  will be smaller than  $E_{\theta}T(B, F)$  for a sizeable  $\theta$ -neighborhood of  $y$ . As for practicality, the integration involved in computing  $L(n, F)$  may make application of  $T(B, F)$  cumbersome. Therefore, choosing a measure  $F^*$  concentrated at a single point or having atoms at a few points and employing  $T(B, F^*)$  may be much more appealing than using  $T(B, F)$  with continuous  $F$ . The range of values of  $B$  where this may be the case is large, and seems to include many of the "practical" cases. [For an indication of this, see Pollak and Siegmund (1985).] While the test associated with  $T(B, F^*)$  may not have power one at all points in the alternative, averaging  $F^*$  with a continuous  $F$  will rectify this, at the same time retaining reasonably good efficiency for most points in the alternative. Therefore, there is an interest in establishing (2) for a wider range of measures  $F$ .

Other questions of interest concern the uniformity (in  $F$ ) of the convergence in (2). These become of importance when several measures are being considered or when the measure  $F$  is random. [This may be the case, for instance, if a machine requires calibration at a value  $\theta = 0$  at the start of each day, a power one test is daily employed on the products to check whether  $\theta = 0$ , and the measure  $F$ , which represents the values of  $\theta$  when  $\theta \neq 0$ , is daily updated. For another example of a random  $F$ , arising in a different context—one where the uniformity of the convergence is crucial—see Pollak (1983).]

Here we show that in case  $yX_i - \psi(y)$  are nonlattice, the convergence in (2) exists for general probability measures  $F$ . We show that this convergence is uniform for all probability measures  $F$  whose support is contained in an arbitrary interval  $[a, b]$  interior to  $\Omega$ , under the restriction that the  $P_y$ -distribution of  $X_i$  be strongly nonlattice [see Stone (1965)] for all  $y \in \Omega$ . (This requirement is fulfilled, for example, in case the observations are normal or exponential. It is not fulfilled if they are binomial or Poisson.) It should be noted that the uniformity result may not hold if the strongly nonlattice assumption is not satisfied (e.g., the Bernoulli case).

Even when the strongly nonlattice assumption is satisfied, the uniformity result is not transparent. The standard approach of decomposing a nonlinear renewal process calls for representing  $L(n, F)$  via

$$(3) \quad \log L(n, F) = \theta \sum_{i=1}^n X_i - n\psi(\theta) + \xi(n, \theta, F),$$

where  $\xi(n, \theta, F)$  are sequences which are slowly varying. If these are slowly varying uniformly in  $\theta$  and  $F$ , letting  $T = T(B, F)$ , this representation would be applied to

$$(4) \quad BP_0(T(B, F) < \infty) = \int E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta)$$

[which is equivalent to (1)] to yield a uniform convergence in (2). The difficulty is that the representation (3) may fail, let alone  $\xi(n, \theta, F)$  slowly vary uniformly [e.g., consider the  $N(\theta, 1)$  case with  $F = \frac{1}{2}\delta_{y-\varepsilon} + \frac{1}{2}\delta_{y+\varepsilon}$ —the representation fails for  $\theta = y$ ]. Looking at it differently, to each  $B, F$  there corresponds a (one-sided or two-sided) boundary  $\gamma(t)$  via the relation

$$B = \int_{\Omega} \exp\{y\gamma(t) - t\psi(y)\} dF(y).$$

The stopping time  $T(B, F)$  is equal to the first time  $n$  that the sequence of partial sums  $\sum_{i=1}^n X_i$  crosses the boundary  $\gamma(t)$ , and one can try to get the asymptotics of the overshoot  $\sum_{i=1}^{T(B, F)} X_i - \gamma(t)$  to account for uniform convergence in (2) via smoothness properties of  $\gamma(t)$ . However, it is generally not the case that for “neighboring” mixing measures  $F_1, F_2$  the corresponding boundaries  $\gamma_1(t), \gamma_2(t)$  are close uniformly in  $B$ . [For instance, consider  $F_1 = F_a, F_2 = (1 - \varepsilon)\delta_a + \varepsilon\delta_b$  where  $0 < a < b$ . An easy calculation shows that  $\gamma_1(0) = (\log B)/a$ , while  $\gamma_2(0) = (\log B - \log \varepsilon)/b + o(1)$ .]

Nevertheless, due to the following reasoning, the uniformity result is true. By virtue of (4) it is enough to show that the representation (3) and the uniformity

in  $\theta$  and  $F$  of the slowly varying characteristics of  $\zeta(n, \theta, F)$  hold not for all  $\theta$ , but for a  $\theta$ -set  $\Delta_B(F)$  having arbitrarily large  $F$ -probability—larger, say, than  $1 - \epsilon$ ,  $\epsilon$  arbitrary. It does not matter if  $\Delta_B(F)$  varies with  $B$  or  $F$ , as long as  $1 - \epsilon$  remains a lower bound for its probability and the slowly varying characteristics of  $\xi(n, \theta, F)$  continue to hold for  $n$  in the vicinity of  $T(B, F)$ . The rigorous presentation of this reasoning is the content of the proof supplied in this article.

**2. General convergence.** We will use the notation of the previous section.

**THEOREM 1.** *Suppose that  $F\{0\} = 0$  and  $F\{\{\theta|\theta X_1 - \psi(\theta) \text{ has a lattice } P_0\text{-distribution}\}\} = 0$ . Then*

$$BP_0\{T(B, F) < \infty\} \rightarrow_{B \rightarrow \infty} \int_{\Omega} \int_0^{\infty} \exp\{-x\} dH_{\theta}(x) dF(\theta).$$

Essentially, the idea of the proof is to apply Lemmas 1 and 5 below and the nonlinear renewal theorem to (4) above.

Let  $a \leq b$  be interior points of  $\Omega$ . Fix  $\delta > 0$ ,  $\beta = \frac{1}{20}$ ,  $\rho = \frac{1}{2}$ , and  $\alpha = \frac{3}{4}$ . Denote

$$T = T(B, F), \quad \bar{X}_n = \sum_{i=1}^n X_i/n, \quad \epsilon = \epsilon(n) = n^{\beta-1/2},$$

$$\kappa = \kappa(n, \theta) = \min\{\epsilon, \frac{1}{2}|\theta|, |\theta|/\sqrt{\psi''(\theta)}\}, \quad \sigma = \sigma(n) = n^{2\beta-1/2},$$

$S[a, b]$  = set of all probability measures  $F$  whose support is contained in  $[a, b]$ , and which satisfy  $F\{0\} = 0$  and  $F\{\{\theta|\theta X_1 - \psi(\theta) \text{ has a lattice distribution}\}\} = 0$ ,

$$I(\theta) = \theta\psi'(\theta) - \psi(\theta), \quad l_{\theta} = (\log B)/I(\theta),$$

$$m_{\theta} = \text{the integer value of } l_{\theta} - (l_{\theta})^{\alpha\rho}/4,$$

$$n_{\theta} = \text{the integer value of } l_{\theta} + (l_{\theta})^{\alpha\rho}/4,$$

$$\epsilon_{\theta} = (n_{\theta})^{\beta-1/2}, \quad \epsilon_{\theta}^* = \min\{\epsilon_{\theta}, \frac{1}{2}|\theta|, |\theta|/\sqrt{\psi''(\theta)}\}, \quad \sigma_{\theta} = (m_{\theta})^{2\beta-1/2},$$

$$\Delta_B = \Delta_B(F) = \{\theta|\theta \neq 0, F\{[\theta - \frac{1}{2}\kappa, \theta + \frac{1}{2}\kappa]\} \geq \delta\kappa \text{ for all } m_{\theta} \leq n \leq n_{\theta}\},$$

$$R_B = R_B(F) = \Delta_B \cap \{\theta|\ |\theta| > (\log B)^{-1/10}\},$$

$$\theta^* = \text{is defined by } \psi'(\theta^*) = \bar{X}_n,$$

$$\theta^{\#} = \theta^{\#}(\theta, B) = |\theta| + [(1 - \rho)l_{\theta}]^{2\beta-1/2},$$

$$n^* = n^*(\theta, B) = \min\{|\theta|^{-2/(1-3\beta)}, 7l_{\theta}\}.$$

The following lemmas are stated in somewhat greater generality than needed for Theorem 1 so as to enable their use for the proof of Theorem 2.

**LEMMA 1.** *There exists  $0 < B_{\delta} < \infty$  such that if  $B \geq B_{\delta}$  then*

$$F\{\Delta_B^{\text{complement}}\} \leq 5(b - a) \delta \quad \text{whenever } F \in S[a, b].$$

**PROOF.** Suppose  $a < 0 < b$ . Suppose  $0 \neq \theta \in [a, b]$  and  $F\{[\theta - \frac{1}{2}\kappa, \theta + \frac{1}{2}\kappa]\} < \delta\kappa$  for some  $m_{\theta} \leq n \leq n_{\theta}$ . There exists  $0 < B_{\delta} < \infty$  (independent of

$\theta, F)$  such that if  $B \geq B_\delta$  then, for such  $\theta$ ,  $F\{[\theta - \frac{1}{2}\epsilon_\theta^*, \theta + \frac{1}{2}\epsilon_\theta^*]\} \leq 2\delta\epsilon_\theta^*$  whenever  $F \in S[a, b]$ . Therefore

$$(\Delta_B)^{\text{complement}} \subseteq \{\theta | F\{(\theta - \frac{1}{2}\epsilon_\theta^*, \theta + \frac{1}{2}\epsilon_\theta^*)\} \leq 2\delta\epsilon_\theta^*\}$$

if  $B \geq B_\delta$ .

Let  $b_0 = b$ , and define recursively  $b_i = \max\{\theta | 0 < \theta \leq b_{i-1} - \frac{1}{2}\epsilon_{b_{i-1}}^*, F\{(\theta - \frac{1}{2}\epsilon_\theta^*, \theta + \frac{1}{2}\epsilon_\theta^*)\} \leq 2\delta\epsilon_\theta^*\}$ ,  $i = 1, 2, \dots$ , and define

$$D_i = \{\theta | b_i - \frac{1}{2}\epsilon_{b_i}^* < \theta < b_i + \frac{1}{2}\epsilon_{b_i}^*\}.$$

Thus,  $(0, b] \cap (\Delta_B)^{\text{complement}} \subseteq \cup_i D_i$ . Clearly,  $F\{D_i\} \leq 2\delta\epsilon_{b_i}^* = 2\delta|D_i|$ . Also  $D_i \cap D_j = \phi$  if  $|i - j| > 1$ , so  $F\{\cup_i D_i\} \leq F\{\cup_j D_{2j}\} + F\{\cup_j D_{2j+1}\} \leq 2\delta b + 2\delta\epsilon_b^* + 2\delta b \leq 5\delta b$ . Consequently,  $F\{(0, b] \cap (\Delta_B)^{\text{complement}}\} \leq 5\delta b$ . A similar argument holds for  $[a, 0)$ , so that  $F\{(\Delta_B)^{\text{complement}}\} \leq 5\delta(b - a)$ . The argument for  $0 \leq a \leq b$  and for  $a \leq b \leq 0$  is analogous.  $\square$

**LEMMA 2.** Let  $0 < \eta < \infty$  and denote

$$\zeta(n, \theta, F) = \log \left[ L(n, F) / \int_{\theta - (1/2)\sigma_\theta}^{\theta + (1/2)\sigma_\theta} L(n, y) dF(y) \right].$$

Then

$$\sup_{F \in S[a, b]} \sup_{\theta \in \Delta_B} \max_{n \geq (1-\rho)l_B} P_\theta \left\{ \max_{j \geq 1} |\zeta(n + j, \theta, F) - \zeta(n, \theta, F)| > \eta \right\} \rightarrow_{B \rightarrow \infty} 0.$$

**PROOF.** Let  $W_n^\theta$  denote the event  $\{|\theta^* - \theta| < \epsilon\}$ . There exists a constant  $c > 0$  independent of  $n, \theta, F$  such that if  $a \leq \theta \leq b$ , then

$$(5) \quad \{|\bar{X}_n - \psi'(\theta)| < c\epsilon\} \subset W_n^\theta.$$

Following the proof of Lemma 2 of Pollak and Siegmund (1975), for  $\Lambda > \theta$

$$\begin{aligned} & P_\theta \{ \bar{X}_n - \psi'(\theta) > z \} \\ (6) \quad &= \int_{\{ \bar{X}_n - \psi'(\theta) > z \}} \exp\{n[(\theta - \Lambda)\bar{X}_n - (\psi(\theta) - \psi(\Lambda))]\} dP_\Lambda \\ &\leq \exp\{-n[(\psi'(\theta) + z)(\Lambda - \theta) - (\psi(\Lambda) - \psi(\theta))]\} \\ &\quad \times P_\Lambda \{ \bar{X}_n - \psi'(\theta) > z \}. \end{aligned}$$

Setting  $z = c\epsilon$  and  $\Lambda = \theta + n^{(1/2)\beta - 1/2}$  yields for large enough  $n$

$$(7) \quad P_\theta \{ \bar{X}_n - \psi'(\theta) > c\epsilon \} < \exp\{-\frac{1}{2}cn^{3\beta/2}\}$$

uniformly for  $a \leq \theta \leq b$ . Hence

$$\sup_{a \leq \theta \leq b} \sum_{n=r}^\infty P_\theta \{ \bar{X}_n - \psi'(\theta) > c\epsilon \} \rightarrow_{r \rightarrow \infty} 0.$$

A similar analysis yields

$$\sup_{a \leq \theta \leq b} \sum_{n=r}^{\infty} P_{\theta} \{ \bar{X}_n - \psi'(\theta) < -c\varepsilon \} \rightarrow_{r \rightarrow \infty} 0.$$

Hence, by (5)

$$(8) \quad \inf_{a \leq \theta \leq b} P_{\theta} \left\{ \bigcap_{n=r}^{\infty} W_n^{\theta} \right\} \rightarrow_{r \rightarrow \infty} 1.$$

Note that

$$\begin{aligned} & \log L(n, \theta + \Delta\theta) \\ &= n [ (\theta + \Delta\theta) \bar{X}_n - \psi(\theta + \Delta\theta) ] \\ &= n \{ \theta \bar{X}_n - \psi(\theta) + \Delta\theta [\bar{x}_n - \psi'(\theta)] - \frac{1}{2} \psi''(\theta) (\Delta\theta)^2 [1 + o(1)] \}, \end{aligned}$$

where  $o(1) \rightarrow 0$  as  $\Delta\theta \rightarrow 0$  uniformly in  $\theta \in [a - \frac{1}{2}\varepsilon, b + \frac{1}{2}\varepsilon]$  if  $a - \frac{1}{2}\varepsilon, b + \frac{1}{2}\varepsilon \in \Omega$ . For large enough  $n$ , for  $\theta = \theta^*$ , this becomes

$$\log L(n, \theta^* + \Delta\theta) = n \{ \theta^* \bar{X}_n - \psi(\theta^*) - \frac{1}{2} \psi''(\theta^*) (\Delta\theta)^2 [1 + o(1)] \}.$$

Therefore, there exists  $B_0$  (independent of  $F$ ) such that if  $B > B_0$  and  $\theta \in \Delta_B$ , on  $W_n^{\theta}$  for  $n \geq (1 - \rho)l_{\theta}$ , if  $F \in S[a, b]$

$$\begin{aligned} & \frac{\int_a^b L(n, y) dF(y)}{\int_{\theta - (1/2)\sigma_{\theta}}^{\theta + (1/2)\sigma_{\theta}} L(n, y) dF(y)} \\ & \leq 1 + \frac{(\int_{\theta + (1/2)\sigma_{\theta}}^b + \int_a^{\theta - (1/2)\sigma_{\theta}}) L(n, y) dF(y)}{\int_{\theta^* - \varepsilon_{\theta}}^{\theta^* + \varepsilon_{\theta}} L(n, y) dF(y)} \\ & \leq 1 + \frac{L(n, \theta^* + \sigma_{\theta}/4) + L(n, \theta^* - \sigma_{\theta}/4)}{\delta \varepsilon_{\theta}^* \min\{L(n, \theta^* - \varepsilon_{\theta}), L(n, \theta^* + \varepsilon_{\theta})\}} \\ & \leq 1 + \frac{2 \exp\{n [\theta^* \bar{X}_n - \psi(\theta^*) - \frac{1}{2} \psi''(\theta^*) (\sigma_{\theta}/4)^2 [1 + o(1)]]\}}{\delta \varepsilon_{\theta}^* \exp\{n [\theta^* \bar{X}_n - \psi(\theta^*) - \frac{1}{2} \psi''(\theta^*) (\varepsilon_{\theta})^2 [1 + o(1)]]\}} \\ & \leq 1 + \frac{2 \exp\{-\frac{1}{64} [\min_{a \leq \theta \leq b} \psi''(\theta)] \sigma_{\theta}^2 (1 - \rho) l_{\theta}\}}{\delta \varepsilon_{\theta}^*} \\ & \rightarrow_{B \rightarrow \infty} 1. \end{aligned}$$

This, together with (8), accounts for Lemma 2.  $\square$

LEMMA 3. Let  $0 < \eta < \infty$ . Let

$$\begin{aligned} K_n(\theta) &= \frac{1}{2} n \psi''(\theta) \{ [\bar{X}_n - \psi'(\theta)] / \psi''(\theta) \}^2, \\ J_n(y, \theta, \lambda) &= \frac{1}{2} n \psi''(\theta) \{ y - \theta - [\bar{X}_n - \psi'(\theta)] / \psi''(\theta) \}^2 \\ & \quad + n(y - \theta)^3 \psi'''(\lambda) / 6. \end{aligned}$$

Let  $\{\lambda_n(\theta)\}_{n=1}^\infty$  be any random sequence such that  $\theta - \sigma \leq \lambda_n(\theta) \leq \theta + \sigma$ . Then

$$(i) \sup_{a \leq \theta \leq b} P_\theta \left\{ \max_{j=1, \dots, \rho n^\alpha} |K_{n+j}(\theta) - K_n(\theta)| \geq \eta \right\} \rightarrow_{n \rightarrow \infty} 0,$$

$$(ii) \sup_{a \leq \theta \leq b} P_\theta \left\{ \max_{\theta - \sigma \leq y \leq \theta + \sigma} \max_{j=1, \dots, \rho n^\alpha} |J_{n+j}(y, \theta, \lambda_{n+j}(\theta)) - J_n(y, \theta, \lambda_n(\theta))| \geq \eta \right\} \rightarrow_{n \rightarrow \infty} 0.$$

PROOF. Part (i) follows from Proposition 1 of Lai and Siegmund (1979), noting that the proof of this Proposition 1 can be carried through uniformly for  $a \leq \theta \leq b$ .

As for part (ii), note that

$$J_n(y, \theta, \lambda) = n(y - \theta)^3 \psi'''(\lambda)/6 + K_n(\theta) + \frac{1}{2} \psi''(\theta)(y - \theta)^2 n - \sum_{i=1}^n [X_i - \psi'(\theta)](y - \theta).$$

Therefore, for  $\theta - \sigma \leq y \leq \theta + \sigma$ ,  $1 \leq j \leq \rho n^\alpha$  and large enough  $n$

$$(9) \quad \begin{aligned} & |J_{n+j}(y, \theta, \lambda_{n+j}(\theta)) - J_n(y, \theta, \lambda_n(\theta))| \\ & \leq \rho n^\alpha \sigma^3 \sup_{a \leq \theta \leq b} |\psi'''(\theta)| + \sup_{a \leq \theta \leq b} |\psi''''(\theta)| \sigma^4 n \\ & \quad + |K_{n+j}(\theta) - K_n(\theta)| + \psi''(\theta) \rho n^\alpha \sigma^2 + \left| \sum_{i=n+1}^{n+j} [X_i - \psi'(\theta)] \right| \sigma \end{aligned}$$

uniformly for  $a \leq \theta \leq b$ . By Kolmogorov's inequality,

$$\begin{aligned} & P_\theta \left\{ \max_{j=1, \dots, \rho n^\alpha} \left| \sum_{i=n+1}^{n+j} [X_i - \psi'(\theta)] \right| \sigma > \eta/4 \right\} \\ & = P_\theta \left\{ \max_{j=1, \dots, \rho n^\alpha} \left( \sum_{i=n+1}^{n+j} [X_i - \psi'(\theta)] \right)^2 > (\eta/4)^2 / \sigma^2 \right\} \\ & \leq \psi''(\theta) \rho n^\alpha \sigma^2 / (\eta/4)^2 \rightarrow_{n \rightarrow \infty} 0 \end{aligned}$$

uniformly for  $a \leq \theta \leq b$ . Part (ii) now follows from (9) and part (i).  $\square$

LEMMA 4. Using the notation of Lemma 3, denote

$$Q_n(\theta, F) = \log \int_{\theta - (1/2)\sigma_\theta}^{\theta + (1/2)\sigma_\theta} \exp\{-J_n(y, \theta, \lambda_n(\theta))\} dF(y).$$

Let  $0 < \eta < \infty$ . Then

$$\sup_{F \in S[a, b]} \sup_{\theta \in \Delta_B} \max_{m_\theta \leq n \leq n_\theta} P_\theta \left( \max_{j=1, \dots, \rho n^\alpha \wedge (n_\theta - n)} |Q_{n+j}(\theta, F) - Q_n(\theta, F)| > \eta \right) \rightarrow_{B \rightarrow \infty} 0.$$

PROOF.

$$\begin{aligned} & Q_{n+j}(\theta, F) - Q_n(\theta, F) \\ &= \log \left\{ \int_{\theta - (1/2)\sigma_\theta}^{\theta + (1/2)\sigma_\theta} \exp \left\{ - \left[ J_{n+j}(y, \theta, \lambda_{n+j}(\theta)) - J_n(y, \theta, \lambda_n(\theta)) \right] \right\} \right. \\ & \quad \times \exp \left\{ - J_n(y, \theta, \lambda_n(\theta)) \right\} dF(y) / \\ & \quad \left. \int_{\theta - (1/2)\sigma_\theta}^{\theta + (1/2)\sigma_\theta} \exp \left\{ - J_n(y, \theta, \lambda_n(\theta)) \right\} dF(y) \right\}. \end{aligned}$$

This is the logarithm of an expectation of  $\exp \{ - [J_{n+j}(y, \theta, \lambda_{n+j}(\theta)) - J_n(y, \theta, \lambda_n(\theta))] \}$ . Lemma 4 is therefore a consequence of Lemma 3(ii).  $\square$

LEMMA 5.

$$\log L(n, F) = \theta \sum_{i=1}^n X_i - n\psi(\theta) + \xi(n, \theta, F),$$

where  $\{\xi(n, \theta, F)\}_{n=1}^\infty$  is a sequence of random variables which satisfies for any  $0 < \eta < \infty$

$$\sup_{F \in S[a, b]} \sup_{\theta \in \Delta_B} \max_{m_\theta \leq n \leq n_\theta} P_\theta \left( \max_{j=1, \dots, \rho n^\alpha \wedge (n_\theta - n)} |\xi(n+j, \theta, F) - \xi(n, \theta, F)| > \eta \right) \rightarrow_{B \rightarrow \infty} 0.$$

PROOF. Using the notation of Lemma 3 and Lemma 4, for  $\theta \in \Delta_B$  and  $m_\theta \leq n \leq n_\theta$  there exists  $\lambda_n(\theta) \in (\theta - \sigma, \theta + \sigma)$  such that

$$\begin{aligned} & \log \int_{\theta - (1/2)\sigma_\theta}^{\theta + (1/2)\sigma_\theta} L(n, y) dF(y) \\ &= \log \int_{\theta - (1/2)\sigma_\theta}^{\theta + (1/2)\sigma_\theta} \exp \{ n [y\bar{X}_n - \psi(y)] \} dF(y) \\ &= \log \int_{\theta - (1/2)\sigma_\theta}^{\theta + (1/2)\sigma_\theta} \exp \left\{ n \left[ \theta \bar{X}_n - \psi(\theta) + (y - \theta)(\bar{X}_n - \psi'(\theta)) \right. \right. \\ & \quad \left. \left. - \frac{1}{2}(y - \theta)^2 \psi''(\theta) - (y - \theta)^3 \psi'''(\lambda_n(\theta)) / 6 \right] \right\} dF(y) \\ &= \theta \sum_{i=1}^n X_i - n\psi(\theta) + K_n(\theta) + Q_n(\theta, F). \end{aligned}$$

Lemma 5 now follows from Lemma 2, Lemma 3(i) and Lemma 4.  $\square$

The following two lemmas are needed in order to apply the nonlinear renewal theorem.



LEMMA 6.

$$\sup_{F \in S[\alpha, b]} \sup_{\theta \in R_B} P_\theta\{T(B, F) > n_\theta\} \rightarrow_{B \rightarrow \infty} 0.$$

PROOF. Denote  $\omega_n = \bar{X}_n - \psi'(\theta)$ .

$$\begin{aligned} L(n, F) &= \int \exp\{n[y\bar{X}_n - \psi(y)]\} dF(y) \\ &\geq \int_{\theta - (1/2)\epsilon}^{\theta + (1/2)\epsilon} \exp\{n[y(\psi'(\theta) + \omega_n) - \psi(y)]\} dF(y) \\ &= \int_{\theta - (1/2)\epsilon}^{\theta + (1/2)\epsilon} \exp\left\{n\left[\theta(\psi'(\theta) + \omega_n) + (y - \theta)(\psi'(\theta) + \omega_n)\right.\right. \\ &\quad \left.\left. - \left[\psi(\theta) + (y - \theta)\psi'(\theta) + \frac{1}{2}(y - \theta)^2\psi''(\bar{\theta})\right]\right\} dF(y), \end{aligned}$$

where  $|\bar{\theta} - \theta| < \frac{1}{2}\epsilon$ ; so for  $\theta \in \Delta_B$ ,  $n \geq l_\theta$  and large enough  $B$ ,

$$(10) \quad L(n, F) \geq \exp\{nI(\theta)\} \exp\{-n|\omega_n|(|\theta| + \frac{1}{2}\epsilon)\} \exp\{-\psi''(\theta)n^{2\beta}\} \delta\kappa.$$

Denote  $z = \{I(\theta)\rho/[8(|\theta| + \frac{1}{2}\epsilon)]\}(l_\theta)^{-(1-\alpha)}$ . By (6) above, setting  $\Lambda = \theta + z/\psi''(\theta)$ ,

$$P_\theta\{\omega_{n_\theta} > z\} \leq \exp\{-n_\theta[(\psi'(\theta) + z)(\Lambda - \theta) - (\psi(\Lambda) - \psi(\theta))]\} \rightarrow_{B \rightarrow \infty} 0$$

uniformly in  $\theta \in R_B$ , and similarly  $P_\theta\{\omega_{n_\theta} < -z\} \rightarrow_{B \rightarrow \infty} 0$ .

Therefore, inserting  $n_\theta$  instead of  $n$  in (10), it follows that

$$\inf_{F \in S[\alpha, b]} \inf_{\theta \in R_B} P_\theta\{L(n_\theta, F) \geq B\} \rightarrow_{B \rightarrow \infty} 1,$$

which is equivalent to the statement of Lemma 6.  $\square$

LEMMA 7.

$$\sup_{F \in S[\alpha, b]} \sup_{\theta \in R_B} P_\theta\{T(B, F) < m_\theta\} \rightarrow_{B \rightarrow \infty} 0.$$

PROOF. Let  $T = T(B, F)$  and denote  $S_n = n\bar{X}_n$ . Following the proof of Lemma 3 of Pollak and Siegmund (1975), for any  $y > 0$ ,

$$\begin{aligned} (11) \quad P_\theta\{T \leq m_\theta\} &\leq P_\theta\{S_{m_\theta} - \psi'(\theta)m_\theta \geq y[\psi''(\theta)m_\theta]^{1/2}\} \\ &\quad + \int_{\{T \leq m_\theta, S_{m_\theta} - \psi'(\theta)m_\theta \leq y[\psi''(\theta)m_\theta]^{1/2}\}} \exp\{\theta S_{m_\theta} - m_\theta\psi(\theta)\} dP_0 \\ &\leq P_\theta\{S_{m_\theta} - \psi'(\theta)m_\theta \geq y[\psi''(\theta)m_\theta]^{1/2}\} \\ &\quad + \exp\{I(\theta)m_\theta + \theta y[\psi''(\theta)m_\theta]^{1/2}\} P_0\{T \leq m_\theta\}. \end{aligned}$$

Since  $P_0\{T \leq m_\theta\} < P_0\{T < \infty\} \leq 1/B$  [cf. Robbins (1970)], for  $y = (m_\theta)^{1/9}/\sqrt{\psi''(\theta)}$ , the second term on the far right side of (11) is less than  $\exp\{-(\log B)^{1/2}\}$  when  $B$  is large enough, for all  $\theta \in R_B$ . Also for large enough

$B$ , for the same value of  $y$ , for all  $\theta \in R_B$

$$\begin{aligned} P_\theta\{S_{m_\theta} - \psi'(\theta)m_\theta \geq y[\psi''(\theta)m_\theta]^{1/2}\} \\ \leq P_\theta\{\bar{X}_{m_\theta} - \psi'(\theta) \geq (m_\theta)^{(1/9)-1/2}\} \\ \leq \exp\{-\frac{1}{2}(m_\theta)^{1/6}\} \rightarrow_{B \rightarrow \infty} 0, \end{aligned}$$

where the last inequality follows from (7) above and the convergence is uniform in  $\theta \in [a, b] - \{0\}$ . This completes the proof of Lemma 7.  $\square$

**PROOF OF THEOREM 1.** Let  $a_1 < b_1 < 0 < a_2 < b_2$  be interior points of  $\Omega$  and denote  $\Gamma = [a_1, b_1] \cup [a_2, b_2]$ . Suppose first that the support of  $F$  is contained in  $\Gamma$ . By virtue of (4) and the definition of  $T$

$$\begin{aligned} (12) \quad \int_{\Delta_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) &\leq BP_0\{T < \infty\} \\ &\leq \int_{\Delta_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) + F\{\Delta_B^{\text{complement}}\}. \end{aligned}$$

Letting  $\Xi(A)$  denote the indicator function of the set  $A$ , define

$$\phi(\theta) = \left| E_\theta \exp\{-[\log L(T, F) - \log B]\} - \int_0^\infty \exp\{-x\} dH_\theta(x) \right| \Xi(\theta \in \Delta_B).$$

Lemmas 5, 6, and 7 ensure that the considerations of the proof of Theorem 1 of Lai and Siegmund (1977) carry through for  $\theta \in \Delta_B$ , so that  $\phi(\theta) \rightarrow_{B \rightarrow \infty} 0$ . Hence  $\int_\Gamma \phi(\theta) dF(\theta) \rightarrow_{B \rightarrow \infty} 0$ , which implies

$$\begin{aligned} (13) \quad \left| \int_{\Delta_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) \right. \\ \left. - \int_{\Delta_B} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) \right| \rightarrow_{B \rightarrow \infty} 0. \end{aligned}$$

Clearly

$$\begin{aligned} (14) \quad \left| \int_\Omega \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) - \int_{\Delta_B} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) \right| \\ \leq F\{\Delta_B^{\text{complement}}\}. \end{aligned}$$

Since  $\delta$  is arbitrary, (12), (13), and (14) in conjunction with Lemma 1 account for Theorem 1 for  $F$  whose support is contained in  $\Gamma$ .

For general  $F$ , suppose first that  $-\infty = \inf\{x|x \in \Omega\}$  and  $\sup\{x|x \in \Omega\} = \infty$ . Let  $\gamma > 0$  and choose  $-\infty < a_1 < b_1 < 0 < a_2 < b_2 < \infty$  such that  $F(\Gamma) \geq (1 - \gamma)$  where  $\Gamma = [a_1, b_1] \cup [a_2, b_2]$ . Let  $B^\# = B/F\{\Gamma\}$  and let  $dF^\#(x) = dF(x)/F\{\Gamma\}$  for  $x \in \Gamma$ ;  $dF^\#(x) = 0$  otherwise. Clearly  $\{T(B, F) < \infty\} \supseteq$

$\{T(B^\#, F^\#) < \infty\}$ . Since  $F^\# \in S[a_1, b_2]$

$$\begin{aligned} \liminf_{B \rightarrow \infty} BP_0\{T(B, F) < \infty\} &\geq \liminf_{B \rightarrow \infty} BP_0\{T(B^\#, F^\#) < \infty\} \\ &\geq \int_{\Gamma} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta)(1 - \gamma). \end{aligned}$$

On the other hand, from (4) it follows that

$$BP_0\{T(B, F) < \infty\} \leq \int_{\Gamma} \exp\{-[\log L(T, F) - \log B]\} dF(\theta) + \gamma$$

and so in a manner similar to the proof for  $F \in S[a, b]$  it follows that

$$\limsup_{B \rightarrow \infty} BP_0\{T(B, F) < \infty\} \leq \int_{\Gamma} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) + \gamma.$$

Decreasing  $\gamma$  towards zero—i.e., increasing  $b_1$  to zero and  $b_2$  to  $\infty$ , and decreasing  $a_1$  to  $-\infty$  and  $a_2$  to 0—concludes the proof.

If  $\inf\{x|x \in \Omega\} > -\infty$  or  $\sup\{x|x \in \Omega\} < \infty$ , a similar proof is valid. The details are omitted.  $\square$

**3. Uniform convergence.** We will continue to use the notation of the previous sections. The distribution of  $X$  is said to be strongly nonlattice [Stone (1965)] if  $\liminf_{|t| \rightarrow \infty} |E \exp\{itX\} - 1| > 0$ .

**THEOREM 2.** *Suppose that the  $P_\gamma$ -distribution of  $X_1$  is strongly nonlattice for all  $\gamma \in \Omega$ . Let  $a \leq b$  be interior points of  $\Omega$ . Then*

$$BP_0\{T(B, F) < \infty\} \rightarrow_{B \rightarrow \infty} \int_a^b \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta)$$

*uniformly in  $F \in S[a, b]$ .*

The proof breaks down into two parts: for  $\theta \in R_B$ , the proof is similar to that of Theorem 1, with the strongly nonlattice property ensuring uniform convergence. For  $\theta \in \Delta_B - R_B$ , the proof shows that  $\theta X_t - \psi(\theta)$  is stochastically small enough to ensure that any “overshoot” is negligible. The details are spelled out in the following lemmas.

**LEMMA 8.** *If  $X$  is strongly nonlattice then so is  $cX + d$  for any  $c \neq 0$ ,  $-\infty < d < \infty$ .*

**PROOF.** It suffices to show that if  $X$  is not strongly nonlattice, then neither is  $cX + d$ . Without loss of generality, let  $c = 1$ .

Suppose there exists a sequence  $\{t_j\}_{j=1}^\infty$ ,  $|t_j| \rightarrow_{j \rightarrow \infty} \infty$  such that  $Ee^{it_j X} \rightarrow_{j \rightarrow \infty} 1$ . Then  $e^{it_j X} \rightarrow_{j \rightarrow \infty} 1$  in probability. Therefore, for any integer  $k$ ,  $e^{ikt_j X} \rightarrow_{j \rightarrow \infty} 1$  in probability, and this convergence is uniform for any finite set of integers  $k = 1, \dots, m$ .

Let  $\eta > 0$ . If  $j$  is large enough, then  $\max_{k=1, \dots, m} |Ee^{ikt_j X} - 1| < \eta$ . Clearly,  $m$  can be chosen to be large enough so that  $\sup_{j \geq 1} \min_{1 \leq k \leq m} |e^{ikt_j d} - 1| < \eta$ . Thus, for any large enough  $j$ , there exists  $k_j \in \{1, \dots, m\}$  such that

$$|Ee^{ik_j t_j (X+d)} - 1| = |(Ee^{ik_j t_j X} - 1) + (e^{ik_j t_j d} - 1) + (Ee^{ik_j t_j X} - 1)(e^{ik_j t_j d} - 1)| \leq 2\eta + \eta^2.$$

Letting  $\eta \rightarrow 0$ , it follows that  $X + d$  is not strongly nonlattice.  $\square$

LEMMA 9. Let  $Y_1, Y_2, \dots$  be strongly nonlattice iid random variables with  $EY_i \geq 0, P(Y_i = 0) < 1$ . Let

$$V = \min \left\{ n \left| \sum_{i=1}^n Y_i > 0 \right. \right\}, \quad Z = \sum_{i=1}^V Y_i.$$

Then  $Z$  is strongly nonlattice.

PROOF. The lemma is trivial if  $P(Y_i \geq 0) = 1$ . Consider the case that  $P(Y_i \geq 0) < 1$ . One must show that  $\liminf_{|t| \rightarrow \infty} |Ee^{itZ} - 1| > 0$ . Suppose this were not the case, but that there exists a sequence  $\{t_j\}_{j=1}^\infty, |t_j| \rightarrow_{j \rightarrow \infty} \infty$  such that  $Ee^{it_j Z} \rightarrow_{j \rightarrow \infty} 1$ . Then

$$(15) \quad e^{it_j Z} \rightarrow_{j \rightarrow \infty} 1$$

in probability. Since  $Ee^{it_j Z} = EE(e^{it_j Z} | Y_1)$ , it follows that on  $\{Y_1 > 0\}$  (on which  $Z = Y_1$ )

$$(16) \quad P(e^{it_j Y_1} \rightarrow_{j \rightarrow \infty} 1 | Y_1 > 0) = 1.$$

Let  $u^* = \sup\{x | P(Y_1 \leq x) < 1\}$ . Clearly,  $u^* > 0$ . Let  $U_1 = (-u^*, 0]$ . If  $u^* < \infty$ , let  $U_k = (-ku^*, -(k-1)u^*], k = 1, 2, \dots$

Suppose  $Y_1 \in U_k$ . Then  $P(V = k + 1, Y_2 > 0, \dots, Y_{k+1} > 0 | Y_1) > 0$ . On this event

$$(17) \quad e^{it_j Z} = e^{it_j Y_1} \prod_{m=2}^{k+1} e^{it_j Y_m}.$$

From (16) it follows that

$$P \left( \prod_{m=2}^{k+1} e^{it_j Y_m} \rightarrow_{j \rightarrow \infty} 1 | Y_1, V = k + 1, Y_2 > 0, \dots, Y_{k+1} > 0 \right) = 1$$

a.s. on  $Y_1 \in U_k$ . From (15) and (17) it therefore follows that

$$P(e^{it_j Y_1} \rightarrow_{j \rightarrow \infty} 1 | Y_1 \in U_k) = 1.$$

Therefore  $P(e^{it_j Y_1} \rightarrow_{j \rightarrow \infty} 1) = 1$  and so  $Ee^{it_j Y_1} \rightarrow_{j \rightarrow \infty} 1$ , contradicting the assumption that  $Y_1$  is strongly nonlattice.  $\square$

LEMMA 10. Let  $S_n^\theta = \theta \sum_{i=1}^n X_i - n\psi(\theta)$ , let  $\tau = \tau(\theta, A) = \min\{n | S_n^\theta > A\}$  and let  $H_\theta$  be the  $P_\theta$ -limiting distribution of  $S_\tau^\theta - A$  as  $A \rightarrow \infty$ . Then the convergence in  $P_\theta$ -distribution of  $S_\tau^\theta - A$  is uniform in  $\theta \in [a, b] - \{0\}$ .

PROOF. Let

$$Y_i = \theta X_i - \psi(\theta),$$

$$V_0 = 0, \quad V_k = \min \left\{ n \mid n > V_{k-1}, \sum_{i=V_{k-1}+1}^n Y_i > 0 \right\}, \quad k = 1, 2, \dots,$$

and

$$Z_k = \sum_{i=V_{k-1}+1}^{V_k} Y_i, \quad k = 1, 2, \dots$$

By Lemmas 8 and 9, the  $P_\theta$ -distribution of  $Z_k$  is strongly nonlattice.

Let  $G_\theta$  denote the renewal function defined by  $G_\theta(x) = \sum_{n=0}^\infty G_\theta^{(n)}(x)$  where  $G_\theta^{(n)}$  is the distribution of  $\sum_{k=1}^n Z_k$ . By Theorem (ii) of Stone (1965), there exists  $r > 0$  ( $r = r(\theta)$ ) such that

$$(18) \quad G_\theta(x) = x/\mu_1 + \mu_2/(2\mu_1^2) + Y_\theta(x),$$

where  $\mu_1 = E_\theta Z_1$ ,  $\mu_2 = E_\theta Z_1^2$  and

$$(19) \quad Y_\theta(x) \exp\{rx\} \rightarrow_{x \rightarrow \infty} 0.$$

Let  $0 < \gamma$ . A check of Stone's (1965) proof reveals that there exists a constant  $r > 0$  independent of  $\theta \in [a, b] - [-\gamma, \gamma]$  such that the convergence in (19) is uniform in  $\theta \in [a, b] - [-\gamma, \gamma]$ . Now

$$(20) \quad \begin{aligned} P_\theta(S_r^\theta - A > x) &= \sum_{n=1}^\infty P_\theta \left( \sum_{i=1}^{n-1} Z_i < A, \sum_{i=1}^n Z_i > A + x \right) \\ &= \sum_{n=1}^\infty \int_0^A [1 - G_\theta^{(1)}(A + x - s)] G_\theta^{(n-1)}(ds) \\ &= \int_0^A [1 - G_\theta^{(1)}(A + x - s)] G_\theta(ds). \end{aligned}$$

Note that

$$(21) \quad \begin{aligned} \int_0^A [1 - G_\theta^{(1)}(A + x - s)] ds &\rightarrow_{A \rightarrow \infty} \int_0^\infty [1 - G_\theta^{(1)}(x + s)] ds \\ &= \int_x^\infty [1 - G_\theta^{(1)}(s)] ds \end{aligned}$$

uniformly for  $x \geq 0$ ,  $\theta \in [a, b] - [-\gamma, \gamma]$ . Also, letting  $Y_\theta$  be as in (19)

$$(22) \quad \begin{aligned} &\int_0^A [1 - G_\theta^{(1)}(A + x - s)] Y_\theta(ds) \\ &= [1 - G_\theta^{(1)}(A + x - s)] Y_\theta(s) \Big|_0^A + \int_0^A Y_\theta(s) G_\theta^{(1)}(A + x - ds) \\ &= [1 - G_\theta^{(1)}(x)] Y_\theta(A) - [1 - G_\theta^{(1)}(A + x)] Y_\theta(0) \\ &\quad - \int_0^A Y_\theta(A - s) G_\theta^{(1)}(x + ds) \\ &\rightarrow_{A \rightarrow \infty} 0 \end{aligned}$$

uniformly for  $x \geq 0$ ,  $\theta \in [a, b] - [-\gamma, \gamma]$ . Combining (18) and (20) together with

(21) and (22) proves that the convergence of the  $P_\theta$ -distribution of  $S_\tau^\theta - A$  to  $H_\theta$  is uniform in  $\theta \in [a, b] - [-\gamma, \gamma]$ .

Now suppose  $\theta \in [-\gamma, \gamma] - \{0\}$ . By Theorem 3 of Lorden (1970), for any  $A \geq 0$ ,

$$E_\theta(S_\tau^\theta - A)^3 \leq \frac{5}{4}E\{[\theta X_1 - \psi(\theta)]^+\}^4/I(\theta) \leq 2\theta^4 E_\theta X_1^4/I(\theta) \rightarrow_{\theta \rightarrow 0} 0.$$

Hence,  $S_\tau^\theta - A \rightarrow_{\theta \rightarrow 0} 0$  in  $P_\theta$ -probability, uniformly in  $A \geq 0$ , and thus also  $H_\theta \rightarrow_{\theta \rightarrow 0} \delta_{\{0\}}$ . Therefore, if  $\eta > 0$ , there exists  $\gamma > 0$  such that the Lévy distance between  $H_\theta$  and the  $P_\theta$ -distribution of  $S_\tau^\theta - A$  is bounded by  $\eta$ , uniformly for  $A \geq 0$ ,  $\theta \in [-\gamma, \gamma] - \{0\}$ . By the above, there exists  $A_\eta$  such that if  $A \geq A_\eta$ , the Lévy distance between  $H_\theta$  and the  $P_\theta$ -distribution of  $S_\tau^\theta - A$  is bounded by  $\eta$  uniformly for  $\theta \in [a, b] - \{0\}$ . Since  $\eta$  is arbitrary, the proof of Lemma 10 is complete.  $\square$

LEMMA 11.

$$\sup_{F \in S[a, b]} \sup_{\theta \in [a, b] - \{0\}} P_\theta\{T(B, F) \leq (1 - \rho)l_\theta\} \rightarrow_{B \rightarrow \infty} 0.$$

PROOF. The proof is analogous to the proof of Lemma 7, replacing  $m_\theta$  by  $(1 - \rho)l_\theta$  and  $y$  by  $\rho[I(\theta)\log B]^{1/2}/[2|\theta|(\psi''(\theta))^{1/2}]$ . The second term on the right side of (11) is less than  $\exp\{-\frac{1}{2}\rho \log B\}$ . There exists a constant  $c_1 > 0$  such that the first term on the right side of (11) is less than  $P_\theta\{\bar{X}_{(1-\rho)l_\theta} - \psi'(\theta) > c_1|\theta|\}$ . By virtue of (6), there exists a constant  $c_2 > 0$  such that this probability is uniformly less than

$$\exp\{-(1 - \rho)l_\theta c_2 \theta^2\} = \exp\{-c_2(1 - \rho)[\theta^2/I(\theta)]\log B\} \rightarrow_{B \rightarrow \infty} 0$$

uniformly for  $\theta \in [a, b] - \{0\}$ .  $\square$

LEMMA 12. Let

$$n^* = \left(\frac{1}{|\theta|}\right)^{2/(1-3\beta)} \vee (7l_\theta).$$

Then

$$\sup_{F \in S[a, b]} \sup_{\theta \in \Delta_B - R_B} P_\theta\{T(B, F) > n^*\} \rightarrow_{B \rightarrow \infty} 0.$$

PROOF. Let  $\omega_n = \bar{X}_n - \psi'(\theta)$ . For large enough  $B$ , for  $\theta \in \Delta_B - R_B$

$$\begin{aligned} L(n, F) &\geq \int_{\theta - (1/2)\epsilon_\theta^*}^{\theta + (1/2)\epsilon_\theta^*} \exp\left\{y \sum_{i=1}^n X_i - n\psi(y)\right\} dF(y) \\ &= e^{nI(\theta)} \int_{\theta - (1/2)\epsilon_\theta^*}^{\theta + (1/2)\epsilon_\theta^*} \exp\left\{n[(y - \theta)\psi'(\theta) \right. \\ &\quad \left. - [\psi(y) - \psi(\theta)] + y\omega_n]\right\} dF(y) \\ &\geq e^{nI(\theta)} \exp\left(-\frac{\psi''(\theta)}{4}n(\epsilon_\theta^*)^2\right) e^{-2|\theta||\omega_n|n} \delta\epsilon_\theta^* \\ &\geq e^{n(I(\theta) - \theta^2/4)} e^{-2|\theta||\omega_n|n} \delta\epsilon_\theta^*. \end{aligned}$$

Since  $|\theta| \geq (n^*)^{-(1-3\beta)/2}$ , it follows that  $n^*\theta^2 \geq (n^*)^{3\beta}$ , and, since also  $l_\theta \leq n^*/7$ , it follows that  $\varepsilon_\theta^* \geq (n^*)^{\beta-1/2}$ . Therefore, for large enough  $B$  (for  $\theta \in \Delta_B - R_B$ )

$$\begin{aligned} L(n^*, F) &\geq \delta \exp\left(n^*\left[\frac{1}{5}\theta^2 - 2|\theta||\omega_{n^*}| - \left[\frac{1}{2} - \beta\right](\log n^*)/n^*\right]\right) \\ &\geq \delta \exp\left(\theta^2 n^* \left[\frac{1}{5} - 2|\omega_{n^*}|/|\theta| - \left[\frac{1}{2} - \beta\right](\log n^*)/(n^*)^{3\beta}\right]\right). \end{aligned}$$

Following the notation and proof of Lemma 2, if  $|\bar{X}_{n^*} - \psi'(\theta)| < c(n^*)^{\beta-1/2}$ , then

$$2|\omega_{n^*}|/|\theta| \leq 2c(n^*)^{\beta-(1/2)+(1-3\beta)/2} = 2c(n^*)^{-\beta/2}$$

so that (for large enough  $B$ )

$$L(n^*, F) \geq \exp\{\theta^2 n^*/6\} \geq B.$$

Lemma 12 now follows from the fact that by (7)

$$\sup_{\theta \in \Delta_B - R_B} P_\theta\left\{|\bar{X}_{n^*} - \psi'(\theta)| > c(n^*)^{\beta-1/2}\right\} \rightarrow_{n^* \rightarrow \infty} 0. \square$$

**LEMMA 13.** *Let  $n^*$  be as in Lemma 12. Denote  $\theta^\# = |\theta| + [(1 - \rho)l_\theta]^{2\beta-1/2}$ . Let  $x > 0$ . Then*

$$\sup_{F \in S[a, b]} \sup_{\theta \in \Delta_B - R_B} P_\theta\left\{\max_{i=1, \dots, n^*} \theta^\# |X_i| > x\right\} \rightarrow_{B \rightarrow \infty} 0.$$

**PROOF.** Let  $(2\eta) > 0$  be in the interior of  $\Omega$ . For all  $\theta \in \Delta_B - R_B$ , when  $B$  is large enough

$$\begin{aligned} P_\theta\left\{\max_{i=1, \dots, n^*} \theta^\# X_i > x\right\} &= 1 - [1 - P_\theta\{\theta^\# X_1 > x\}]^{n^*} \\ &= 1 - [1 - P_\theta\{\exp\{\eta X_1\} > \exp\{\eta x/\theta^\#\}]^{n^*} \\ &\leq 1 - [1 - E_\theta \exp\{\eta X_1\} / \exp\{\eta x/\theta^\#\}]^{n^*} \\ &\leq 1 - [1 - \exp\{\psi(2\eta)\} / \exp\{\eta x/\theta^\#\}]^{n^*} \\ &\rightarrow_{B \rightarrow \infty} 0. \end{aligned}$$

A similar analysis for  $P_\theta\{\min_{i=1, \dots, n^*} \theta^\# X_i < -x\}$  completes the proof of Lemma 13.  $\square$

**PROOF OF THEOREM 2.** By virtue of (1) and the definition of  $T$

$$\begin{aligned} (23) \quad &\int_{\Delta_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) \leq BP_0\{T < \infty\} \\ &\leq \int_{\Delta_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) + F\{\Delta_B^{\text{complement}}\}. \end{aligned}$$

Lemmas 5, 6, 7, and 10 ensure that the considerations of the proof of Theorem 1 of Lai and Siegmund (1977) carry through uniformly for  $\theta \in R_B$ ,  $F \in S[a, b]$ , so

that

$$(24) \quad \sup_{F \in S[\alpha, b]} \left| \int_{R_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) - \int_{R_B} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) \right| \rightarrow_{B \rightarrow \infty} 0.$$

Let  $\gamma > 0$ . By (24) there exists  $B_1$  such that if  $B > B_1$

$$(25) \quad \sup_{F \in S[\alpha, b]} \left| \int_{R_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) - \int_{R_B} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) \right| < \gamma/4.$$

CASE I. If  $F\{\Delta_B - R_B\} \leq \gamma/8$ , then it clearly follows from (25) that if  $B > B_1$

$$(26) \quad \left| \int_{\Delta_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) - \int_{\Delta_B} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) \right| < \gamma/2.$$

CASE II. If  $F\{\Delta_B - R_B\} > \gamma/8$ , let  $\theta^\#$  be as in Lemma 13. By virtue of Lemma 11 and Lemma 2, if  $\eta > 0$ , then

$$\sup_{F \in S[\alpha, b]} \sup_{\theta \in \Delta_B - R_B} P_\theta \left\{ \left| \log L(T, F) - \log \int_{\theta - (\theta^\# - |\theta|)}^{\theta + (\theta^\# - |\theta|)} L(T, y) dF(y) \right| > \eta \right\} \rightarrow_{B \rightarrow \infty} 0.$$

Clearly

$$\int_{\theta - (\theta^\# - |\theta|)}^{\theta + (\theta^\# - |\theta|)} L(T, y) dF(y) \leq \exp\{\theta^\# |X_T|\} B.$$

Lemmas 12 and 13 therefore imply that  $\log L(T, F) - \log B \rightarrow_{B \rightarrow \infty} 0$  in  $P_\theta$ -probability uniformly for  $\theta \in \Delta_B - R_B$ . Since  $H_\theta \rightarrow_{\theta \rightarrow 0} \delta_{\{0\}}$  (see the proof of Lemma 10), it follows that there exists  $B_2$  such that if  $B > B_2$ , then

$$\left| \int_{\Delta_B - R_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) - \int_{\Delta_B - R_B} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) \right| < \gamma/4,$$



so that by (25) if  $B > \max(B_1, B_2)$ , then

$$(27) \quad \left| \int_{\Delta_B} E_\theta \exp\{-[\log L(T, F) - \log B]\} dF(\theta) - \int_{\Delta_B} \int_0^\infty \exp\{-x\} dH_\theta dF(\theta) \right| < \gamma/2.$$

Clearly

$$(28) \quad \left| \int_{\Delta_B} \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) - \int_a^b \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) \right| \leq F\{\Delta_B^{\text{complement}}\}.$$

By virtue of Lemma 1,  $\delta$  can be chosen so that

$$(29) \quad F\{\Delta_B^{\text{complement}}\} < \gamma/4.$$

Now (23), (25), (26)/(27), (28), and (29) imply that there exists  $B_3$  such that if  $B > B_3$  then

$$\sup_{F \in S[a, b]} \left| BP_0\{T < \infty\} - \int_a^b \int_0^\infty \exp\{-x\} dH_\theta(x) dF(\theta) \right| < \gamma.$$

Since  $\gamma$  is arbitrary, this completes the proof of Theorem 2.  $\square$

**4. Remarks.** The set  $\{\theta | \theta X_1 - \psi(\theta) \text{ has a lattice } P_\theta\text{-distribution}\}$  is at most countable [Woodroffe (1982), Section 6.2], so that the restriction in Theorem 1 is not prohibitive. If  $F$  gives positive probability to a value  $\theta$  for which  $\theta X_1 - \psi(\theta)$  has a lattice  $P_\theta$ -distribution, then it can be shown that

$$\log L(n, F) = \theta \sum_{i=1}^n X_i - n\psi(\theta) + \log F\{\theta\} + o_p(1),$$

where  $o_p(1) \rightarrow_{n \rightarrow \infty} 0$  in probability. Therefore, Theorem 1 will not hold as is; the lattice property is not asymptotically negligible.

The strongly nonlattice property is utilized to attain uniformity in the convergence of the distribution of  $S_\tau^\theta - A$  (Lemma 10). If other means can be found to yield uniformity—such as regarding probability measures  $F$  whose support is a finite set of points—then the requirement of the strongly nonlattice property may be relaxed.

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