

## CONVERGENCE RATES OF LARGE DEVIATION PROBABILITIES IN THE MULTIDIMENSIONAL CASE<sup>1</sup>

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Let  $\{W_n\}_{n=1,2,\dots}$  denote a sequence of  $k$ -dimensional random vectors on a probability space  $(\Omega, \mathcal{A}, P)$ . Using moment-generating function techniques sufficient conditions are given for the existence of limits  $\rho(A) = \lim_{n \rightarrow \infty} [P(W_n \in k_n A)]^{1/k_n}$  for certain subsets  $A \subset R^k$ , where  $\{k_n\}_{n=1,2,\dots}$  is a divergent sequence of positive real numbers. The results are multivariate analogs of well-known large deviation theorems on the real line.

**1. Introduction.** In 1975 G. L. Sievers stated some large deviation results for a sequence  $\{P_n\}_{n=1,2,\dots}$  of probability measures on  $(R^k, B^k)$ , where  $R^k$  is  $k$ -dimensional Euclidean space and  $B^k$  the system of Borel subsets. Assuming that the limit  $e(A) = \lim_{n \rightarrow \infty} n^{-1} \log P_n(A)$ ,  $A \in B^k$ , exists, Sievers developed methods for determining  $e(A)$  from  $e(B)$  for "simpler" sets  $B$  of the form  $B = \{x = (x_1, \dots, x_k)^T \in R^k : x_1 * a_1, \dots, x_k * a_k\}$ , where the  $*$ 's are either  $\geq$  or  $\leq$ , and  $a = (a_1, \dots, a_k)^T$  is a fixed vector in  $R^k$ . The existence of the limits  $e(B)$  was also assumed.

In this paper two theorems are proven which give sufficient conditions for the existence of limits such as  $e(A)$  or  $e(B)$  described above. These conditions correspond to certain properties of the moment-generating functions of the underlying probability measures and are multivariate analogs of well-known assumptions for large deviation theorems on the real line (cf. Sievers (1969), Plachky and Steinebach (1975)).

Theorem 3.1 is concerned with large deviation probabilities for cones in  $R^k$ . Included is the case involving Sievers' simple sets  $B$ . Theorem 3.2 deals with the existence of the limits  $e(A)$  for the complements of certain bounded subsets of  $R^k$ .

Both theorems can be applied directly to weighted sums of independent, identically distributed (i.i.d.) random vectors with finite moment-generating functions. For details see [12], Chapter 2.2, where some  $k$ -dimensional analogs of the large deviation results of Book (1973, 1975) concerning weighted sums of i.i.d. random variables on the real line are derived.

**2. Notations. Certain facts about convex functions.** Let  $a = (a_1, \dots, a_k)^T$ ,  $b = (b_1, \dots, b_k)^T$  be vectors in  $R^k$ . Then we write  $a \leq b$  ( $a \triangleleft b$ ) if  $a_i \leq b_i$

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( $a_i < b_i$ ) for all  $i = 1, \dots, k$ , and the set  $[a, b] = \{x \in R^k : a \leq x \leq b\}$  is referred to as a  $k$ -dimensional closed interval. Corresponding definitions hold for  $(a, b) = \{x \in R^k : a \triangleleft x \triangleleft b\}$  ( $a$   $k$ -dimensional open interval),  $[a, b)$ , and  $(a, b]$  ( $a$   $k$ -dimensional half-open interval). The inner product of  $a, b \in R^k$  is denoted as  $\langle a, b \rangle$ , i.e.,  $\langle a, b \rangle = \sum_{i=1}^k a_i b_i$ . Furthermore,  $|a| = \langle a, a \rangle^{\frac{1}{2}}$ .

The closure (interior, boundary, complement) of  $A \subset R^k$  is denoted as  $\bar{A}$  ( $A^0$ ,  $\partial A$ ,  $\complement A$ ), and, for fixed  $\lambda \in R^1$ ,  $a \in R^k$ , let  $\lambda(a + A) = \{\lambda(a + x) : x \in A\}$ . For a positive real number  $\varepsilon$  the set  $U_\varepsilon(x_0) = \{x \in R^k : |x - x_0| < \varepsilon\}$  is called an  $\varepsilon$ -neighborhood of  $x_0$ . A subset  $C \subset R^k$  is called a cone if it is closed under positive scalar multiplication, i.e.,  $\lambda x \in C$  when  $x \in C$  and  $\lambda > 0$ . The origin itself may or may not be included here. The polar cone  $C^+$  is defined as the set  $C^+ = \{x \in R^k : \langle x, y \rangle \geq 0 \text{ for all } y \in C\}$ . For the properties of  $C^+$  consult, e.g., Karlin (1962), pages 397–406. A subset  $D \subset R^k$  is called convex if  $\lambda x + (1 - \lambda)y \in D$  when  $x, y \in D$ ,  $0 < \lambda < 1$ .

Let  $f$  be a real-valued function on a subset  $D$  of  $R^k$ . Then  $(\partial f / \partial t_i)(t_0)$ ,  $i = 1, \dots, k$ ,  $t = (t_1, \dots, t_k)^T$ ,  $t_0 = (t_{10}, \dots, t_{k0})^T$ , will denote the partial derivatives of  $f$  at  $t_0$ ,  $f'(t_0)$  the gradient  $((\partial f / \partial t_1)(t_0), \dots, (\partial f / \partial t_k)(t_0))^T$ , and  $f''(t_0)$  the Hessian matrix of second partial derivatives  $(\partial^2 f / \partial t_i \partial t_j)(t_0)$ . A real-valued function  $f$  on a convex subset  $D \subset R^k$  is called convex if  $f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$  for all  $x, y \in D$  and  $\lambda$  in the real open interval  $(0, 1)$ . Moreover,  $f$  is called strictly convex if strict inequality holds.

In proving the main results of this paper we shall make frequent use of certain standard facts about convex functions and their conjugates which we collate below.

LEMMA 2.1. Let  $f$  be a convex function on an open convex set  $D_0 \subset R^k$ . Then

$$(2.1) \quad f \text{ is continuous on } D_0;$$

$$(2.2) \quad f \text{ is (Fréchet) differentiable almost everywhere in } D_0,$$

i.e., there is a linear transformation  $T: R^k \rightarrow R$  such that for sufficiently small  $|h|$   $f(t_0 + h) = f(t_0) + T(h) + |h|\varepsilon(t_0, h)$ . Here  $\varepsilon(t_0, h) \in R$  goes to zero as  $|h| \rightarrow 0$ , and  $t_0 \in D \subset D_0$ , where  $D_0 \setminus D$  has Lebesgue measure zero. Moreover, if  $t$  is a real variable,  $D_0 \setminus D$  is a countable set. All partial derivatives  $(\partial f / \partial t_i)(t_0)$ ,  $i = 1, \dots, k$ , exist in  $D$ , and  $T(h) = \langle f'(t_0), h \rangle$ .

$$(2.3) \quad f' \text{ is continuous on } D;$$

$$(2.4) \quad f' \text{ is isotone on } D, \quad \text{i.e.,} \\ \langle f'(t) - f'(s), t - s \rangle \geq 0 \quad \text{for all } t \neq s \in D,$$

and strict inequality holds if  $f$  is strictly convex.

Let  $f$  be a real-valued function having continuous second partial derivatives on an open convex set  $D_0 \subset R^k$ . It holds that

$$(2.5) \quad f \text{ is convex in } D_0$$

iff the Hessian matrix  $f''(t)$  is nonnegative definite for each  $t \in D_0$ . Moreover, if the Hessian matrix is positive definite on  $D_0$ , then  $f$  is strictly convex, and the set  $f'(D_0) = \{f'(t) : t \in D_0\}$  is open.

Let  $\{f_n\}_{n=1,2,\dots}$  be a sequence of (Fréchet) differentiable convex functions on an open convex set  $D_0 \subset R^k$  such that  $\lim_{n \rightarrow \infty} f_n(t) = f(t)$  exists for each  $t \in D_0$ . It is apparent that  $f$  is convex in  $D_0$ . Moreover, the assertion

$$(2.6) \quad \lim_{n \rightarrow \infty} f'_n(t) = f'(t)$$

holds almost everywhere in  $D_0$ .

Let  $f^* : D_0^* \rightarrow R$  denote the conjugate function of  $f$  defined as  $f^*(t^*) = \sup_{t \in D_0} [\langle t, t^* \rangle - f(t)]$  with domain  $D_0^* = \{t^* \in R^k : f^*(t^*) < \infty\}$ . Then

$$(2.7) \quad f^* \text{ is convex on } D_0^* .$$

PROOF. See, e.g., Rockafellar (1970) and Roberts and Varberg (1973), though the proofs require slight modification to establish the results (2.4) and (2.6). To obtain (2.6), for instance, note that for  $t = (t_1, \dots, t_k)^T \in D$  and  $t_2, \dots, t_k$  being fixed the convexity of the functions  $f_n(\cdot, t_2, \dots, t_k)$  yields the inequalities

$$\begin{aligned} \frac{f_n(t_1 - h_1, t_2, \dots, t_k) - f_n(t_1, \dots, t_k)}{-h_1} &\leq \frac{\partial}{\partial t_1} f_n(t_1, \dots, t_k) \\ &\leq \frac{f_n(t_1 + h_1, t_2, \dots, t_k) - f_n(t_1, \dots, t_k)}{h_1} , \end{aligned}$$

where  $h_1 > 0$  can be chosen arbitrarily small. Letting first  $n$  tend to infinity and then  $h_1$  tend to zero, we have by the differentiability of  $f(\cdot, t_2, \dots, t_k)$  at  $t_1$

$$\lim_{n \rightarrow \infty} \frac{\partial}{\partial t_1} f_n(t_1, \dots, t_k) = \frac{\partial}{\partial t_1} f(t_1, \dots, t_k) .$$

Same arguments for the other components, too, imply the assertion (2.6).

**3. Results.** As already mentioned, the first theorem states a certain large deviation property for cones  $C$  in the Euclidean space  $R^k$ .

**THEOREM 3.1.** Let  $\{W_n\}_{n=1,2,\dots}$  be a sequence of  $k$ -dimensional random variables on a probability space  $(\Omega, \mathcal{A}, P)$ , and  $D_0$  an open, convex subset of  $R^k$ . Suppose further that

- (i)  $\varphi_n(t) = \int \exp \langle t, W_n \rangle dP < \infty$  for all  $t \in D_0$ ;
- (ii)  $\lim_{n \rightarrow \infty} k_n^{-1} \psi_n(t) = c(t) \in R$  for all  $t \in D_0$ ; where  $\psi_n(t) = \log \varphi_n(t)$  and  $\{k_n\}_{n=1,2,\dots}$  is a divergent sequence of positive real numbers, remembering here that  $\varphi_n(t)$  is positive;
- (iii)  $c$  is strictly convex in  $D_0$ .

Now let  $D$  be the subset of  $D_0$ , where  $c'$  exists,  $C$  a cone in  $R^k$  with  $C^0 \neq \emptyset$ , and  $\{a_n\}_{n=1,2,\dots} \subset R^k$  a convergent sequence with

$$(iv) \quad \lim_{n \rightarrow \infty} a_n = a \in \{c'(t) : t \in D \cap C^+\}^0 .$$

Then it follows that

$$(3.1) \quad \lim_{n \rightarrow \infty} [P(W_n \in k_n(a_n + C))]^{1/k_n} = \inf_{t \in D_0} \exp[c(t) - \langle t, a \rangle] \equiv \rho(a),$$

and the infimum equals  $\exp[c(h) - \langle h, a \rangle]$ , where  $h$  is the unique solution of  $a = c'(h)$ .

The second theorem deals with another sort of multivariate probabilities of large deviations, which appears, for instance, when studying convergence rates in the multidimensional version of the law of large numbers.

**THEOREM 3.2.** *Let  $\{W_n\}_{n=1,2,\dots}$  be a sequence of  $k$ -dimensional random variables on a probability space  $(\Omega, \mathcal{A}, P)$  with  $EW_n = 0$ ,  $n = 1, 2, \dots$ . Let  $D_0 \subset R^k$  be an open, bounded convex set including the origin. Suppose also that*

- (i')  $\varphi_n(t) = \int \exp \langle t, W_n \rangle dP < \infty$  for all  $t \in D_0$ ;
- (ii')  $\lim_{n \rightarrow \infty} k_n^{-1} \psi_n(t) = c(t) \in R$  for all  $t \in D_0$ , where  $\psi_n(t) = \log \varphi_n(t)$  and  $\{k_n\}_{n=1,2,\dots}$  is a divergent sequence of positive real numbers;
- (iii')  $c$  is strictly convex in  $D_0$ .

We now let  $D$  be the subset of  $D_0$ , where  $c'$  exists, and suppose that  $0 \in D$ . Consider a bounded set  $A \subset R^k$  with  $0 \in A^0$  and

- (iv')  $a \in \{c'(t) : t \in D\}^0$  for all  $a \in \partial A$ ;
- (v')  $U_\varepsilon(a^*) \cap (CA)^0 \neq \emptyset \quad \forall \varepsilon > 0$ , where  $a^*$  is a point in  $\partial A$  such that the conjugate  $c^*$  of  $c$  attains its infimum on  $\partial A$  at  $a^*$  (cf. (2.7)).

It then follows that

$$(3.2) \quad \lim_{n \rightarrow \infty} [P(W_n \notin k_n A)]^{1/k_n} = \sup_{a \in \partial A} \rho(a) = \rho(a^*),$$

where  $\rho(a) = \exp[c(h) - \langle h, a \rangle]$  and  $h$  is the unique solution of  $a = c'(h)$ .

**REMARK 3.1.** Assumption (iv'), the strict convexity of  $c$ , and  $c(0) = \lim_{n \rightarrow \infty} k_n^{-1} \log \varphi_n(0) = 0$ , together yield the inequalities

$$0 < \rho(a) < 1 \quad \text{for all } a \in \partial A.$$

In particular,  $\log \rho(a)$  is well defined on  $\partial A$ , and even on  $B = \{a \in R^k : \rho(a) > 0\}$ . Accordingly, we have

$$\begin{aligned} -\log \rho(a) &= -\log (\inf_{t \in D_0} \exp[c(t) - \langle t, a \rangle]) \\ &= \sup_{t \in D_0} [\langle t, a \rangle - c(t)] = c^*(a). \end{aligned}$$

Thus, the continuity of  $\rho$  on  $B^0$  follows from (2.7) and (2.1). The last equality in (3.2) then holds from the definition of  $a^*$  in (v').

The proofs of both Theorems 3.1 and 3.2 are based on a certain convergence property of the conjugate distributions  $P_{t,n}$ ,  $n = 1, 2, \dots$ , of  $P$ . Here the  $P_{t,n}$  are defined as

$$(3.3) \quad P_{t,n}(A) = \int_A \frac{\exp \langle t, W_n \rangle}{\varphi_n(t)} dP,$$

for  $A \in \mathcal{A}$ , and for a fixed  $t \in D_0$ .

LEMMA 3.1. Consider the function  $c$  in Theorems 3.1 or 3.2. Further let  $(a_1, a_2)$  be an open interval in  $R^k$ , and put  $a = c'(t)$ , where  $t$  is a fixed vector in  $D$ . Supposing that  $a \in (a_1, a_2)$ , it follows from the assumptions of Theorem 3.1 or 3.2 that

$$(3.4) \quad \lim_{n \rightarrow \infty} P_{t,n}(W_n \in k_n(a_1, a_2)) = 1.$$

REMARK 3.2. Conjugate distributions defined according to (3.3) are known to appear in a certain guise in statistical mechanics when microcanonical measures on a space of possible configurations of a thermodynamical system are defined (cf. Lanford (1973), page 44). The results of Theorems 3.1 and 3.2 thereby could be described as deducing the existence of the thermodynamical limit for the microcanonical ensemble from that of the canonical ensemble (cf. Lanford (1973), page 52, Theorem A5.1 and Corollary A5.2, where the opposite implication is deduced, and also Ruelle (1969), Chapter 3).

REMARK 3.3. The assumptions of Theorems 3.1 and 3.2 may be greatly simplified if  $k_n = n$ , and  $W_n$  is the  $n$ th partial sum of a sequence  $\{X_i\}_{i=1,2,\dots}$  of i.i.d. random vectors with finite moment-generating function  $\varphi(t) = \int \exp\langle t, X_1 \rangle dP$  for  $t \in D_0$ . The distribution of the  $X_i$  is not concentrated on a hyperplane of  $R^k$ . It follows that  $\varphi_n(t) = [\varphi(t)]^n$ , and thus  $c$  equals the cumulant-generating function  $\log \varphi$  of the  $X_i$ . From the properties of  $\log \varphi$  we know that partial derivatives of any order exist in  $D_0$ . In particular, the Hessian matrix  $[\log \varphi(t)]''$  exists in  $D_0$ , and is positive definite since  $[\log \varphi(t)]''$  is the dispersion matrix of  $X_1$  with respect to the conjugate distribution  $P_t$  defined as  $P_t(A) = \int_A (\exp\langle t, X_1 \rangle / \varphi(t)) dP$ ,  $A \in \mathcal{A}$ ,  $t \in D_0$ . Hence, by (2.5),  $\log \varphi$  is strictly convex in  $D_0$ , and  $\{(\log \varphi(t))' : t \in D_0\}$  is a nonempty open set, if  $C$  has a nonempty interior. Assumption (iv) of Theorem 3.1 can thus be fulfilled. Furthermore, relation (v') in Theorem 3.2 is valid if  $A$  is either a closed or convex set.

4. Proofs.

PROOF OF LEMMA 3.1. Put  $W_n = (W_{1n}, \dots, W_{kn})^T$ ,  $a_j = (a_{1j}, \dots, a_{kj})^T$ ,  $j = 1, 2$ ,  $a = (a_1, \dots, a_k)^T$ . Then,

$$(4.1) \quad P_{t,n}(W_n \notin k_n(a_1, a_2)) \leq \sum_{i=1}^k P_{t,n}(W_{in} \leq k_n a_{i1}) + \sum_{i=1}^k P_{t,n}(W_{in} \geq k_n a_{i2}).$$

Now, for negative  $\tau_{i1}$ , it follows that

$$(4.2) \quad P_{t,n}(W_{in} \leq k_n a_{i1}) \leq \int \frac{\exp\{\langle t, W_n \rangle + \tau_{i1}(W_{in} - k_n a_{i1})\}}{\varphi_n(t)} dP \\ = \frac{\varphi_n(t + \tau_{i1} e_i) \exp(-k_n \tau_{i1} a_{i1})}{\varphi_n(t)},$$

where  $e_i$  denotes the  $i$ th column of a  $k \times k$  identity matrix. Remembering that the cumulant-generating functions  $\phi_n = \log \varphi_n$  are convex and differentiable in  $D_0$ , we have

$$(4.3) \quad \log \varphi_n(t) \geq \log \varphi_n(t + \tau_{i1} e_i) - \langle \phi_n'(t + \tau_{i1} e_i), \tau_{i1} e_i \rangle.$$

Relations (4.2) and (4.3) yield the upper bound

$$(4.4) \quad P_{t,n}(W_{in} \leq k_n a_{i1}) \leq \exp \left\{ k_n \tau_{i1} \left[ k_n^{-1} \frac{\partial \psi_n}{\partial t_i} (t + \tau_{i1} e_i) - a_{i1} \right] \right\}.$$

For fixed  $t_j, j \neq i$ , where  $t = (t_1, \dots, t_k)^T$ , it holds that  $\psi_n$  and  $c$  are strictly convex functions of  $t_i, i = 1, \dots, k$ . Using (2.2), (2.3), and (2.4) and recalling that  $a = c'(t) \in (a_1, a_2)$ , we can choose  $\tau_{i1}$  such that  $(\partial c / \partial t_i)(t + \tau_{i1} e_i)$  both exists and lies inside the open interval  $(a_{i1}, a_{i2})$ . In particular,

$$(4.5) \quad \frac{\partial c}{\partial t_i} (t + \tau_{i1} e_i) > a_{i1}, \quad i = 1, \dots, k.$$

Finally, we may note that in view of (2.6)

$$(4.6) \quad \lim_{n \rightarrow \infty} k_n^{-1} \frac{\partial \psi_n}{\partial t_i} (t + \tau_{i1} e_i) = \frac{\partial c}{\partial t_i} (t + \tau_{i1} e_i), \quad i = 1, \dots, k.$$

Using (4.4), (4.5), and (4.6) we thus have

$$(4.7) \quad \lim_{n \rightarrow \infty} P_{t,n}(W_{in} \leq k_n a_{i1}) = 0, \quad i = 1, \dots, k.$$

By similar arguments one also obtains

$$(4.8) \quad \lim_{n \rightarrow \infty} P_{t,n}(W_{in} \geq k_n a_{i2}) = 0, \quad i = 1, \dots, k.$$

Use of (4.1) renders the proof complete.

Large deviation results on the real line analogous to (3.1) or (3.2) are usually proven by deriving upper and lower bounds for the probabilities to be studied, and then showing that both bounds are equal. The upper bound is thereby often easy to obtain whereas the hard work is in the lower bound (see, e.g., Bahadur (1971), page 5, and Landford (1973), page 40). In the same vein the proofs of Theorems 3.1 and 3.2 are divided into two parts. The easier upper bound is first derived and then an estimate of the lower bound is obtained.

**PROOF OF THEOREM 3.1:** (a) *Upper bound:* The relation  $W_n \in k_n(a_n + C)$  is equivalent to  $k_n^{-1}(W_n - k_n a_n) \in C$ . Then, from the definition of the polar cone  $C^+, \langle t, W_n - k_n a_n \rangle \geq 0$  for  $t \in C^+$ . Hence, it follows that  $P(W_n \in k_n(a_n + C)) \leq \varphi_n(t) \exp(-k_n \langle t, a_n \rangle)$ , for  $t \in D \cap C^+$  and using assumption (ii),

$$(4.9) \quad \limsup_{n \rightarrow \infty} [P(W_n \in k_n(a_n + C))]^{1/k_n} \leq \exp[c(h) - \langle h, a \rangle]$$

for  $h \in D \cap C^+$ . Since  $a = c'(h)$  and  $h \in D_0$  the equality  $\exp[c(h) - \langle h, a \rangle] = \inf_{t \in D_0} \exp[c(t) - \langle t, a \rangle]$  is obvious.

(b) *Lower bound:* Assumption (iv) implies the existence of a  $k$ -dimensional open interval  $(a_1, a_2) \subset U_\varepsilon(a) \cap (a + C)^0$  with  $(a_1, a_2) \cap \{c'(t) : t \in D\} \neq \emptyset$  for an arbitrary  $\varepsilon > 0$ . Let  $t = t(\varepsilon)$  be such that  $c'(t) \in (a_1, a_2)$ . Then for  $n$  sufficiently large

$$(4.10) \quad \begin{aligned} &P(W_n \in k_n(a_n + C)) \\ &\geq P(W_n \in k_n(a_1, a_2)) \\ &= \varphi_n(t) \int_{\{W_n \in k_n(a_1, a_2)\}} \exp(-\langle t, W_n \rangle) \frac{\exp \langle t, W_n \rangle}{\varphi_n(t)} dP \\ &\geq \varphi_n(t) \exp(-k_n \langle t, a_0 \rangle) P_{t,n}(W_n \in k_n(a_1, a_2)), \end{aligned}$$

where  $a_0 = (a_{10}, \dots, a_{k0})^T$  is defined by  $a_{i0} = a_{i1}$  (if  $t_i \leq 0$ ) and  $a_{i0} = a_{i2}$  (if  $t_i > 0$ ). In view of Lemma 3.1 and assumption (i) relation (4.10) yields

$$(4.11) \quad \begin{aligned} \liminf_{n \rightarrow \infty} [P(W_n \in k_n(a_n + C))]^{1/k_n} &\geq \exp[c(t) - \langle t, a_0 \rangle] \\ &\geq \inf_{t \in D_0} \exp[c(t) - \langle t, a_0 \rangle] = \rho(a_0). \end{aligned}$$

Now  $\lim_{\varepsilon \rightarrow 0} a_0 = a \in \{c'(t) : t \in D \cap C^+\}^0$ , and  $\rho(a) > 0$ . By the continuity of  $\rho$  in a neighborhood of  $a$  (cf. Remark 3.1) relation (4.11) then implies that

$$(4.12) \quad \liminf_{n \rightarrow \infty} [P(W_n \in k_n(a_n + C))]^{1/k_n} \geq \rho(a).$$

From (4.9) and (4.12) the proof is now complete.

Before the proof of Theorem 3.2 we note some useful facts:

REMARK 4.1. Using (2.4) the strict convexity of  $c$  implies that the solution  $h$  of  $a = c'(h)$  in Theorem 3.2 is unique. Hence  $h = h(a)$ , and  $h \neq 0$  if  $a \neq 0$ , because  $\rho(0) = \inf_{t \in D_0} \exp[c(t)] = \exp[c(0)]$  in view of the equalities  $c'(0) = \lim_{n \rightarrow \infty} k_n^{-1} \psi_n'(0) = \lim_{n \rightarrow \infty} k_n^{-1} EW_n = 0$ .

REMARK 4.2. For  $h(a) \neq 0$  we define the hyperplane

$$H(a) = \{x \in R^k : \langle h(a), x - a \rangle = 0\}$$

with the open half-spaces

$$\begin{aligned} H^+(a) &= \{x \in R^k : \langle h(a), x - a \rangle > 0\}, \\ H^-(a) &= \{x \in R^k : \langle h(a), x - a \rangle < 0\}. \end{aligned}$$

As mentioned in Remark 3.1 it follows that

$$\rho(a) = \exp[c(h(a)) - \langle h(a), a \rangle] < 1$$

for all  $a \in \partial A$ , which in turn implies that

$$(4.13) \quad \langle h(a), a \rangle > c(h(a)),$$

and  $0 \in H^-(a)$ , since  $c(h(a)) > c(0) = 0$ , for  $a \in \partial A$ . Moreover, from  $c'(0) = 0 \in A^0$ ,  $c'(h(a)) = a \in \partial A$ , and the continuity of  $c'$  on  $D$  (cf. (2.3)), it follows that the vectors  $h(a)$  are outside a neighborhood of the origin. Hence a real number  $c_0$  exists such that

$$(4.14) \quad c(h(a)) \geq c_0 > 0 = c(0) = \inf_{t \in D_0} c(t)$$

for all  $a \in \partial A$ .

REMARK 4.3. For  $\varepsilon > 0$  and  $a \in \partial A$  let  $H_\varepsilon(a)$  denote the supporting hyperplane of the  $\varepsilon$ -neighborhood  $U_\varepsilon(a)$  which parallels  $H(a)$  and is tangential to  $U_\varepsilon(a)$  at a unique point  $a_\varepsilon \in H^-(a)$ . We then have

$$H_\varepsilon(a) = \{x \in R^k : \langle h(a), x - a_\varepsilon \rangle = 0\},$$

and  $H_\varepsilon(a)$  also determines the open half-spaces

$$\begin{aligned} H_\varepsilon^+(a) &= \{x \in R^k : \langle h(a), x - a_\varepsilon \rangle > 0\}, \\ H_\varepsilon^-(a) &= \{x \in R^k : \langle h(a), x - a_\varepsilon \rangle < 0\}. \end{aligned}$$

Using (4.13), (4.14) and the boundedness of  $\{h(a) : a \in \partial A\}$ , a positive real number  $\varepsilon_0$  can be chosen such that

$$(4.15) \quad 0 \in H_\varepsilon^-(a) \quad \text{for all } a \in \partial A \quad \text{and all } \varepsilon \in (0, \varepsilon_0).$$

REMARK 4.4. For every  $\varepsilon \in (0, \varepsilon_0)$  there exists a finite number of neighborhoods  $U_\varepsilon(a)$ ,  $a \in \partial A$  (say)  $U_\varepsilon(a_1), \dots, U_\varepsilon(a_K)$ , where  $K = K(\varepsilon)$  and with

$$(4.16) \quad \partial A \subset \bigcup_{i=1}^K U_\varepsilon(a_i).$$

We now show that

$$(4.17) \quad B_K \equiv \bigcap_{i=1}^K H_\varepsilon^-(a_i) \subset A^0.$$

First we note from (4.15) that  $B_K$  is a convex set containing the origin. Also,

$$(4.18) \quad a \notin B_K \quad \text{for all } a \in \partial A$$

from (4.16) and the definition of  $H_\varepsilon^-(a_i)$ . But then  $a \notin B_K$  for all  $a \in \mathbb{C}\bar{A}$ , as otherwise  $a \in B_K$  for some  $a \in \mathbb{C}\bar{A}$  and  $0 \in B_K$  would imply that  $\{\lambda a : \lambda \in [0, 1]\}$  is a subset of  $B_K$ . This would contradict (4.18).

PROOF OF THEOREM 3.2: (a) *Upper bound:* From (4.17) we have for  $\varepsilon \in (0, \varepsilon_0)$

$$(4.19) \quad \begin{aligned} P(W_n \notin k_n A) &\leq P(W_n \notin k_n(\bigcap_{i=1}^K H_\varepsilon^-(a_i))) \\ &= P(W_n \in k_n(\bigcup_{i=1}^K (H_\varepsilon^-(a_i) \cup H_\varepsilon^+(a_i)))) \\ &\leq \sum_{i=1}^K P(W_n \in k_n(H_\varepsilon^-(a_i) \cup H_\varepsilon^+(a_i))) \\ &= \sum_{i=1}^K P(\langle h(a_i), W_n - k_n a_{i,\varepsilon} \rangle \geq 0). \end{aligned}$$

Here  $a_{i,\varepsilon}$  is the unique point in  $H^-(a_i)$  at which the supporting hyperplane  $H_\varepsilon^-(a_i)$  of  $U_\varepsilon(a_i)$  that parallels  $H(a_i)$  is tangential to  $U_\varepsilon(a_i)$  (cf. Remark 4.3).

The probabilities for the sum can be bounded by

$$(4.20) \quad \begin{aligned} P(\langle h(a_i), W_n - k_n a_{i,\varepsilon} \rangle \geq 0) \\ \leq E \exp \langle h(a_i), W_n - k_n a_{i,\varepsilon} \rangle \\ = \varphi_n(h(a_i)) \exp(-\langle h(a_i), k_n a_{i,\varepsilon} \rangle) \exp \langle h(a_i), k_n(a_i - a_{i,\varepsilon}) \rangle. \end{aligned}$$

As already mentioned, the set  $\{h(a) : a \in \partial A\}$  is bounded, i.e., we may say  $|h(a)| \leq h_0$  for all  $a \in \partial A$ . Now

$$(4.21) \quad |\langle h(a_i), k_n(a_i - a_{i,\varepsilon}) \rangle| \leq |h(a_i)| |k_n(a_i - a_{i,\varepsilon})| \leq k_n h_0 \varepsilon.$$

Finally, assumption (ii') yields

$$(4.22) \quad \begin{aligned} \lim_{n \rightarrow \infty} [\varphi_n(h(a_i)) \exp(-\langle h(a_i), k_n a_{i,\varepsilon} \rangle)]^{1/k_n} \\ = \exp[c(h(a_i)) - \langle h(a_i), a_i \rangle] = \rho(a_i). \end{aligned}$$

Using relations (4.19)—(4.22) it follows that

$$(4.23) \quad \begin{aligned} \limsup_{n \rightarrow \infty} [P(W_n \notin k_n A)]^{1/k_n} &\leq e^{h_0 \varepsilon} \max_{i=1, \dots, K} \rho(a_i) \\ &\leq e^{h_0 \varepsilon} \sup_{a \in \partial A} \rho(a). \end{aligned}$$

Upon letting  $\varepsilon$  tend to zero in (4.23) the upper bound is derived, i.e.,

$$(4.24) \quad \limsup_{n \rightarrow \infty} [P(W_n \notin k_n A)]^{1/k_n} \leq \sup_{a \in \partial A} \rho(a) = \rho(a^*).$$



(b) *Lower bound:* For every  $\varepsilon > 0$  assumptions (iv') and (v') imply the existence of a  $k$ -dimensional open interval

$$(a_1, a_2) \subset U_\varepsilon(a^*) \cap (CA)^0 \quad \text{with} \quad (a_1, a_2) \cap \{c'(t) : t \in D\} \neq \emptyset .$$

By the same arguments as those used in part (b) of the proof of Theorem 3.1 it then follows that

$$(4.25) \quad \liminf_{n \rightarrow \infty} [P(W_n \notin k_n A)]^{1/k_n} \geq \rho(a^*) .$$

This completes the proof.

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