

## 2 Random real zeroes

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### 2a No derivatives

We consider random trigonometric sums of the form

$$X(t) = \sum_{k=1}^N a_k \operatorname{Re}((X_{2k-1} + iX_{2k})e^{i\lambda_k t}) = \sum_{k=1}^N a_k (X_{2k-1} \cos \lambda_k t - X_{2k} \sin \lambda_k t)$$

where  $a_1, \dots, a_N > 0$ ,  $0 < \lambda_1 < \dots < \lambda_N < \infty$ , and  $X_1, \dots, X_{2N}$  are independent standard normal (that is, distributed  $N(0, 1)$ ) random variables.

The distribution of  $X(\cdot)$  is shift-invariant. In other words,  $X(\cdot)$  is a stationary random process.

It may happen that all  $\lambda_k/(2\pi)$  are integers, and then  $X(t+1) = X(t)$ , but generally  $X(\cdot)$  need not be periodic.

ASSUMPTION A:

$$\sum_{k=1}^N a_k^2 = 1.$$

That is,  $X(0) \sim N(0, 1)$ . Otherwise we may rescale  $X$ .

ASSUMPTION  $A_n$ : assumption A holds, and in addition,

$$\forall \lambda \in [0, \infty) \quad \sum_{k: \lambda_k \in [\lambda, \lambda+1]} a_k^2 \leq \frac{1}{n}.$$

The correlation function

$$\mathbb{E}(X(0)X(t)) = \sum_{k=1}^N a_k^2 \cos \lambda_k t$$

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<sup>1</sup>This section is, to a large extent, a one-dimensional counterpart of the work: F. Nazarov, M. Sodin (2009), “On the number of nodal domains of random spherical harmonics”, *American Journal of Mathematics* **131**:5, 1337–1357.

need not decay in  $n$  exponentially. For example,  $N \gg n$ ,  $\lambda_k = nk/N$ ,  $a_k = 1/\sqrt{N}$ ; then

$$\mathbb{E}(X(0)X(t)) \approx \frac{1}{n} \int_0^n \cos \lambda t \, d\lambda = \frac{\sin nt}{nt};$$

another example gives

$$\mathbb{E}(X(0)X(t)) \approx \frac{2}{n^2} \int_0^n (n - \lambda) \cos \lambda t \, d\lambda = \left( \frac{\sin nt/2}{nt/2} \right)^2.$$

In amazing contrast, many interesting probabilities decay in  $n$  exponentially.

**2a1 Lemma.** Let  $X$  satisfy assumption  $A$ , and a measurable function  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$  be  $\gamma^1$ -integrable (that is,  $\int |\varphi| \, d\gamma^1 < \infty$ ). Then the random variable

$$\xi = \int_0^1 \varphi(X(t)) \, dt$$

is integrable, and

$$\mathbb{E} \xi = \int \varphi \, d\gamma^1.$$

**2a2 Theorem.** Let  $X$  satisfy assumption  $A_n$ , and a function  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$  be continuous almost everywhere, and

$$\sup_x \frac{|\varphi(x)|}{1 + |x|} < \infty.$$

Then the random variable  $\xi$  introduced above satisfies, for every  $\varepsilon > 0$ ,

$$\mathbb{P}(|\xi - \mathbb{E} \xi| \geq \varepsilon) \leq 2e^{-c_{\varepsilon, \varphi} n}$$

for some  $c_{\varepsilon, \varphi} > 0$  (dependent on  $\varepsilon$  and  $\varphi$  only, not on  $n$ ).

The same holds for  $\mathbb{R}^d$ -valued processes  $X(\cdot)$  provided that  $X(0) \sim \gamma^d$ . For  $f \in C[0, 1]$  denote

$$T(f) = \inf_g \int_0^1 |f(t) - g(t)| \, dt$$

where the infimum is taken over all measurable  $g : (0, 1) \rightarrow \mathbb{R}$  that send Lebesgue measure to  $\gamma^1$ .

**2a3 Theorem.** Let  $X$  satisfy assumption  $A_n$ . Then

$$\mathbb{P}(T(X(\cdot)) \geq \varepsilon) \leq 2e^{-c_{\varepsilon} n}$$

for some  $c_{\varepsilon} > 0$  dependent on  $\varepsilon$  only.

The same holds for  $\mathbb{R}^d$ -valued processes  $X(\cdot)$  (with  $\gamma^d$  in place of  $\gamma$ ).

A trivial rescaling of  $t$  by arbitrary  $L > 0$  turns Assumption  $A_n$ , Lemma 2a1 and Theorem 2a2 into the following.

ASSUMPTION  $A_{n,L}$ : assumption  $A$  holds, and in addition,

$$\forall \lambda \in [0, \infty) \quad \sum_{k: \lambda_k \in [\lambda, \lambda + \frac{1}{L}]} a_k^2 \leq \frac{1}{n}.$$

**2a4 Lemma.** Let  $X$  satisfy assumption  $A$ , and a measurable function  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$  be  $\gamma^1$ -integrable. Then the random variable

$$\xi = \frac{1}{L} \int_0^L \varphi(X(t)) dt$$

is integrable, and

$$\mathbb{E} \xi = \int \varphi d\gamma^1.$$

**2a5 Theorem.** Let  $X$  satisfy assumption  $A_{n,L}$ , and a function  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$  be continuous almost everywhere, and

$$\sup_x \frac{|\varphi(x)|}{1 + |x|} < \infty.$$

Then the random variable  $\xi$  introduced above satisfies, for every  $\varepsilon > 0$ ,

$$\mathbb{P}(|\xi - \mathbb{E} \xi| \geq \varepsilon) \leq 2e^{-c_{\varepsilon, \varphi} n}$$

for some  $c_{\varepsilon, \varphi} > 0$ .

## 2b One derivative

ASSUMPTION  $B$ :

$$\sum_{k=1}^N a_k^2 = 1 \quad \text{and} \quad \sum_{k=1}^N \lambda_k^2 a_k^2 = 1.$$

That is,  $X(0) \sim N(0, 1)$  and  $X'(0) \sim N(0, 1)$ . Otherwise we may rescale  $t$ . In fact,  $X(0)$  and  $X'(0)$  are independent; thus,  $(X(0), X'(0)) \sim \gamma^2$ .

ASSUMPTION  $B_{n,L}$ : assumption  $B$  holds, and in addition,

$$\forall \lambda \in [0, \infty) \quad \sum_{k: \lambda_k \in [\lambda, \lambda + \frac{1}{L}]} (1 + \lambda_k^2) a_k^2 \leq \frac{1}{n}.$$

If  $X$  satisfies  $B_{n,L}$  then the 2-dimensional process  $(X, X')$  satisfies  $A_{n,L}$ . Thus, Theorem 2a2 (2-dim version) may be applied to random variables of the form

$$\frac{1}{L} \int_0^L \varphi(X(t), X'(t)) dt.$$

But now we turn to a more interesting random variable.

**2b1 Theorem.** Let  $X$  satisfy assumption  $B$ , and a measurable function  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$  satisfy  $\int |\varphi(y)| |y| e^{-y^2/2} dy < \infty$ . Then the random variable

$$\xi = \frac{1}{L} \sum_{t \in [0, L], X(t)=0} \varphi(X'(t))$$

is integrable, and

$$\mathbb{E} \xi = \frac{1}{2\pi} \int \varphi(y) |y| e^{-y^2/2} dy.$$

(Assumption  $B$  does not contain  $L$ , but anyway, the theorem above holds for all  $L > 0$ .)

In particular (for  $\varphi(\cdot) = 1$ ), the expected number of zeroes per unit time is equal to  $1/\pi$ .<sup>1</sup>

The expected number (per unit time) of zeroes  $t$  such that  $|X'(t)| \leq \varepsilon$  is  $O(\varepsilon^2)$  as  $\varepsilon \rightarrow 0+$ .

## 2c Two derivatives

ASSUMPTION  $C_M$ :

$$\sum_{k=1}^N a_k^2 = 1, \quad \sum_{k=1}^N \lambda_k^2 a_k^2 = 1, \quad \text{and} \quad \sum_{k=1}^N \lambda_k^4 a_k^2 \leq M.$$

Thus,  $(X(0), X'(0)) \sim \gamma^2$ , and  $\mathbb{E} |X''(0)|^2 \leq M$ . In fact,  $X'(0)$  and  $X''(0)$  are independent, but  $\mathbb{E} (X(0)X''(0)) = -1$ .

ASSUMPTION  $C_{M,n,L}$ : assumption  $C_M$  holds, and in addition,

$$\forall \lambda \in [0, \infty) \quad \sum_{k: \lambda_k \in [\lambda, \lambda + \frac{1}{L}]} (1 + \lambda_k^2)^2 a_k^2 \leq \frac{1}{n}.$$

Clearly,  $C_{M,n,L}$  implies  $B_{n,L}$ .

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<sup>1</sup>“Rice’s formula” (Kac 1943, Rice 1945, Bunimovich 1951, Grenander and Rosenblatt 1957, Ivanov 1960, Bulinskaya 1961, Itô 1964, Ylvisaker 1965 et al. See [1, Sect. 10.3]). “... the famous Rice formula, undoubtedly one of the most important results in the applications of smooth stochastic processes” (R.J. Adler and J.E. Taylor, “Random fields and geometry”, Springer 2007; see Preface, page viii).

**2c1 Theorem.** Let  $X$  satisfy assumption  $C_{M,n,L}$ , and a function  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$  be continuous almost everywhere,  $\varphi(0) = 0$ , and

$$\sup_{x \neq 0} \frac{|\varphi(x)|}{|x|} < \infty .$$

Then the random variable

$$\xi = \frac{1}{L} \sum_{t \in [0,L], X(t)=0} \varphi(X'(t))$$

is integrable,

$$\mathbb{E} \xi = \frac{1}{2\pi} \int \varphi(y) |y| e^{-y^2/2} dy ,$$

and

$$\mathbb{P} ( |\xi - \mathbb{E} \xi| \geq \varepsilon ) \leq 2e^{-c_{M,\varepsilon,\varphi} n}$$

for some  $c_{M,\varepsilon,\varphi} > 0$ .

## References

- [1] H. Cramér, M.R. Leadbetter, *Stationary and related stochastic processes*, Wiley 1967.