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# Extremes of $\alpha(t)$ -locally stationary Gaussian processes with non-constant variances

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#### ABSTRACT

With motivation from [9], in this paper we derive the exact tail asymptotics of  $\alpha(t)$ -locally stationary Gaussian processes with non-constant variance functions. We show that some certain variance functions lead to qualitatively new results. © 2016 Elsevier Inc. All rights reserved.

#### 1. Introduction and main result

For X(t),  $t \in [0,T]$ , T > 0 a centered stationary Gaussian process with unit variance and continuous sample paths Pickands derived in [20] that

$$\mathbb{P}\left\{\sup_{t\in[0,T]}X(t)>u\right\}\sim T\mathcal{H}_{\alpha}a^{1/\alpha}u^{2/\alpha}\mathbb{P}\left\{X(0)>u\right\},\quad u\to\infty,$$
(1)

provided that the correlation function r satisfies

$$1 - r(t) \sim a |t|^{\alpha}, \quad t \downarrow 0, \quad a > 0, \quad \text{and } r(t) < 1, \ \forall \ t \neq 0,$$
 (2)

with  $\alpha \in (0,2]$  (~ means asymptotic equivalence when the argument tends to 0 or  $\infty$ ). Here the classical Pickands constant  $\mathcal{H}_{\alpha}$  is defined by

$$\mathcal{H}_{\alpha} = \lim_{T \to \infty} T^{-1} \mathbb{E} \left\{ \sup_{t \in [0,T]} e^{\sqrt{2}B_{\alpha}(t) - t^{\alpha}} \right\},\,$$

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where  $B_{\alpha}(t), t \ge 0$  is a standard fractional Brownian motion with Hurst index  $\alpha/2 \in (0, 1]$ , see [20,21,5,12, 10,14,7,23,11,13,6,15] for various properties of  $\mathcal{H}_{\alpha}$ .

The deep contribution [3] introduced the class of locally stationary Gaussian processes with index  $\alpha$ , i.e., a centered Gaussian process  $X(t), t \in [0,T]$  with a constant variance function, say equal to 1, and correlation function satisfying

$$r(t,t+h) = 1 - a(t)|h|^{\alpha} + o(|t|^{\alpha}), \ h \to 0,$$

uniformly with respect to  $t \in [0, T]$ , where  $\alpha \in (0, 2]$  and a(t) is a bounded, strictly positive and continuous function.

Clearly, the class of locally stationary Gaussian processes includes the stationary ones. It allows for some minor fluctuations of dependence at t and at the same time keeps stationary structure at the local scale. See [3,4,18] for studies on the locally stationary Gaussian processes with index  $\alpha$ .

In [9] the tail asymptotics of the supremum of  $\alpha(t)$ -locally stationary Gaussian processes are investigated. Such processes and random fields are of interest in various applications, see [9] and the recent contributions [2,16,17]. Following the definition in [9], a centered Gaussian process  $X(t), t \in [0, T]$  with continuous sample paths and unit variance is  $\alpha(t)$ -locally stationary if the correlation function  $r(\cdot, \cdot)$  satisfies the following conditions:

(i)  $\alpha(t) \in C([0,T])$  and  $\alpha(t) \in (0,2]$  for all  $t \in [0,T]$ ;

- (ii)  $a(t) \in C([0,T])$  and  $0 < \inf\{a(t) : t \in [0,T]\} \le \sup\{a(t) : t \in [0,T]\} < \infty;$
- (iii) uniformly for  $t \in [0, T]$

$$1 - r(t, t+h) = a(t)|h|^{\alpha(t)} + o(|h|^{\alpha(t)}), \ h \to 0,$$

where  $f(t) \in C(\mathcal{T})$  means that f(t) is continuous on  $\mathcal{T} \subset \mathbb{R}$ .

In this paper, we shall consider the case that the variance function  $\sigma^2(t) = Var(X(t))$  is not constant, assuming instead that:

(iv)  $\sigma(t)$  attains its maximum equal to 1 over [0, T] at the unique point  $t_0 \in [0, T]$  and for some constants  $c, \gamma > 0$ ,

$$\frac{1}{\sigma(t)} = 1 + c e^{-|t-t_0|^{-\gamma}} (1+o(1)), \quad t \to t_0.$$

A crucial assumption in our result is that similar to the variance function, the function  $\alpha(t)$  has a certain behavior around the extreme point  $t_0$ . Specifically, as in [9] we shall assume:

(v) there exist  $\beta, \delta, b > 0$  such that

$$\alpha(t+t_0) = \alpha(t_0) + b|t|^\beta + o(|t|^{\beta+\delta}), \quad t \to 0.$$

**Remark 1.1.** We remark that  $t_0$  does not need to be the unique point such that  $\alpha(t)$  is minimal on [0, T], which is different from [9]. For instance,  $[0, T] = [0, 2\pi]$ ,  $t_0 = 0$  and  $\alpha(t) = 1 + \frac{1}{2}\sin(t)$ , then 0 is not the minimum point of  $\alpha(t)$  over  $[0, 2\pi]$  which means assumptions about  $\alpha(t)$  in [9] are not satisfied but assumption (v) here is satisfied with

$$\alpha(t) = 1 + \frac{1}{2}|t| + o(|t|^{\frac{3}{2}}), \ t \to 0.$$

Below we set  $\alpha := \alpha(t_0)$ ,  $a := a(t_0)$  and write  $\Psi$  for the survival function of an N(0,1) random variable. Further, define  $0^a = \infty$  for a < 0. Our main result is stated in the next theorem. Download English Version:

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