

# A Gaussian Construction for an Erdős Problem on Zeros of Successive Derivatives

Checked draft for discussion

April 2026

## Abstract

We give a probabilistic construction of a transcendental entire function  $f$  with the following property: for every nonempty open set  $U \subset \mathbb{C}$ , all sufficiently high derivatives  $f^{(n)}$  have a zero in  $U$ . This implies that, for every infinite increasing sequence  $n_1 < n_2 < \dots$ , the union of zero sets

$$\bigcup_{k \geq 1} \{z \in \mathbb{C} : f^{(n_k)}(z) = 0\}$$

is dense in  $\mathbb{C}$ . The examples are almost sure samples of the Gaussian entire function

$$f(z) = \sum_{k=0}^{\infty} \xi_k \frac{z^k}{(k!)^{1-\beta}}, \quad \frac{1}{2} < \beta < 1,$$

where the  $\xi_k$  are independent standard complex Gaussian variables. The proof has two main ingredients: a saddle-point estimate for the covariance kernels of the derivatives, and an Offord-type exponential tail estimate for smooth zero statistics. For the latter, we include a short self-contained proof in the form needed here.

## 1 Statement and cofinite reformulation

The literal version of Erdős Problem #906 asks for a nonzero entire function  $f$  such that, for every infinite increasing sequence  $n_1 < n_2 < \dots$ , the set

$$\{z \in \mathbb{C} : f^{(n_k)}(z) = 0 \text{ for some } k\}$$

is dense in  $\mathbb{C}$ . Without an additional hypothesis this is trivial: every nonzero polynomial works, since all sufficiently high derivatives vanish identically. The intended question is therefore the transcendental version. We prove the following stronger cofinite form.

**Theorem 1.1** (Main theorem). *There exists a transcendental entire function  $f$  such that for every nonempty open set  $U \subset \mathbb{C}$  there is an integer  $N(U)$  for which*

$$f^{(n)} \text{ has a zero in } U \quad (n \geq N(U)).$$

Consequently, for every infinite increasing sequence  $n_1 < n_2 < \dots$ , the set

$$\bigcup_{k \geq 1} \{z \in \mathbb{C} : f^{(n_k)}(z) = 0\}$$

is dense in  $\mathbb{C}$ .

We first record the elementary equivalence between the subsequence formulation and the cofinite formulation.

**Lemma 1.2** (Cofinite reformulation). *Let  $f$  be an entire function. The following assertions are equivalent.*

(i) *For every infinite increasing sequence  $n_1 < n_2 < \dots$  of nonnegative integers, the set*

$$\bigcup_{k \geq 1} \{z \in \mathbb{C} : f^{(n_k)}(z) = 0\}$$

*is dense in  $\mathbb{C}$ .*

(ii) *For every nonempty open set  $U \subset \mathbb{C}$ , the set*

$$A_U := \{n \in \mathbb{N}_0 : f^{(n)} \text{ has a zero in } U\}$$

*is cofinite in  $\mathbb{N}_0$ .*

*Proof.* Assume (ii). Let  $n_1 < n_2 < \dots$  be infinite and let  $U \subset \mathbb{C}$  be nonempty and open. Since  $A_U$  is cofinite and  $n_k \rightarrow \infty$ , some  $n_k$  belongs to  $A_U$ . Thus the union of zero sets meets  $U$ ; since  $U$  was arbitrary, the union is dense.

Conversely, suppose that (ii) fails for some nonempty open set  $U$ . Then  $\mathbb{N}_0 \setminus A_U$  is infinite. Enumerate this complement increasingly as  $n_1 < n_2 < \dots$ . No derivative  $f^{(n_k)}$  has a zero in  $U$ , so the corresponding union of zero sets misses  $U$  and is not dense.  $\square$

## 2 The random entire function

Fix

$$\frac{1}{2} < \beta < 1$$

throughout the proof. Let  $\xi_0, \xi_1, \xi_2, \dots$  be independent standard complex Gaussian random variables, normalized by

$$\mathbb{P}(\xi_j \in dw) = \frac{1}{\pi} e^{-|w|^2} dA(w).$$

Define

$$f(z) = \sum_{k=0}^{\infty} \xi_k \frac{z^k}{(k!)^{1-\beta}}. \tag{1}$$

**Lemma 2.1.** *Almost surely, (1) defines a transcendental entire function.*

*Proof.* For every  $\varepsilon > 0$ ,

$$\sum_{k=1}^{\infty} \mathbb{P}\{|\xi_k| > e^{\varepsilon k}\} = \sum_{k=1}^{\infty} e^{-e^{2\varepsilon k}} < \infty.$$

By Borel–Cantelli, almost surely  $|\xi_k| \leq e^{\varepsilon k}$  for all sufficiently large  $k$ . Since

$$((k!)^{1-\beta})^{1/k} \sim (k/e)^{1-\beta} \rightarrow \infty,$$

the Taylor series has infinite radius of convergence. Also  $\mathbb{P}(\xi_k = 0) = 0$  for each  $k$ , hence, by countable intersection, almost surely all coefficients are nonzero. Thus  $f$  is not a polynomial.  $\square$

For  $n \geq 0$ , termwise differentiation gives

$$F_n(z) := f^{(n)}(z) = \sum_{m=0}^{\infty} \xi_{n+m} \frac{((n+m)!)^\beta}{m!} z^m. \quad (2)$$

For each fixed  $n$ ,  $F_n$  is a Gaussian analytic function. Its covariance kernel on the diagonal is

$$K_n(z, z) := \mathbb{E}|F_n(z)|^2 = \sum_{m=0}^{\infty} \frac{((n+m)!)^{2\beta}}{(m!)^2} |z|^{2m}. \quad (3)$$

Let  $Z_n$  denote the zero counting measure of  $F_n$ , counting multiplicity, and set  $\mu_n := \mathbb{E}Z_n$ .

We shall use the following standard form of the Edelman–Kostlan formula. It is included to fix normalizations.

**Lemma 2.2** (Edelman–Kostlan formula). *Let  $X$  be a Gaussian analytic function on a plane domain  $G$ , and suppose that its covariance  $K_X(z, z) = \mathbb{E}|X(z)|^2$  is positive on  $G$ . If  $Z_X$  is the zero counting measure of  $X$ , then*

$$\mathbb{E}Z_X = \frac{1}{4\pi} \Delta \log K_X(z, z),$$

in the sense of distributions, where  $\Delta = \partial_x^2 + \partial_y^2$ .

*Proof.* The Poincaré–Lelong formula in this normalization is

$$Z_X = \frac{1}{2\pi} \Delta \log |X|.$$

For each fixed  $z$ , the random variable  $X(z)$  has the same distribution as  $K_X(z, z)^{1/2}\zeta$ , where  $\zeta$  is a standard complex Gaussian. Therefore

$$\mathbb{E} \log |X(z)| = \frac{1}{2} \log K_X(z, z) + \mathbb{E} \log |\zeta|.$$

The last term is a finite constant, whose Laplacian is zero. Taking expectations in the distributional identity gives the result.  $\square$

Consequently,

$$\mu_n = \frac{1}{4\pi} \Delta \log K_n(z, z). \quad (4)$$

### 3 Expected zero measures

The deterministic core of the proof is the following asymptotic for the expected zero measures of the derivatives.

**Proposition 3.1.** *As  $n \rightarrow \infty$ ,*

$$n^{-\beta} \mu_n \longrightarrow \nu := \frac{1}{2\pi|z|} dA(z) \quad (5)$$

vaguely on  $\mathbb{C}$ . Equivalently, for every  $\varphi \in C_c^\infty(\mathbb{C})$ ,

$$\frac{1}{n^\beta} \int_{\mathbb{C}} \varphi d\mu_n \longrightarrow \int_{\mathbb{C}} \varphi(z) \frac{dA(z)}{2\pi|z|}. \quad (6)$$

The density  $1/|z|$  is locally integrable at the origin.

Write  $\lambda_n := n^\beta$ . Factoring out the constant  $(n!)^{2\beta}$  from (3), we have

$$K_n(z, z) = (n!)^{2\beta} A_n(|z|), \quad (7)$$

where

$$A_n(r) = \sum_{m=0}^{\infty} a_{n,m}(r), \quad a_{n,m}(r) = \frac{r^{2m}}{(m!)^2} \prod_{\ell=1}^m (n + \ell)^{2\beta}. \quad (8)$$

The empty product is 1.

**Lemma 3.2** (Laplace estimate). *For every  $R > 0$ ,*

$$\sup_{0 \leq r \leq R} \left| \frac{1}{\lambda_n} \log A_n(r) - 2r \right| \rightarrow 0. \quad (9)$$

Consequently,

$$\frac{1}{n^\beta} (\log K_n(z, z) - 2\beta \log(n!)) \rightarrow 2|z| \quad (10)$$

locally uniformly on  $\mathbb{C}$ .

*Proof.* We prove (9). For the lower bound, fix  $0 < \varepsilon < R$  and let  $r \in [\varepsilon, R]$ . Put  $m_n = \lfloor r\lambda_n \rfloor$ . Since  $\beta < 1$ , we have  $m_n/n \rightarrow 0$  uniformly for  $r \in [\varepsilon, R]$ . Also, uniformly in this range,

$$\sum_{\ell=1}^{m_n} \log(n + \ell) = m_n \log n + O\left(\frac{m_n^2}{n}\right) = m_n \log n + o(\lambda_n),$$

and Stirling's formula gives

$$\log(m_n!) = m_n \log m_n - m_n + o(\lambda_n).$$

Using  $\lambda_n = n^\beta$ , we obtain

$$\begin{aligned} \log a_{n,m_n}(r) &= 2m_n \log r + 2\beta \sum_{\ell=1}^{m_n} \log(n + \ell) - 2 \log(m_n!) \\ &= 2m_n \log r + 2m_n \log \lambda_n - 2m_n \log m_n + 2m_n + o(\lambda_n) \\ &= 2r\lambda_n + o(\lambda_n), \end{aligned}$$

uniformly for  $r \in [\varepsilon, R]$ . Hence

$$\liminf_{n \rightarrow \infty} \inf_{\varepsilon \leq r \leq R} \left( \frac{1}{\lambda_n} \log A_n(r) - 2r \right) \geq 0.$$

For  $0 \leq r < \varepsilon$ , the single term  $m = 0$  gives  $A_n(r) \geq 1$ , hence

$$\frac{1}{\lambda_n} \log A_n(r) \geq 0 \geq 2r - 2\varepsilon.$$

Letting  $\varepsilon \downarrow 0$  gives the required uniform lower bound on  $[0, R]$ .

For the upper bound, observe that the ratio of consecutive terms is

$$\frac{a_{n,m+1}(r)}{a_{n,m}(r)} = r^2 \frac{(n+m+1)^{2\beta}}{(m+1)^2}. \quad (11)$$

Choose  $M > 0$  so large that  $R^2 2^{2\beta} / M^2 < 1/4$ . If  $M\lambda_n \leq m \leq n$ , then, for all sufficiently large  $n$  and all  $r \leq R$ , the ratio in (11) is at most  $1/4$ . If  $m > n$ , then the same ratio is bounded by  $C_R m^{-2(1-\beta)}$ , hence is also at most  $1/4$  for all sufficiently large  $n$ . Thus, uniformly for  $0 \leq r \leq R$ , the tail  $m \geq M\lambda_n$  is geometrically dominated by its first term. It follows that, after replacing  $M$  by  $M + 1$  if necessary,

$$A_n(r) \leq C_R(1 + \lambda_n) \max_{0 \leq m \leq M\lambda_n} a_{n,m}(r) \quad (12)$$

for all large  $n$ .

It remains to estimate the maximum in (12). Let  $x = m/\lambda_n$ . If  $r = 0$ , then  $a_{n,m}(0) = 0$  for  $m > 0$  and  $a_{n,0}(0) = 1$ , so the desired upper bound is immediate. Assume therefore that  $r > 0$ . For  $1 \leq m \leq M\lambda_n$ , the estimate

$$\sum_{\ell=1}^m \log(n + \ell) = m \log n + O\left(\frac{m^2}{n}\right) = m \log n + o(\lambda_n)$$

is uniform, because  $m \leq M\lambda_n$  and  $\lambda_n/n \rightarrow 0$ . The elementary lower bound  $\log(m!) \geq m \log m - m$  then gives, uniformly for  $0 < r \leq R$  and  $1 \leq m \leq M\lambda_n$ ,

$$\frac{1}{\lambda_n} \log a_{n,m}(r) \leq 2x \log \frac{er}{x} + o(1), \quad (13)$$

with the usual convention that the right hand side is 0 when  $x = 0$ . The case  $m = 0$  is also covered, since  $a_{n,0}(r) = 1$ . For fixed  $r > 0$ , the function

$$x \mapsto 2x \log \frac{er}{x} \quad (x > 0)$$

has maximum  $2r$ , attained at  $x = r$ . Hence, for every  $\delta > 0$  and all sufficiently large  $n$ ,

$$a_{n,m}(r) \leq \exp\{(2r + \delta)\lambda_n\}$$

for all  $0 \leq r \leq R$  and  $0 \leq m \leq M\lambda_n$ . Combining this with (12) and using  $\log \lambda_n = o(\lambda_n)$  gives

$$\limsup_{n \rightarrow \infty} \sup_{0 \leq r \leq R} \left( \frac{1}{\lambda_n} \log A_n(r) - 2r \right) \leq 0.$$

Together with the lower bound, this proves (9). Formula (10) follows from (7).  $\square$

*Proof of Proposition 3.1.* Define

$$u_n(z) = \frac{1}{n^\beta} (\log K_n(z, z) - 2\beta \log(n!)).$$

By Lemma 3.2,  $u_n \rightarrow u := 2|z|$  locally uniformly, hence in  $L^1_{\text{loc}}$ . Taking distributional Laplacians and using (4), we get

$$n^{-\beta} \mu_n = \frac{1}{4\pi} \Delta u_n \longrightarrow \frac{1}{4\pi} \Delta(2|z|).$$

Since  $\Delta|z| = 1/|z|$  on  $\mathbb{C} \setminus \{0\}$  and this identity has no extra atom at the origin in the distributional sense, the limit is

$$\frac{1}{2\pi|z|} dA(z).$$

This proves the proposition.  $\square$

## 4 An Offord estimate for smooth zero statistics

The next estimate is the only probabilistic input beyond elementary Gaussian facts. It is a standard Offord-type estimate; the proof below is included because the constants and exact normalization are unimportant for the application.

**Lemma 4.1** (Logarithmic tail bound). *Let  $\zeta$  be a standard complex Gaussian random variable. There is a universal constant  $C$  such that, for every event  $E$  with probability  $p \in [0, 1]$ ,*

$$\mathbb{E}(|\log |\zeta|| \mathbf{1}_E) \leq Cp \log \frac{e}{p},$$

where the right hand side is interpreted as 0 when  $p = 0$ .

*Proof.* Set  $Y = |\log |\zeta||$ . For  $s \geq 0$ ,

$$\mathbb{P}\{Y > s\} = \mathbb{P}\{|\zeta| > e^s\} + \mathbb{P}\{|\zeta| < e^{-s}\} = e^{-e^{2s}} + 1 - e^{-e^{-2s}} \leq 2e^{-s}.$$

For any  $a \geq 0$  and any event  $E$  of probability  $p$ ,

$$\mathbb{E}(Y \mathbf{1}_E) \leq ap + \mathbb{E}(Y \mathbf{1}_{Y>a}) = ap + a\mathbb{P}\{Y > a\} + \int_a^\infty \mathbb{P}\{Y > s\} ds \leq ap + 2(a+1)e^{-a}.$$

Choosing  $a = \log(e/p)$  for  $0 < p \leq 1$  gives  $a \geq 1$  and  $e^{-a} = p/e$ , so the last expression is at most a universal constant times  $p \log(e/p)$ .  $\square$

**Lemma 4.2** (Offord estimate). *Let  $X$  be a Gaussian analytic function on a plane domain  $G$ , and suppose  $K_X(z, z) = \mathbb{E}|X(z)|^2$  is positive on  $G$ . Let  $Z_X$  be the zero counting measure of  $X$  and  $\mu_X = \mathbb{E}Z_X$ . Then there are universal constants  $C_1, C_2 > 0$  such that, for every  $\phi \in C_c^\infty(G)$  and every  $t > 0$ ,*

$$\mathbb{P}\left\{\left|\int_G \phi d(Z_X - \mu_X)\right| \geq t\right\} \leq C_1 \exp\left(-C_2 \frac{t}{\|\Delta\phi\|_{L^1(G)}}\right), \quad (14)$$

provided  $\Delta\phi \not\equiv 0$ . If  $\Delta\phi \equiv 0$ , the left hand side is zero.

*Proof.* Put

$$L_X(z) := \log |X(z)| - \frac{1}{2} \log K_X(z, z).$$

This is a random locally integrable function on  $G$ ; we use it only as a real-valued normalized logarithm, not as the logarithm of an analytic normalization. For each fixed  $z \in G$ ,  $L_X(z)$  has the same distribution as  $\log |\zeta|$ , where  $\zeta$  is a standard complex Gaussian. By Poincaré–Lelong and Lemma 2.2,

$$\int_G \phi d(Z_X - \mu_X) = \frac{1}{2\pi} \int_G \Delta\phi(z) L_X(z) dA(z) \quad (15)$$

in the distributional sense.

Let  $I$  denote the left hand side of (15), let  $E_t = \{|I| \geq t\}$ , and put  $p = \mathbb{P}(E_t)$ . If  $p = 0$  there is nothing to prove. Otherwise, using (15), Fubini's theorem, and Lemma 4.1, we get

$$\begin{aligned} tp &\leq \mathbb{E}(|I| \mathbf{1}_{E_t}) \\ &\leq \frac{1}{2\pi} \int_G |\Delta\phi(z)| \mathbb{E}(|L_X(z)| \mathbf{1}_{E_t}) dA(z) \\ &\leq C \|\Delta\phi\|_{L^1(G)} p \log \frac{e}{p}. \end{aligned}$$

After cancelling  $p$ , this implies

$$p \leq e \exp\left(-\frac{t}{C\|\Delta\phi\|_{L^1(G)}}\right),$$

which is (14) after adjusting the constants.  $\square$

## 5 Summable hole probabilities

**Proposition 5.1** (Hole probabilities). *Let  $U \subset \mathbb{C}$  be nonempty and open. Then there are constants  $c_U, C_U > 0$  such that, for all sufficiently large  $n$ ,*

$$\mathbb{P}\{F_n \text{ has no zero in } U\} \leq C_U e^{-c_U n^\beta}. \quad (16)$$

In particular,

$$\sum_{n=1}^{\infty} \mathbb{P}\{F_n \text{ has no zero in } U\} < \infty. \quad (17)$$

*Proof.* Choose a nonzero, nonnegative test function  $\phi \in C_c^\infty(U)$ . Since a compactly supported harmonic function on a plane domain is zero, this choice has  $\Delta\phi \not\equiv 0$ . Proposition 3.1 gives

$$\frac{1}{n^\beta} \int \phi d\mu_n \longrightarrow c_\phi := \int_{\mathbb{C}} \phi(z) \frac{dA(z)}{2\pi|z|} > 0.$$

Therefore, for all sufficiently large  $n$ ,

$$M_n := \int \phi d\mu_n \geq \frac{c_\phi}{2} n^\beta. \quad (18)$$

On the event that  $F_n$  has no zero in  $U$ , we have  $\int \phi dZ_n = 0$ , because  $\text{supp } \phi \subset U$ . Hence

$$\left| \int \phi d(Z_n - \mu_n) \right| = M_n.$$

Applying Lemma 4.2 to  $X = F_n$  and  $t = M_n$ , and then using (18), gives

$$\mathbb{P}\{F_n \text{ has no zero in } U\} \leq C_1 \exp\left(-C_2 \frac{M_n}{\|\Delta\phi\|_1}\right) \leq C_U e^{-c_U n^\beta}.$$

The series in (17) converges because  $\beta > 0$ .  $\square$

## 6 Almost sure proof of the main theorem

*Proof of Theorem 1.1.* Let  $\mathcal{B}$  be a countable base for the topology of  $\mathbb{C}$ , for instance all open disks with centers in  $\mathbb{Q} + i\mathbb{Q}$  and positive rational radii.

Fix  $B \in \mathcal{B}$ . By Proposition 5.1,

$$\sum_{n=1}^{\infty} \mathbb{P}\{F_n \text{ has no zero in } B\} < \infty.$$

The first Borel–Cantelli lemma, which does not require independence, implies that almost surely only finitely many of the events

$$\{F_n \text{ has no zero in } B\}$$

occur. Equivalently, almost surely there exists  $N(B)$  such that  $F_n = f^{(n)}$  has a zero in  $B$  for every  $n \geq N(B)$ .

Intersect this probability-one event over the countable family  $\mathcal{B}$ , and also intersect with the probability-one event from Lemma 2.1. For every outcome in the resulting event,  $f$  is transcendental entire and satisfies

$$\forall B \in \mathcal{B} \quad \exists N(B) \quad \forall n \geq N(B), \quad f^{(n)} \text{ has a zero in } B.$$

Fix such an outcome.

Let  $U \subset \mathbb{C}$  be nonempty and open. Choose  $B \in \mathcal{B}$  with  $B \subset U$ . Then  $f^{(n)}$  has a zero in  $B \subset U$  for every  $n \geq N(B)$ . This proves the cofinite property.

Finally, if  $n_1 < n_2 < \dots$  is any infinite increasing sequence and  $U \subset \mathbb{C}$  is nonempty and open, choose  $B \in \mathcal{B}$  with  $B \subset U$ . Since  $n_k \rightarrow \infty$ , some  $n_k$  is at least  $N(B)$ , and then  $f^{(n_k)}$  has a zero in  $U$ . Hence the union of zero sets along the subsequence meets every nonempty open set, and is dense in  $\mathbb{C}$ .  $\square$

## 7 Order of the examples

The construction gives examples of every prescribed order strictly larger than 2.

**Proposition 7.1.** *For the random function (1), almost surely*

$$\text{ord}(f) = \frac{1}{1-\beta} > 2.$$

Consequently, for every  $p > 2$ , choosing  $\beta = 1 - 1/p$  gives examples of order  $p$ .

*Proof.* Let

$$c_k = \frac{\xi_k}{(k!)^{1-\beta}}$$

be the Taylor coefficients of  $f$ . Fix  $\varepsilon > 0$ . Since

$$\sum_{k=1}^{\infty} \mathbb{P}\{|\xi_k| > e^{\varepsilon k \log k}\} < \infty$$

and

$$\sum_{k=1}^{\infty} \mathbb{P}\{|\xi_k| < e^{-\varepsilon k \log k}\} < \infty,$$

Borel–Cantelli gives, almost surely,

$$e^{-\varepsilon k \log k} \leq |\xi_k| \leq e^{\varepsilon k \log k}$$

for all sufficiently large  $k$ . Applying this to the countable sequence  $\varepsilon = 1/j$  and intersecting the resulting probability-one events gives

$$\log |\xi_k| = o(k \log k)$$

almost surely. Therefore

$$-\log |c_k| = (1 - \beta) \log(k!) + o(k \log k) = (1 - \beta)k \log k + o(k \log k).$$

The standard coefficient formula for the order of an entire function,

$$\text{ord}(f) = \left( \liminf_{k \rightarrow \infty} \frac{-\log |c_k|}{k \log k} \right)^{-1},$$

therefore gives  $\text{ord}(f) = 1/(1 - \beta)$ . Because  $\beta > 1/2$ , this order is greater than 2.  $\square$

## 8 Referee-style audit of the proof

We close by spelling out the main points at which hidden gaps could otherwise enter.

First, no independence between the events for different  $n$  is used. The proof uses only the first Borel–Cantelli lemma, so the strong dependence among the derivatives  $f^{(n)}$  is harmless.

Second, the Offord estimate is used only for smooth linear statistics. The passage from a hole event to a linear-statistic deviation is made by choosing  $\phi \geq 0$  with compact support inside the open set  $U$ . If  $F_n$  has no zero in  $U$ , then  $\int \phi dZ_n = 0$  exactly. In the proof of the Offord estimate, the normalized logarithm  $L_X(z) = \log |X(z)| - \frac{1}{2} \log K_X(z, z)$  is used pointwise; no analyticity of a normalized process is asserted or needed.

Third, the singular density  $1/|z|$  in Proposition 3.1 is locally integrable and has no atom at the origin. Therefore every nonzero nonnegative test function  $\phi \in C_c^\infty(U)$  has a positive limiting expected statistic, even when  $U$  is a small neighborhood of the origin.

Finally, the constant factor  $(n!)^{2\beta}$  in the covariance kernel is removed only inside a logarithm and contributes a spatial constant. Its Laplacian is zero, so it has no effect on the expected zero measure.

## References

- [1] T. F. Bloom, *Erdős Problem #906*, Erdős Problems website. <https://www.erdosproblems.com/906>
- [2] A. C. Offord, *The distribution of zeros of power series whose coefficients are independent random variables*, Indian Journal of Mathematics **9** (1967), 175–196.
- [3] A. C. Offord, *The distribution of the values of a random function in the unit disk*, Studia Mathematica **41** (1972), 71–106.
- [4] M. Sodin, *Zeros of Gaussian analytic functions*, Mathematical Research Letters **7** (2000), 371–381. Preprint: <https://arxiv.org/abs/math/0007030>.