

Mathematical Foundations of Infinite-Dimensional Statistical Models

2.7 Asymptotics for Extremes of Stationary Gaussian Processes

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Asymptotics for Extremes of Stationary Gaussian Processes

- ▶ Focus on the limiting distribution of $\max_{1 \leq k \leq n} |X_n|$ for Gaussian sequences and of $\sup_{0 \leq t \leq T} |X(t)|$ for Gaussian processes on \mathbb{R}^+ .
- ▶ Only theoretical interest because the speed of convergence for limits theorems is very slow (for sequences, of the order of $1/\log n$).
- ▶ Here we consider stationary processes.

Gaussian Sequences

- **Theorem 2.7.1** Let $g_i, i \leq n$, be independent standard normal random variables. For each $n \geq e^2$, set

$$a_n = (2 \log n)^{1/2}, \quad b_n = a_n - \frac{\log \log n + \log \pi}{2a_n}$$

Then, for all $x \in \mathbb{R}$,

$$\lim_{n \rightarrow \infty} \Pr \left\{ a_n \left(\max_{1 \leq i \leq n} |g_i| - b_n \right) \leq x \right\} = \exp(-e^{-x}). \quad (2.90)$$

Gaussian Sequences

- **Proof of Theorem 2.7.1** For g standard normal, let $\Phi(x) = \Pr\{|g| \leq x\}$, and let $\phi(x) = (2/\pi)^{1/2} e^{-x^2/2}$, $x \geq 0$. Then, for $u > 0$,

$$\frac{u}{1+u^2} \phi(u) \leq 1 - \Phi(u) \leq \frac{1}{u} \phi(u).$$

For $x \in \mathbb{R}$ fixed, set $u_n = x/a_n + b_n$. By definition of u_n , we have

$$\frac{n\phi(u_n)}{u_n} \rightarrow e^{-x}, \text{ as } n \rightarrow \infty.$$

Finally,

$$\begin{aligned} \Pr \left\{ \max_{i \leq n} |g_i| \leq u_n \right\} &= \Phi^n(u_n) = [1 - (1 - \Phi(u_n))]^n \\ &\approx e^{-n(1-\Phi(u_n))} \approx e^{-n\phi(u_n)/u_n} \rightarrow \exp(-e^{-x}), \end{aligned}$$

as $n \rightarrow \infty$.

Gaussian Sequences

- ▶ **Definition 2.7.2** A sequence $\{\xi_n\}$ of r.v.s is stationary if for any $n_1, \dots, n_m \in \mathbb{N}$, $m > 0$ and for any $k \geq 0$, the joint probability law of the r.v.s $\xi_{n_1+k}, \dots, \xi_{n_m+k}$ does not depend on k .
- ▶ If $\{\xi_n\}$ is a centred (jointly) Gaussian sequence, then it is stationary iff the covariances $\mathbb{E}(\xi_m \xi_{n+m})$ do not depend on m . In this case, $r(n) = \mathbb{E}(\xi_m \xi_{n+m})$ is the covariance function of the sequence.

Gaussian Sequences

- **Lemma 2.7.3** Let $\{r_j\}$ be a sequence of numbers s.t. $\sup_n |r_n| < 1$ and $r_n \log n \rightarrow 0$ as $n \rightarrow \infty$, and let $\{u_n\}$ be a sequence of positive constants s.t. $\sup_n n(1 - \Phi(u_n)) < \infty$. Then

$$\lim_{n \rightarrow \infty} n \sum_{j=1}^n |r_j| \exp\left(-\frac{u_n^2}{1 + |r_j|}\right) = 0. \quad (2.93)$$

Gaussian Sequences

- **Theorem 2.7.4** Let $\{\xi_i\}_{n=1}^{\infty}$ be a stationary sequence of standard normal random variables s.t. its sequence of covariances $r_n = \mathbb{E}(\xi_m \xi_{n+m})$ satisfies $r_n \log n \rightarrow 0$. For each $n \geq e^2$, set

$$a_n = (2 \log n)^{1/2}, \quad b_n = a_n - \frac{\log \log n + \log \pi}{2a_n}$$

Then, for all $x \in \mathbb{R}$,

$$\lim_{n \rightarrow \infty} \Pr \left\{ a_n \left(\max_{1 \leq i \leq n} |\xi_i| - b_n \right) \leq x \right\} = \exp(-e^{-x}). \quad (2.96)$$

Proof By Thm. 2.4.7, Lemma 2.7.3,

$$\left| \Pr \left\{ a_n \left(\max_{1 \leq i \leq n} |\xi_i| - b_n \right) \leq x \right\} - \Pr \left\{ a_n \left(\max_{1 \leq i \leq n} |g_i| - b_n \right) \leq x \right\} \right| \leq \frac{2}{\pi} n \sum_{j=1}^n |r_j| \exp \left(-\frac{u_n^2}{1 + |r_j|} \right)$$

Gaussian Processes Indexed by $[0, \infty)$

- ▶ A stochastic process $\xi(t)$, $t \geq 0$, is *stationary* if the finite-dimensional marginal distributions of $\xi(t_1 + s), \dots, \xi(t_n + s)$ do not depend on s , for all n and for any s s.t. $t_i + s \geq 0, i = 1, \dots, n$.
- ▶ If $\xi(t)$, $t \geq 0$ is a centred GP, then it is stationary iff the covariances $\mathbb{E}(\xi(s)\xi(s+t))$ do not depend on s . In this case, $r(t) = \mathbb{E}(\xi(0)\xi(t))$ is the covariance function of the GP.
- ▶ A stationary GP is *normalized* if $\xi(t)$, $t \geq 0$ is standard normal.

Gaussian Processes Indexed by $[0, \infty)$

- ▶ Now let $\xi(t)$, $t \in [0, \infty)$, be a normalized stationary GP with continuous sample paths and with covariance $r(t) = \mathbb{E}(\xi(s)\xi(s+t))$ satisfying, for some $\alpha \in (0, 2]$ and $C \in (0, \infty)$,

$$r(t) < 1, \text{ for } t > 0 \text{ and } r(t) = 1 - C|t|^\alpha + o(|t|)^\alpha, \text{ as } t \rightarrow 0. \quad (2.97)$$

At the end of this section, for some a_T and b_T , we want to get

$$\lim_{T \rightarrow \infty} \Pr \left\{ a_T \left(\sup_{t \in [0, T]} |\xi(t)| - b_T \right) \leq x \right\} = \exp(-e^{-x}),$$

for all $x \in \mathbb{R}$.

Gaussian Processes Indexed by $[0, \infty)$

- ▶ The proof is divided into three parts:
 1. A limit theorem for the high excursions of the process over a fixed finite interval;
 2. The process is replaced by its absolute value;
 3. The limit theorem for an interval increasing to infinity.

Gaussian Processes Indexed by $[0, \infty)$

- ▶ Let $\bar{\Phi}(u) = (2\pi)^{-1/2} \int_u^\infty e^{-x^2/2} dx$ be the tail probability function for the standard normal distribution.
- ▶ For each $\alpha \in (0, 2]$, $\zeta(t)$ is a (auxiliary) nonstationary GP with mean $-|t|^\alpha$ and covariance $r_\zeta(s, t) = t^\alpha + s^\alpha - |t - s|^\alpha$.
- ▶ **Theorem 2.7.5** Let $\xi(t)$, $t \in [0, \infty)$, be a normalized centred stationary GP with continuous sample paths and with covariance $r(t)$ satisfying condition (2.97) for some $\alpha \in [0, 2]$ and $C \in (0, \infty)$. Then, for any $h \in (0, \infty)$,

$$\lim_{u \rightarrow \infty} \frac{1}{u^{2/\alpha} h \bar{\Phi}(u)} \Pr \left\{ \sup_{t \in [0, h]} \xi(t) > u \right\} = C^{1/\alpha} H_\alpha, \quad (2.99)$$

where $H_\alpha = \lim_{a \downarrow 0} \frac{1}{a} \Pr \left\{ \sup_{k \in \mathbb{N}} \zeta(ak) + \eta \leq 0 \right\} \in (0, \infty)$.

Gaussian Processes Indexed by $[0, \infty)$

- ▶ Corollary 2.7.6 Under the assumptions of Thm. 2.7.5,

$$\lim_{u \rightarrow \infty} \frac{1}{u^{2/\alpha} h \bar{\Phi}(u)} \Pr \left\{ \sup_{t \in [0, h]} |\xi(t)| > u \right\} = 2C^{1/\alpha} H_\alpha. \quad (2.108)$$

Gaussian Processes Indexed by $[0, \infty)$

- ▶ Now we obtain a limit theorem for the interval $[0, h]$ increasing to infinity.
- ▶ **Proposition 2.7.8** Let $\xi(t)$, $t \in [0, \infty)$, be a normalized centred stationary GP with continuous sample paths and with covariance $r(t)$ satisfying $r(t) < 1$, for $t > 0$, $r(t) = 1 - C|t|^\alpha + o(|t|^\alpha)$, for some $\alpha \in (0, 2]$ as $t \rightarrow 0$, and $r(t) \log t \rightarrow 0$, as $t \rightarrow \infty$. Let $\tau > 0$, and for each T , let $u = u_T$ be s.t. condition $T\mu(u_T) \rightarrow \tau$, as $T \rightarrow \infty$. Then

$$\lim_{T \rightarrow \infty} \Pr \left\{ \sup_{0 \leq t \leq T} |\xi(t)| \leq u_T \right\} \rightarrow e^{-\tau}.$$

Gaussian Processes Indexed by $[0, \infty)$

- **Theorem 2.7.9** Let $\xi(t)$, $t \in [0, \infty)$, be a normalized centred stationary GP with continuous sample paths and with covariance $r(t)$ satisfying $r(t) < 1$, for $t > 0$, $r(t) = 1 - C|t|^\alpha + o(|t|^\alpha)$, for some $\alpha \in (0, 2]$ as $t \rightarrow 0$, and $r(t) \log t \rightarrow 0$, as $t \rightarrow \infty$. Set $a_T = (2 \log T)^{1/2}$ and

$$b_T = a_T + \frac{(2 - \alpha)/\alpha \log \log T + 2 \log((2\pi)^{-1/2} 2^{(2+\alpha)/\alpha} C^{1/\alpha} H_\alpha)}{2a_T}. \quad (2.120)$$

Then, for all $x \in \mathbb{R}$,

$$\lim_{T \rightarrow \infty} \Pr \left\{ a_T \left(\sup_{t \in [0, T]} |\xi(t)| - b_T \right) \leq x \right\} = \exp(-e^{-x}).$$

Notes

- ▶ A centred GP $Y(t)$, $t \in \mathbb{R}$, is cyclostationary if its covariance function $t \mapsto r(t, t + \nu) := \mathbb{E}(Y(t)Y(t + \nu))$ is periodic in t for every $\nu \in \mathbb{R}$ with period independent of ν .
- ▶ **Theorem 2.8.3** Let $X(t)$, $t \in \mathbb{R}$ be a cyclostationary, centred GP with period 1, variance $\sigma_X(t)$ and covariance $r_X(s, t)$. Under the some conditions, for all $x \in \mathbb{R}$, we have

$$\lim_{T \rightarrow \infty} \Pr \left\{ a_T \left(\sup_{t \in [0, T]} |X(t)| - b_T \right) \leq x \right\} = \exp(-e^{-x}),$$

where $a_T = \sqrt{2 \log T}$ and

$$b_T = a_T - \frac{\log \log T + \log \pi - \log \left(1 - \frac{\mathbb{E}(X'(t_0))^2}{\sigma_X''(t_0)} \right)}{2a_T}.$$