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Rates of convergence of extreme for asymmetric normal distribution[☆]

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1. Introduction

Let $\{Y_n, n \geq 1\}$ be a sequence of independent and identically distributed (i.i.d.) random variables with common distribution function $F(x)$, let $M_n = \max\{Y_1, \ldots, Y_n\}$ denote the partial maximum. If there exist normalizing constants $a_n > 0$, $b_n \in R$ and non-degenerate distribution $G(x)$ such that

$$
\lim_{n\to\infty} P(M_n \le a_n x + b_n) = \lim_{n\to\infty} F^n(a_n x + b_n) = G(x)
$$
\n(1.1)

for all continuity points of *G*, then *G* must belong to one of the following three classes:

$$
\Phi_{\alpha}(x) = \begin{cases} 0, & x < 0, \\ exp\{-x^{-\alpha}\}, & x \ge 0, \end{cases}
$$

$$
\Psi_{\alpha}(x) = \begin{cases} exp\{-(-x)^{\alpha}\}, & x < 0, \\ 1, & x \ge 0, \end{cases}
$$

for some $\alpha > 0$ and

$$
\Lambda(x) = \exp\{-e^{-x}\}, \quad x \in R.
$$

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a b s t r a c t

In this paper, we derive the exact uniform convergence rate of the asymmetric normal distribution of the maximum and minimum to its extreme value limit. © 2013 Elsevier B.V. All rights reserved.

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If [\(1.1\)](#page-0-4) holds, we say that *F* belongs to one of the max domains of attraction of *G*, denoted by $F \in D_{\text{max}}(G)$. Criteria for $F \in$ $D_{\text{max}}(G)$ and the choice of normalizing constants, a_n and b_n , can be found in [de](#page--1-0) [Haan](#page--1-0) [\(1970\)](#page--1-0), [Galambos](#page--1-1) [\(1987\)](#page--1-1), [Leadbetter](#page--1-2) [et al.](#page--1-2) [\(1983\)](#page--1-2) and [Resnick](#page--1-3) [\(1987\)](#page--1-3).

Similarly, let $W_n = \min\{Y_1, \ldots, Y_n\}$ denote the partial minimum, there must exist normalizing constants $c_n > 0$, $d_n \in R$ and non-degenerate distribution $L(x)$ such that

$$
\lim_{n \to \infty} P(W_n \le c_n x + d_n) = \lim_{n \to \infty} \{1 - [1 - F(c_n x + d_n)]^n\} = L(x)
$$
\n(1.2)

for all continuity points of *L*(*x*), then *L* must belong to one of the following three classes:

$$
L_1(x) = \begin{cases} 1 - \exp\{-(x)^{-\alpha}\}, & x < 0, \\ 1, & x \ge 0, \end{cases}
$$

$$
L_2(x) = \begin{cases} 0, & x < 0, \\ 1 - \exp\{-x^{\alpha}\}, & x \ge 0, \end{cases}
$$

for some $\alpha > 0$ and

$$
L_3(x) = 1 - \exp{-e^x}, \quad x \in R.
$$

If [\(1.2\)](#page-1-0) holds, we say that *F* belongs to one of the min domains of attraction of *L*, denoted by $F \in D_{\text{min}}(L)$. Criteria for $F \in D_{\min}(L)$ and the choice of normalizing constants, c_n and d_n , can be found in [Galambos](#page--1-1) [\(1987\)](#page--1-1).

One interesting problem in extreme value is the convergence rate of $F^n(a_nx + b_n)$ to any one of the extreme value distributions. For the uniform convergence rate of $F^n(a_nx + b_n)$ to its extreme value limit, $\Lambda(x)$, [Hall](#page--1-4) [and](#page--1-4) [Wellner](#page--1-4) [\(1979\)](#page--1-4) showed that the convergence rate is proportional to 1/*n* if *F* is exponential. For the normal distribution, [Hall](#page--1-5) [\(1979\)](#page--1-5) proved the following result:

$$
\frac{c_1}{\log n} < \sup_{x \in R} |\Phi^n(a_n x + b_n) - \Lambda(x)| < \frac{c_2}{\log n}
$$

for $n > n_0$, where constants $0 < c_1 < c_2$ and $\Phi(x)$ denotes the standard normal distribution function, and the norming constants a_n and b_n are given by

$$
2\pi b_n^2 \exp(b_n^2) = n^2, \qquad a_n = b_n^{-1}.
$$

[Hall](#page--1-5) [\(1979\)](#page--1-5) showed that 1/ log *n* is the best convergence rate for the maxima of normal random variables. For related work on the uniform convergence rate of extremes, see [Smith](#page--1-6) [\(1982\)](#page--1-6), [Falk](#page--1-7) [\(1986\)](#page--1-7) and [Kaufmann](#page--1-8) [\(1995\)](#page--1-8). For general distribution functions, [de](#page--1-9) [Haan](#page--1-9) [and](#page--1-9) [Resnick](#page--1-9) [\(1996\)](#page--1-9) gave the rate of uniform convergence under a second order regular variation condition involved in a distribution function *F* with second derivative *F*["] which was weakened to the case of *F* without any derivative requirement in [Cheng](#page--1-10) [and](#page--1-10) [Jiang](#page--1-10) [\(2001\)](#page--1-10). For recent work, see [Peng](#page--1-11) [et al.\(2010\)](#page--1-11), [Lin](#page--1-12) [et al.\(2011\)](#page--1-12), [Liao](#page--1-13) [and](#page--1-13) [Peng](#page--1-13) [\(2012\)](#page--1-13) and [Liao](#page--1-14) [et al.](#page--1-14) [\(2012\)](#page--1-14): [Peng](#page--1-11) [et al.](#page--1-11) [\(2010\)](#page--1-11) derived convergence rates of the distribution of maxima for random sequences obeying the general error distribution; [Lin](#page--1-12) [et al.](#page--1-12) [\(2011\)](#page--1-12) obtained convergence rates of the distribution of maxima for random sequences obeying the short-tailed symmetric distribution; [Liao](#page--1-13) [and](#page--1-13) [Peng](#page--1-13) [\(2012\)](#page--1-13) discussed convergence rates of the distribution of maxima for random sequences obeying the lognormal distribution; [Liao](#page--1-14) [et al.](#page--1-14) [\(2012\)](#page--1-14) established two different convergence rates of the distribution of maxima for random sequences obeying the skew normal distribution. Our aim in this paper is to consider the uniform convergence rate of [\(1.1\)](#page-0-4) and [\(1.2\)](#page-1-0) when X_n follows the asymmetric normal distribution (AX distribution). The AM distribution is one of the most widely applied distributions in statistics [\(Alan](#page--1-15) [and](#page--1-15) [Pond,](#page--1-15) [1988;](#page--1-15) [Aigner](#page--1-16) [et al.,](#page--1-16) [1977;](#page--1-16) [George](#page--1-17) [and](#page--1-17) [Greg,](#page--1-17) [1977;](#page--1-17) [Micceri,](#page--1-18) [1989\)](#page--1-18). For example, [Kato](#page--1-19) [et al.](#page--1-19) [\(2002\)](#page--1-19) apply it to pattern recognition. The asymmetric normal distribution has attracted applications in subsidence analysis and training and overtraining markers in sport events. Some references are [Johnson](#page--1-20) [et al.](#page--1-20) [\(1994\)](#page--1-20), [Azzalini](#page--1-21) [and](#page--1-21) [Dalla](#page--1-21) [\(1996\)](#page--1-21) and [Bennett](#page--1-22) [\(2003\)](#page--1-22). The probability density function (pdf) is given by:

$$
F'(x) = \frac{2}{\sqrt{2\pi}(\sigma_l + \sigma_r)} \begin{cases} \exp\left\{-\frac{(x-\theta)^2}{2\sigma_l^2}\right\}, & x \le \theta, \\ \exp\left\{-\frac{(x-\theta)^2}{2\sigma_r^2}\right\}, & x > \theta, \end{cases}
$$
(1.3)

where $\sigma_l > 0$, $\sigma_r > 0$ and $\theta \in R$.

This paper is organized as follows: a lemma and some auxiliary results are given in Section [2.](#page-1-1) Main results are given in Section [3.](#page--1-23) Their proofs are deferred to Section [4.](#page--1-24)

2. Preliminaries

In order to obtain the uniform convergence rate of extremes from the AM distribution, we will give a lemma and two propositions. The following lemma gives Mills-type ratios for the $A\mathcal{N}$ distribution.

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