EMPIRICAL DISTRIBUTION FUNCTIONS AND STRONG APPROXIMATION THEOREMS FOR DEPENDENT RANDOM VARIABLES. A PROBLEM OF BAKER IN PROBABILISTIC NUMBER THEORY

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Dedicated to Wolfgang M. Schmidt on his 60th birthday

Abstract. Let \( \mathcal{F} = \{ q_1, \ldots, q_t \} \) be a finite set of coprime integers and let \( \{ n_1, n_2, \ldots \} \) denote the multiplicative semigroup generated by \( \mathcal{F} \), and arranged in increasing order. Let \( D_N(\omega) \) denote the discrepancy of the sequence \( \{ n_k \omega \}_{k=1}^\infty \mod 1 \), \( \omega \in [0, 1) \). In this paper we solve a problem posed by R.C. Baker [3], by proving that for all \( \omega \) except on a set of Lebesgue measure 0

\[
\frac{1}{4} \leq \limsup_{N \to \infty} \frac{N D_N(\omega)}{\sqrt{N \log \log N}} \leq C.
\]

Here the constant \( C \) only depends on the total number of primes involved in the prime factorization of \( q_1, \ldots, q_t \). The lower bound is obtained from a strong approximation theorem for the partial sums of the sequence \( \{ \cos 2\pi n_k \omega \}_{k=1}^\infty \) by sums of independent standard normal random variables.

1. Introduction

Let \( \{ n_k, k \geq 1 \} \) be an increasing sequence of positive integers and let \(([0, 1), \mathcal{B}, \mathcal{P})\) denote the unit interval with Lebesgue measurability. Then

\[ \eta_k(\omega) := n_k \omega \mod 1 \]

are random variables defined on \(([0, 1), \mathcal{B}, \mathcal{P})\) which are uniformly distributed in the probabilistic sense, i.e.,

\[ P\{ \omega : \eta_k(\omega) \leq x \} = x, \quad 0 \leq x \leq 1, \quad k \geq 1. \]

If \( F_N \) denotes the empirical distribution function of the sequence \( \{ \eta_k \} \) at stage \( N \), i.e.,

\[ F_N(s) = F_N(s; \omega) := N^{-1} \# \{ k \leq N : \eta_k(\omega) \leq s \}, \quad 0 \leq s \leq 1, \]

then

\[ D_N(\omega) := \sup_{0 \leq s \leq 1} |F_N(s) - s| \]

denotes the discrepancy (or the Kolmogorov-Smirnov statistic) of the sequence \( \{ n_k \omega \}_{k=1}^\infty \mod 1 \). (For the basic definition and facts on uniform distribution...
mod 1, see e.g. Kuipers and Niederreiter [13].) Recall that \( D_N \) is a random variable, i.e., a measurable function from \( ([0, 1), \mathcal{B}, P) \to ([0, 1), \mathcal{B}, P) \).

R.C. Baker [2] proved that for given \( \epsilon > 0 \),

\[
D_N(\omega) \ll N^{-\frac{1}{4}}(\log N)^{\frac{1}{4}+\epsilon}
\]

for almost all \( \omega \), i.e., for all \( \omega \) except on a set of Lebesgue measure 0. The best result before Baker's paper had \( \frac{1}{2} \) instead of \( \frac{1}{4} \) in the exponent of \( \log N \) and was independently obtained by Cassels [7] and Erdős and Koksma [10]. This is of course without any further assumptions on \( \{n_k\} \).

In 1962, Erdős [9, p. 56] conjectured that for some \( c \left( \geq \frac{1}{2} \right) \)

\[
D_N(\omega) = o(N^{-\frac{1}{4}}(\log \log N)^c) \quad \text{a.e.}
\]

(1.1)

Apparently unaware of Erdős' conjecture R.C. Baker [2] conjectured that for any \( \epsilon > 0 \),

\[
D_N(\omega) = o(N^{-\frac{1}{4}}(\log N)^{\epsilon}) \quad \text{a.e.}
\]

For exponentially fast growing sequences \( \{n_k\} \), i.e., for sequences with

\[
n_{k+1}/n_k \geq q > 1, \quad k \geq 1,
\]

(1.2) is indeed true. In [19] a bounded law of the iterated logarithm was proved in the form

\[
\frac{1}{4} \leq \limsup_{N \to \infty} \frac{ND_N(\omega)}{\sqrt{N \log \log N}} \leq C(q) \quad \text{a.e.}
\]

(1.3)

Here \( C(q) \ll 1/\log q \), \( q \perp 1 \), only depends on \( q \). This established a longstanding conjecture of Erdős and Gál. By a different method it was shown in [20] that (1.3) continues to hold without the assumption that the \( n_k \)'s are integers. It is still unknown whether the limes superior equals a constant a.e. Also, the question raised in [19], whether or not the right side (1.3) continues to hold if (1.2) is replaced by

\[
n_{k+1}/n_k \geq 1 + k^{-q}, \quad q < \frac{1}{2},
\]

still remains open. The left side of (1.3) follows from a result of Berkes [4] and a well-known inequality of Koksma (see [13, p. 143]). \( q = \frac{1}{2} \) is the value in (1.4) where the finer probabilistic properties of the partial sums of the sequence \( \{\cos 2\pi n_k \omega\} \) begin to break down (see Berkes [5]). Thus there is some reason to believe that (1.3) is, in general, false for sequences satisfying (1.4) with \( q \geq \frac{1}{2} \).

From now on let \( \{n_k\}^\infty_{k=1} := \{q_1^{\alpha_1} \ldots q_r^{\alpha_r}, \alpha_i \geq 0 \text{ integer} \} \) and arranged in increasing order where \( \{q_1, \ldots, q_r\} \) is a finite set of coprime integers. Let \( r^* \) denote the total number of primes occurring in the prime factorization of \( q_1, \ldots, q_r \). Let \( D_N(\omega) \) denote the discrepancy of the sequence \( \{n_k \omega\}^N_{k=1} \mod 1 \).

The following theorem solves a problem of R.C. Baker [3], posed around 1979.
Theorem 1. There is a constant $C$, depending on $\tau^*$ only such that for almost all $\omega$

$$\frac{1}{4} \leq \limsup_{N \to \infty} \frac{ND_N(\omega)}{\sqrt{N \log \log N}} \leq C.$$ 

Over the years the analytic properties of the sequence $\{n_k \omega\}$ have attracted considerable attention in connection with Khintchine's conjecture. Marstrand [15] proved that for bounded and measurable $f$ with period 1

$$(1.5) \quad \frac{1}{N} \sum_{k \leq N} f(n_k \omega) \to \int_0^1 f(t) \, dt \quad \text{a.e.}$$

Recently (1.5) has been proved by Nair [18] to continue to hold under the weakened assumption $f \in L^1([0, 1])$, thereby answering another question raised by Baker [1].

The lower bound in Theorem 1 will follow from the following strong approximation theorem. Let $\Omega = [0, 1]^2$, be the unit square with Lebesgue measurability. Write $\omega = (\omega_1, \omega_2) \in \Omega$.

Theorem 2. There exists a sequence $\{Y_k(\omega_1, \omega_2)\}_{k=1}^{\infty}$ of independent standard Gaussian random variables defined on $[0, 1]^2$ such that for almost all $\omega = (\omega_1, \omega_2) \in [0, 1)^2$

$$2^k \sum_{k \leq N} \cos 2\pi n_k \omega_1 - \sum_{k \leq N} Y_k(\omega_1, \omega_2) \ll N^{1-\lambda}$$

for some $\lambda > 0$, depending on $\tau^*$ only.

By treating $e^{2\pi i \omega} = (\cos 2\pi \omega, \sin 2\pi \omega)$ as a two-dimensional random vector the same proof yields

Theorem 3. There exists a sequence $\{Y_k(\omega_1, \omega_2)\}_{k=1}^{\infty}$ of independent standard complex valued Gaussian random variables defined on $[0, 1)^2$ such that for almost all $\omega = (\omega_1, \omega_2) \in [0, 1)^2$

$$2^k \sum_{k \leq N} e^{2\pi i n_k \omega_1} - \sum_{k \leq N} Y_k(\omega_1, \omega_2) \ll N^{1-\lambda}$$

for some $\lambda > 0$, depending on $\tau^*$ only.

Remark. Since for almost all $\omega \in [0, 1)^2$

$$\limsup_{N \to \infty} \frac{\sum_{k \leq N} Y_k(\omega_1, \omega_2)}{\sqrt{2N \log \log N}} = 1$$

by the classical law of the iterated logarithm for 2-dimensional standard Gaussian random vectors, Theorem 3 implies that for almost all $\omega_1 \in [0, 1)$

$$\limsup_{N \to \infty} \frac{\sum_{k \leq N} e^{2\pi i n_k \omega_1}}{\sqrt{N \log \log N}} = 1$$

The inequality of Koksma with the constant improved by Niederreiter (see [13, p. 143]) implies the lower bound in Theorem 1. The upper bound in Theorem 1 will follow from the following stronger version.
Theorem 4. Let \( \alpha < 1/(4\tau) \). There is a constant \( C \), depending on \( \tau^* \) only, with the following property. For almost all \( \omega \in [0, 1) \) there is an \( N_0 = N_0(\omega, \alpha) \) such that for all \( N \geq N_0 \) and all \( s \) and \( t \) with \( 0 \leq s < t \leq 1 \)

\[
\max_{k \leq N} k|F_k(t) - F_k(s) - (t-s)| \leq C(t-s)^\alpha(N \log \log N)^{1/2} + N^{1/2}.
\]

As an immediate consequence we obtain the following corollary. For a proof, see [20, pp. 325–326].

Corollary. Define

\[
f_N(t) = N(F_N(t) - t)(2N \log \log N)^{-1}, \quad 0 < t < 1, \quad N \geq 1.
\]

Then the sequence \( \{f_N(t), N \geq 1\} \) is with probability 1 relatively compact in \( D[0, 1] \) endowed with the supremum norm.

It follows from a theorem of Tijdeman [24] that the sequence \( \{n_k\} \) satisfies a growth condition (1.4) with an effectively computable constant \( q > 0 \). It is easy to see that "on average" (1.4) holds with \( q = 1 - 1/\tau \). Since in general we can assume \( \tau \geq 2 \) (the case \( \tau = 1 \) is taken care of by (1.2)) it follows by the remarks preceding Theorem 1 that probability methods alone will not suffice for the proof of our results. In fact, the proofs will employ a mix of martingale inequalities, strong approximation theorems for martingales, inequalities from the theory of uniform distribution mod 1, theorems on the finiteness of the number of solutions of \( S \)-unit equations, and Tijdeman's theorem.

2. Preliminary results

For fixed \( s \) and \( t \) with \( 0 < s < t \leq 1 \) we write

\[
L = [s, t), \quad x_\nu = x_\nu(s, t; \omega) = 1_L(n_\nu \omega) - (t-s).
\]

Here \( 1_L(\cdot) \) denotes the indicator function of \( L \), extended with period 1. We also write for any function \( f : [0, 1]^2 \to \mathbb{R} \)

\[
\|f\| = \sup_{0 \leq s < t \leq 1} |f(s, t)|.
\]

Lemma 2.1. Let \( \mathcal{N} \) be a finite set of positive integers with \( \text{card} \mathcal{N} = N \), say. Then, as \( N \to \infty \)

\[
E \sup_{T} \left\| \sum_{\nu \in \mathcal{N}, \nu \leq T} x_\nu \right\|^6 \ll N^3 \log^6 N
\]

where the constant implied by \( \ll \) only depends on \( \tau^* \).

Proof. Note that \( \| \cdot \| \) in (2.3) is a random variable since the supremum over \( s \) and \( t \) needs to be extended only over a countable dense subset of \( [0, 1]^2 \).

Let \( \mathcal{M} \subset \mathcal{N} \) be an arbitrary subset of \( \mathcal{N} \) with \( \text{card} \mathcal{M} = m \), say. Then by the Erdős-Turán inequality [13, p. 112] we have for each \( R \geq 1 \) and each \( \omega \in [0, 1) \)

\[
\left\| \sum_{\nu \in \mathcal{M}} x_\nu \right\| \leq \frac{6m}{R} + 2 \sum_{r=1}^{R} \frac{1}{r} \left| \sum_{\nu \in \mathcal{M}} e(rn_\nu \omega) \right|. 
\]
Here we use the standard notation $e(x) = e^{2\pi ix}$. Setting $R = m$ we obtain

$$E \left\{ \max_{M \leq x \leq N} \left\| \sum_{\nu \in \mathcal{M}} x_\nu \right\| \right\} \leq 126 + 4^6 E \left\{ \left( \max_{1 \leq r \leq N} \left\| \sum_{\nu \in \mathcal{M}} e(rn_\nu) \right\| \right)^6 \right\}.$$ 

In order to apply one of Billingsley's maximal inequalities [6, p. 102, Problem 5] or better still [17, Theorem 3.1] (see Lemma A.1 in the appendix) we shall show that uniformly in $r = 1, 2, \ldots$ and uniformly in $\mathcal{M} \subset \mathcal{N}$

$$\int_0^1 \left| \sum_{\nu \in \mathcal{M}} e(rn_\nu) \right|^6 d\omega \leq A m^3$$

for some constant $A$, depending on $\tau^*$ only. The integral in (2.4) is bounded by the number of solutions of the Diophantine equation

$$(2.5;6) \sum_{i=1}^{6} \pm n_{\nu_i} = 0$$

with $\nu_i \in \mathcal{M}$, $1 \leq i \leq 6$. Recall that each $n_{\nu_i}$ is a product of at most $\tau^*$ prime powers. Hence by the van der Porten-Schlickewei-Evertse theorem (see [22], [11], and [12, pp. 116–117]) there are only finitely many solutions of (2.5;6) with

$$(2.6;6) \gcd(n_{\nu_1}, n_{\nu_2}, n_{\nu_3}, n_{\nu_4}, n_{\nu_5}, n_{\nu_6}) = 1$$

provided that

$$(2.7) \text{no proper subsum of (2.5;6) vanishes.}$$

Denote this finite number of solutions of (2.5;6) subject to (2.6;6) and (2.7) by $B_6$ and denote the following auxiliary argument $\mathcal{A}_6$:

Let $k \in \mathcal{M}$ and let $(n_{\nu_i}, 1 \leq i \leq 6)$ be one of these $B_6$ solutions of (2.5;6). If $n_k/n_{\nu_i}$ happens to be an integer then the numbers $n_{\nu_i} n_k/n_{\nu_i}$ ($1 \leq i \leq 6$) could possibly yield a solution of (2.5;6) (provided, of course, that all these six numbers correspond to some $n_{k_i}, k_i \in \mathcal{M}$). Thus the number of solutions of (2.5;6) subject to (2.7) does not exceed $B_6 \cdot m$. This is the end of the argument $\mathcal{A}_6$.

Since $n_\nu > 0$ a proper subsum of (2.5;6) can vanish only if either both

$$(2.5;3) n_{\nu_1} \pm n_{\nu_2} \pm n_{\nu_3} = 0 \quad \text{and} \quad n_{\nu_4} \pm n_{\nu_5} \pm n_{\nu_6} = 0$$

or if both

$$(2.5;4) n_{\nu_1} \pm n_{\nu_2} \pm n_{\nu_3} \pm n_{\nu_4} = 0$$

and

$$(2.5;2) n_{\nu_1} \pm n_{\nu_6} = 0.$$

By two arguments $\mathcal{A}_3$ (with a self-evident explanation) we see that the number of solutions of (2.5;6) generated by (2.5;3) does not exceed $(B_3 m)^2$. By an argument $\mathcal{A}_4$ we see that the number of solutions of (2.5;4) does not exceed $B_4 m$ provided that no proper subsum of (2.5;4) vanishes. But this can only happen if both $n_{\nu_1} - n_{\nu_2} = 0$ and $n_{\nu_3} - n_{\nu_4} = 0$ and then the number of solutions
does not exceed \( m^2 \). Collecting all these estimates and noting that (2.5;2) has at most \( m \) solutions we conclude that the total number of solutions of (2.5;6) with \( \nu_i \in \mathcal{M} \), \( 1 \leq i \leq 6 \), does not exceed

\[
6!(m^3 + B_3^2 m^2 + B_4 m^2 + B_6 m).
\]

Thus (2.4) holds with \( A = 6!(1 + B_3^2 + B_4 + B_6) \).

Hence by Lemma A.1 with \( \alpha = 3 \), and \( \gamma = 6 \) we obtain for some constant \( A^* \)

\[
E \sup \left| \sum_{\nu \in \mathcal{M}, \nu \leq T} e(\nu_\nu \cdot) \right|^6 \leq A^* N^3.
\]

Thus by Minkowski's inequality the left-hand side of (2.3) does not exceed

\[
\ll 1 + N^3 \left( \sum_{r \leq N} \frac{1}{r} \right)^6 \ll N^3 \log^6 N.
\]

**Lemma 2.2.** Let \( \mathcal{M} \) be a finite set of positive integers with \( \text{card} \mathcal{M} = N \), say, and let \( \varepsilon > 0 \). Then for \( 0 < s < t < 1 \)

\[
E \left( \sum_{\nu \in \mathcal{M}} x_\nu(s, t) \right)^2 \ll (t - s)^{1-\varepsilon} N,
\]

where the constant implied by \( \ll \) only depends on \( \tau^* \) and \( \varepsilon \). Moreover,

\[
E \left( \sum_{\nu \in \mathcal{M}} x_\nu(s, t) \right)^4 \ll (t - s)^{\frac{1}{2}(1-\varepsilon)} N^2 (\log N)^3.
\]

**Proof.** The integral in (2.9) plays an important role in the metric theory of uniform distribution mod 1. Perhaps the quickest way to prove this lemma is to expand \( 1_L(\cdot) - (t - s) \) into a Fourier series

\[
1_L(\omega) - (t - s) = \sum_{h \neq 0} c_h e(h \omega)
\]

where

\[
|c_h| \leq \frac{1}{\pi |h|}, \quad h \neq 0.
\]

Let \( || \cdot ||_2 = \left( \int_0^1 | \cdot |^2 d\omega \right)^{\frac{1}{2}} \) denote the \( L^2 \)-norm on \([0, 1]\) and let \( 1{\cdot, \cdot} = \delta \).
denote the Kronecker symbol. Then by Minkowski’s inequality

\[ E^{\frac{1}{2}} \left\{ \left( \sum_{\mu \in \mathbb{N}} x_\mu(s, t) \right)^2 \right\} \leq \sum_{u \geq 0} \left( \sum_{\mu, \nu \in \mathbb{N}} c_\mu \sum_{\nu \in \mathbb{N}} \epsilon(h n_\nu) \right) \left( \sum_{\mu, \nu \in \mathbb{N}} c_\mu c_\nu \{h n_\mu, k n_\nu\} \right)^{\frac{1}{2}} \]

\[ \leq \sum_{u \geq 0} \left( \sum_{\mu, \nu \in \mathbb{N}} \sum_{2^u \leq |h|, |k| < 2^{u+1}} |c_\mu|^2 \sum_{\nu \in \mathbb{N}} \sum_{2^u \leq |k| < 2^{u+1}} 1\{h n_\mu, k n_\nu\} \right)^{\frac{1}{2}} \]

For fixed \( u, \nu, \) and \( h \) the inner sum is easily seen to be \( \ll u^{T^*} \). Indeed, this sum equals the number of solutions of the Diophantine equation \( h n_\mu = k n_\nu \). If \( n_\mu = p_1^{\delta_1} \ldots p_r^{\delta_r} \) and \( h = p_1^{\gamma_1} \ldots p_r^{\gamma_r} M \), where \( M \) is not divisible by \( p_1, \ldots, p_r \) then \( k = p_1^{\gamma_1} \ldots p_r^{\gamma_r} M \) for some nonnegative integers \( \delta_1, \ldots, \delta_r \).

But \( 2^u \leq k < 2^{u+1} \) implies

\[ p_1^{\delta_1} \ldots p_r^{\delta_r} \leq 2^{u+1} \text{ or } \delta_1 \log p_1 + \ldots + \delta_r \log p_r \leq (u + 1) \log 2. \]

The number of lattice points \( (\delta_1, \ldots, \delta_r) \), \( \delta_i \geq 0 \) is about the volume of the corresponding tetrahedron and this is \( \ll u^{T^*} \). Thus the number of possible \( k \)'s is \( \ll u^{T^*} \). This proves the claim.

Hence we can continue the chain of inequalities and obtain, by (2.12), Cauchy’s and Hölder’s inequality

\[ \leq 2/\pi N^{\frac{1}{2}} \sum_{u \geq 0} u^{T^*} \left( \sum_{2^u \leq h < 2^{u+1}} |c_h|^2 \right)^{\frac{1}{2}} \]

\[ \ll N^{\frac{1}{2}} \sum_{u \geq 0} u^{T^*} 2^{-u} \left( \sum_{2^u \leq h < 2^{u+1}} |c_h|^{2-\varepsilon} \right)^{\frac{1}{2}} \]

\[ \ll N^{\frac{1}{2}} \left( \sum_{u \geq 0} u^{T^*} 2^{-2u} \right)^{\frac{1}{2}} \left( \sum_{u \geq 0} \sum_{2^u \leq h < 2^{u+1}} |c_h|^{2-\varepsilon} \right)^{\frac{1}{2}} \]

(2.13)

\[ \ll N^{\frac{1}{2}} \left( \sum_{h \neq 0} |c_h|^{2-\varepsilon} \right)^{\frac{1}{2}} \ll N^{\frac{1}{2}} \left( \sum_{h \neq 0} |c_h|^{2-3\varepsilon} h^{-2\varepsilon} \right)^{\frac{1}{2}} \]

\[ \ll N^{\frac{1}{2}} \left( \sum_{h \neq 0} |c_h|^{2-\varepsilon} \right)^{\frac{1}{2}} \ll N^{\frac{1}{2}} \left( \sum_{h \neq 0} |c_h|^{2-3\varepsilon} h^{-2\varepsilon} \right)^{\frac{1}{2}} \]

\[ \ll N^{\frac{1}{2}} \|1_L(\cdot) - (t-s)\|_{L^2}^{\frac{1}{2} - \frac{\varepsilon}{2}} \ll N^{\frac{1}{2}} (t-s)^{\frac{1}{2} - \frac{\varepsilon}{2}}. \]
This proves (2.9). Relation (2.10) follows by interpolation. Indeed, by [14, p. 156] \( \varphi(r) := \log E[X^r] \) is a convex function in \( r \), for any random variable \( X \). Thus by (2.9) and Lemma 2.1
\[
\varphi(4) = \varphi(\frac{1}{2} + \frac{1}{2}6) \leq \frac{1}{2}(\varphi(2) + \varphi(6))
\]
\[
\ll \frac{1}{2}((1 - \varepsilon) \log(t - s) + 4 \log N + 6 \log \log N).
\]
This proves (2.10).

**Lemma 2.3.** Let \( \mathcal{N} \) be as in Lemma 2.2 and let
\[
\varphi_m(\omega) = \sum_{|h| \geq m} c_h e(h\omega)
\]
where the \( c_h \) are defined by (2.11). Then for \( \lambda > 0 \)
\[
E \left| \sum_{h \in \mathcal{N}} \varphi_m(n_k \omega) \right|^2 \ll Nm^{-1}(t - s)^{\frac{1}{2} - \lambda}.
\]

**Proof.** Let \( u_0 \) be the largest integer with \( 2^{u_0} \leq m \). We follow the proof of Lemma 2.2 until (2.13) and obtain
\[
E \left\{ \left| \sum_{h \in \mathcal{N}} \varphi_m(n_k \omega) \right|^2 \right\} \ll N^{\frac{1}{2}} \left( \sum_{u \geq u_0} u^{2-2\varepsilon} \right)^{\frac{1}{2}} \left( \sum_{u \geq u_0} \sum_{2^u \leq |h| < 2^{u+1}} |c_h|^{2-\varepsilon} \right)^{\frac{1}{2}}
\]
\[
\ll N^{\frac{1}{4}} 2^{-u_0} u_0^{\frac{5}{2}} (t - s)^{\frac{1}{2} - \frac{3}{4}}
\]
by the remainder of the proof of Lemma 2.2. We let \( \varepsilon > \frac{1}{2} \) and obtain the bound
\[
\ll N^{\frac{1}{4}} m^{-\frac{1}{4}} (t - s)^{\frac{1}{4} - \lambda} \quad \text{for} \lambda > 0.
\]

3. **Proof of Theorem 4**

For \( j = 0, 1, 2, \ldots \) we let \( H_j \) denote the set of all indices \( k \) such that
\[
(3.1) \quad \exp(j^2) < n_k < \exp((j + 1)^2 - j^2)
\]
and we let \( I_j \) denote the set of all indices \( k \) such that
\[
(3.2) \quad \exp((j + 1)^2 - j^2) < n_k < \exp((j + 1)^2).
\]
Thus if \( n_k = q_1^{\alpha_1} \cdots q_t^{\alpha_t} \) then \( k \in H_j \) iff
\[
(3.2) \quad j^2 < \alpha_1 \log q_1 + \cdots + \alpha_t \log q_t < (j + 1)^2 - j^2.
\]

Let \( h_j \) denote the largest member of \( H_j \). Estimating the number of lattice points \( (\alpha_1, \ldots, \alpha_t) \) subject to (3.2) we obtain as \( j \to \infty \)
\[
(3.3) \quad h_j \sim \frac{1}{\log q_1 \cdots \log q_t} \frac{1}{\tau!} ((j + 1)^2 - j^2)^\tau.
\]
and so
\[
(3.4) \quad \text{card } H_j \gg j^{2\tau - 1}, \quad \text{card } I_j \ll j^{2\tau - 7/4}.
\]

The blocks \( I_j \) have been introduced to provide the proper spacing between the large blocks \( H_j \). This is standard technique in the theory of weak dependence. Moreover, the contributions of the blocks \( I_j \) is negligible. Recall that \( \| \cdot \| \) has been defined in (2.2).
Lemma 3.1. With probability 1 we have as \( n \to \infty \)

\[
\left\| \sum_{j \leq n} \sum_{u \in I_j} x_u \right\| \ll h_n^{\frac{1}{2} - \frac{1}{3}a}.
\]

Proof. We apply Lemma 2.1 with \( \mathcal{N} = \bigcup_{j \leq n} I_j \). Then by (3.4)

\[
E \left\| \sum_{j \leq n} \sum_{u \in I_j} x_u \right\|^6 \ll \left( \sum_{j \leq n} \text{card } I_j \right)^3 \left( \log \left( \sum_{j \leq n} \text{card } I_j \right) \right)^6
\]

\[
\ll \left( \sum_{j \leq n} j^{2t-7/4} \right)^3 \left( \log n \right)^6 \ll n^{6t-2}.
\]

Thus by Markov’s inequality we have

\[
P \left( \left\| \sum_{j \leq n} \sum_{u \in I_j} x_u \right\| \geq h_n^{\frac{1}{2} - \frac{1}{3}a} \right) \ll h_n^{-3+\frac{1}{3}a} n^{6t-2} \ll n^{-2+3a}.\]

The lemma follows from the convergence part of the Borel Cantelli lemma.

Having properly disposed of the blocks \( I_j \) we can now concentrate on the blocks \( H_j \). Let \( r_k \) denote the largest integer \( r \) with

\[
2^r \leq n_k k^{12}
\]

and let \( \mathcal{H}_k \) be the \( \sigma \)-field generated by the intervals

\[
U_{\nu k} = [\nu 2^{-r_k}, (\nu + 1)2^{-r_k}), \quad 0 \leq \nu < 2^{r_k}.
\]

We set

\[
\mathcal{F}_j = \mathcal{H}_j, \quad \xi_{\nu} = E(x_{\nu} | \mathcal{F}_j), \quad \nu \in H_j
\]

and

\[
w_j = \sum_{\nu \in H_j} x_{\nu}, \quad y_j = E(w_j | \mathcal{F}_j) = \sum_{\nu \in H_j} \xi_{\nu}.
\]

The proof of Theorem 4 follows the pattern of [20, Theorem 4.1] except for the details. We first show that \( \{y_j\} \) is close to \( \{w_j\} \), uniformly in \( 0 \leq s < t \leq 1 \). Second, we approximate \( \{y_j\} \) by a martingale difference \( \{Y_j\} \). After truncating \( Y_j \) and recentering at conditional expectations we apply an exponential inequality. This will finally yield the desired upper bound.

Lemma 3.2. With probability 1

\[
\sum_{j=1}^{\infty} \| y_j - w_j \| < \infty.
\]

Proof. This will follow from the Beppo Levi theorem and

\[
\sum E\| y_j - w_j \| < \infty.
\]

In order to show (3.9) we recall that by [20, Lemma 4.2.3] for each \( 0 \leq s < t \leq 1 \)

\[
E|\xi_{\nu} - x_{\nu}|^2 \ll \nu^{-12}
\]
with an absolute constant implied by \( \ll \). We shall use (3.10) to show that in fact

\[
E\|\xi_{\nu} - \chi_{\nu}\| \ll \nu^{-3}
\]

and this will yield (3.9) and thus the lemma.

Let \( m \) and \( M \) be integers with \( 0 < m < M \) which will be chosen suitably later. We write \( s \) and \( t \) in binary expansion

\[
s = \sum_{i=1}^{\infty} \sigma_i 2^{-i}, \quad \sigma_i = 0, 1,
\]

\[
t = \sum_{i=1}^{\infty} \tau_i 2^{-i}, \quad \tau_i = 0, 1.
\]

Then

\[
s = a 2^{-m} + \sum_{i=m+1}^{M} \sigma_i 2^{-i} + \theta_1 2^{-M},
\]

\[
t = b 2^{-m} + \sum_{i=m+1}^{M} \tau_i 2^{-i} + \theta_2 2^{-M}
\]

where \( a \) and \( b \) are integers with \( 0 < a, b < 2^m \) and \( 0 < \theta_1, \theta_2 \leq 1 \). We also write for \( s < t \)

\[
Z = Z(s, t) = Z_{\nu}(s, t) = |\xi_{\nu}(s, t) - \chi_{\nu}(s, t)|
\]

and observe that for \( s < r < t \)

\[
Z(s, t) \leq Z(s, r) + Z(r, t),
\]

\[
Z(r, t) \leq Z(s, t) + Z(s, r).
\]

We set

\[
m = 0, \quad M = \left[ \frac{3}{\log 2} \log \nu \right] + 1.
\]

We apply (3.14) repeatedly and obtain in view of (3.12)

\[
Z(s, t) \leq \sum_{i=1}^{M} Z(a_i 2^{-i}, (a_i + 1) 2^{-i}) + \sum_{i=1}^{M} Z(b_i 2^{-i}, (b_i + 1) 2^{-i})
\]

\[
+ Z(a_{M+1} 2^{-M}, (a_{M+1} + 1) 2^{-M})
\]

\[
+ Z(b_{M+1} 2^{-M}, (b_{M+1} + 1) 2^{-M}) + 2^{-M+1}
\]

where \( a_i, b_i \ (1 \leq i \leq M+1) \) are integers with \( 0 \leq a_i, b_i < 2^i \ (1 \leq i \leq M+1) \).

The last term is explained by the fact that for \( 0 \leq h < 2^M \) and \( 0 \leq \theta \leq 1 \)

\[
Z(h 2^{-M}, (h + \theta) 2^{-M}) \leq Z(h 2^{-M}, (h + 1) 2^{-M}) + 2^{-M}
\]

by an application of (3.14). Thus by (3.10) and (3.15) we obtain

\[
E\|Z\| \leq 4 \sum_{i=1}^{M} 2^{i-1} \sum_{a=0}^{2^i-1} E\{Z(a 2^{-i}, (a + 1) 2^{-i})\} + 2^{-M+1} \leq 2^M \nu^{-12} \ll \nu^{-3}.
\]

This proves (3.11) and thus the lemma.
Lemma 3.3. The random variables $y_j$ can be represented in the form

$$y_j = Y_j + v_j$$

where $(Y_j, \mathcal{F}_j, j \geq 1)$ is a martingale difference sequence and

$$v_j \ll (t-s) \exp \left( -\frac{1}{2} j^{\frac{1}{4}} \right) \quad \text{a.s.}$$

with an absolute constant implied by $\ll$.

Proof. We put $Y_j = y_j - E(y_j | \mathcal{F}_{j-1})$. By (3.8) $Y_j$ is $\mathcal{F}_j$-measurable and $E(Y_j | \mathcal{F}_{j-1}) = 0$. Thus $(Y_j, \mathcal{F}_j, j \geq 1)$ is a martingale difference sequence and

$$v_j := y_j - Y_j = E(y_j | \mathcal{F}_{j-1})$$

needs to be estimated. Write $r = r_{h_{j-1}}$ and for $0 \leq k < 2^r$ let

$$U = [k2^{-r}, (k+1)2^{-r}).$$

By [20, Lemma 4.2.1]

$$\left| \int_U x_\nu(\omega) \, d\omega \right| \leq 4(t-s)n^{\nu-1}. $$

Hence by the definition of a conditional expectation

$$E(x_\nu | \mathcal{F}_{j-1}) = \sum_{k=0}^{2^r-1} 1_u(\cdot)2^r \int_U x_\nu(\omega) \, d\omega \ll 2^r(t-s)n^{\nu-1}. $$

Thus by (3.3), (3.5), and (3.8) we have for $\nu \in H_j$

$$v_j = E(y_j | \mathcal{F}_{j-1}) = \sum_{\nu \in H_j} E(E(x_\nu | \mathcal{F}_j) | \mathcal{F}_{j-1})$$

$$= \sum_{\nu \in H_j} E(x_\nu | \mathcal{F}_{j-1}) \ll n_{h_{j-1}} h_{j-1}^{2^r} h_{j}^{(t-s) \exp(-j^2)}$$

$$\ll (t-s) j^{2^r} \exp(j^2 - (j-1)^{\frac{1}{4}} - j^2)$$

$$\ll (t-s) \exp \left( -\frac{1}{2} j^{\frac{1}{4}} \right) \quad \text{a.s.}$$

Lemma 3.4 (Main lemma). Let $C$ be the constant implied by $\ll$ in Lemma 2.2. Then

$$\sum_{j \leq n} E(w_j^2 | \mathcal{F}_{j-1}) \leq 2C(t-s)^{\frac{1}{4} - \epsilon} h_n$$

except on a set of measure $\ll n^{-1/32}$.

This will follow from (2.9) and the following two lemmas.

Lemma 3.5. Let $0 < \epsilon < \frac{1}{12}$. Then as $n \to \infty$

$$P \left( \max_{k \leq n} \left| \sum_{j \leq k} (w_j^2 - E(w_j^2 | \mathcal{F}_{j-1})) \right| \geq (t-s)^{\frac{1}{4} - \epsilon} h_n n^{-\epsilon} \right) \ll n^{-\frac{1}{2}}.$$

Proof. By (3.10) and Minkowski's inequality

$$E|y_j - w_j|^2 \leq \left( \sum_{\nu \in H_j} E^\frac{1}{4} |\xi_\nu - x_\nu|^2 \right)^2 \ll h_j^{-10}.$$
Thus by (2.9) and since \(E y_j^2 \leq E w_j^2\) by the conditional version of Jensen’s inequality
\[
E|y_j^2 - w_j^2| \leq (E|y_j - w_j|^2 E w_j^2)^{\frac{1}{2}}
\]
\[
\leq h_j^{-5} (\text{card } H_j)^{\frac{1}{4}} (t - s)^{\frac{1}{4} - \varepsilon} \leq (t - s)^{\frac{1}{4} - \varepsilon} j^{-8\varepsilon}
\]
and
\[
(E|y_j - w_j|^2 E w_j^2)^{\frac{1}{2}} \leq (E|y_j - w_j|^2 E w_j^2)^{\frac{1}{2}} \leq (t - s)^{\frac{1}{4} - \varepsilon} j^{-8\varepsilon}.
\]
Hence
\[
E|E(y_j^2 | \mathcal{F}_{j-1}) - E(w_j^2 | \mathcal{F}_{j-1})| \leq E|y_j^2 - w_j^2| \leq (t - s)^{\frac{1}{4} - \varepsilon} j^{-8\varepsilon}.
\]

\[
(3.18)
\]
\[
E|E(y_j^2 | \mathcal{F}_{j-1}) - E(w_j^2 | \mathcal{F}_{j-1})| \leq E|y_j^2 - w_j^2| \leq (t - s)^{\frac{1}{4} - \varepsilon} j^{-8\varepsilon}.
\]

Hence
\[
P \left( \sum_{j \leq n} |y_j^2 - E(y_j^2 | \mathcal{F}_{j-1}) - (w_j^2 - E(w_j^2 | \mathcal{F}_{j-1}))| \geq (t - s)^{\frac{1}{4} - \varepsilon} h_n n^{-\varepsilon} \right)
\]
\[
\leq h_n^{-1} n^{-\varepsilon} \sum_{j \leq n} j^{-8\varepsilon} \leq n^{-2\varepsilon + \varepsilon},
\]
and so for the proof of the lemma it is enough to show that
\[
P \left( \max_{k \leq n} \left| \sum_{j \leq k} (y_j^2 - E(y_j^2 | \mathcal{F}_{j-1})) \right| \geq (t - s)^{\frac{1}{4} - \varepsilon} h_n n^{-\varepsilon} \right) \leq n^{-\frac{1}{4}}.
\]

The purpose of this exercise was to make sure that we are dealing with a martingale difference sequence, namely with \(\{y_j^2 - E(y_j^2 | \mathcal{F}_{j-1}), \mathcal{F}_j, j \geq 1\}\) which requires \(\mathcal{F}_j\)-measurability of the \(j\)th difference. By Doob’s maximal inequality for martingales [8, p. 314] the probability in question does not exceed
\[
(t - s)^{\frac{1}{4} + 2\varepsilon} h_n^{-2} n^{2\varepsilon} E \left( \sum_{j \leq n} (y_j^2 - E(y_j^2 | \mathcal{F}_{j-1})) \right)^2
\]
\[
\leq (t - s)^{\frac{1}{4} + 2\varepsilon} h_n^{-2} n^{2\varepsilon} \sum_{j \leq n} E y_j^4
\]
\[
\leq (t - s)^{\frac{1}{4} + 2\varepsilon} h_n^{-2} n^{2\varepsilon} \sum_{j \leq n} E w_j^4
\]
\[
\leq h_n^{-2} n^{2\varepsilon} \sum_{j \leq n} (\text{card } H_j)^2 (\log j)^3 \leq n^{-1 + 3\varepsilon}
\]
using the orthogonality of martingale differences, Jensen’s inequality, and (2.10), (3.3), and (3.4).

**Lemma 3.6.** Let \(0 < \varepsilon < \frac{1}{32}\). Then as \(n \to \infty\)
\[
P \left( \max_{k \leq n} \left| \sum_{j \leq k} (w_j^2 - E w_j^2) \right| \geq (t - s)^{\frac{1}{4} - \varepsilon} h_n n^{-\varepsilon} \right) \leq n^{-\frac{1}{4}}.
\]

**Proof.** We set
\[
m_j = \lfloor j^{\frac{1}{4}} \rfloor,
\]
\[
w_j^* = \sum_{\nu \in H_j} \varphi_m(n_\nu \omega),
\]
\[
(3.20)
\]

The purpose of this exercise was to make sure that we are dealing with a martingale difference sequence, namely with \(\{y_j^2 - E(y_j^2 | \mathcal{F}_{j-1}), \mathcal{F}_j, j \geq 1\}\) which requires \(\mathcal{F}_j\)-measurability of the \(j\)th difference. By Doob’s maximal inequality for martingales [8, p. 314] the probability in question does not exceed
\[
(t - s)^{\frac{1}{4} + 2\varepsilon} h_n^{-2} n^{2\varepsilon} E \left( \sum_{j \leq n} (y_j^2 - E(y_j^2 | \mathcal{F}_{j-1})) \right)^2
\]
\[
\leq (t - s)^{\frac{1}{4} + 2\varepsilon} h_n^{-2} n^{2\varepsilon} \sum_{j \leq n} E y_j^4
\]
\[
\leq (t - s)^{\frac{1}{4} + 2\varepsilon} h_n^{-2} n^{2\varepsilon} \sum_{j \leq n} E w_j^4
\]
\[
\leq h_n^{-2} n^{2\varepsilon} \sum_{j \leq n} (\text{card } H_j)^2 (\log j)^3 \leq n^{-1 + 3\varepsilon}
\]
using the orthogonality of martingale differences, Jensen’s inequality, and (2.10), (3.3), and (3.4).

**Lemma 3.6.** Let \(0 < \varepsilon < \frac{1}{32}\). Then as \(n \to \infty\)
\[
P \left( \max_{k \leq n} \left| \sum_{j \leq k} (w_j^2 - E w_j^2) \right| \geq (t - s)^{\frac{1}{4} - \varepsilon} h_n n^{-\varepsilon} \right) \leq n^{-\frac{1}{4}}.
\]

**Proof.** We set
\[
m_j = \lfloor j^{\frac{1}{4}} \rfloor,
\]
\[
w_j^* = \sum_{\nu \in H_j} \varphi_m(n_\nu \omega),
\]
\[
(3.20)
\]
and
\[ u_j = w_j - w_j^*. \]
Then
\[ \left| (w_j^2 - Ew_j^2) - (w_j^* - Eu_j^*) \right| \leq |w_j|^2 + E|w_j|^2 + 2|w_j^* u_j| + 2(E|w_j|^2 Eu_j^*)^{1/2}. \]
Now by Lemma 2.3
\[ E|w_j|^2 \ll \text{card } H_j m_j^{-1} (t-s)^{1-\varepsilon}. \]
Thus by (3.19), (3.3), and (3.4)
\[ \sum_{j \leq n} E|w_j|^2 \ll (t-s)^{1-\varepsilon} \sum_{j \leq n} m_j^{-1} \text{card } H_j \ll (t-s)^{1-\varepsilon} n^{-1} h_n. \]
Consequently, by Markov's inequality
\[ P \left( \left\| \sum_{j \leq n} w_j^* \right\|^2 \geq (t-s)^{1-\varepsilon} h_n n^{-\varepsilon} \right) \ll n^{-1}. \]
Similarly, since by (2.9)
\[ E|u_j|^2 \leq 2E|w_j|^2 + 2E|w_j^*|^2 \ll \text{card } H_j (t-s)^{1-\varepsilon} \]
we have
\[ \sum_{j \leq n} (E|u_j|^2 E|w_j|^2)^{1/2} \ll (t-s)^{1-\varepsilon} \sum_{j \leq n} m_j^{-1} \text{card } H_j \ll (t-s)^{1-\varepsilon} n^{-1/2} h_n. \]
Consequently
\[ P \left( \left\| \sum_{j \leq n} u_j w_j^* \right\| \geq (t-s)^{1-\varepsilon} h_n n^{-\varepsilon} \right) \ll n^{-1/2}. \]
Thus in order to complete the proof of the lemma we need to estimate
\[ P \left( \left\| \max_{k \leq n} \sum_{j \leq n} (u_j^2 - Eu_j^2) \right\| \geq (t-s)^{1-\varepsilon} h_n n^{-\varepsilon} \right). \]
We shall apply Lemma A.2. Let \( j < k \leq n \). We need to estimate
\[ E \left\| \sum_{j<q \leq k} (u_q^2 - Eu_q^2) \right\|^2. \]
Now, by (3.20), (3.21), and (2.11)
\[ \sum_{j<p \leq q \leq k} E((u_p^2 - Eu_p^2)(u_q^2 - Eu_q^2)) \]
\[ = \sum_{q \leq k} \sum_{0 \leq |h_1|, |h_2| \leq c_q} c_{h_1} c_{h_2} \sum_{0 \leq |\nu_1|, |\mu_2| \leq c_p} \sum_{\nu_1, \nu_2 \in H_q} \sum_{j<p \leq q} \sum_{\mu_1, \mu_2 \in H_p} I(h, \nu_1, \mu_1, \nu_2, \mu_2, \mu). \]
where
\[ I(h, \nu, \mu) = \int_0^1 e((h_1 n_{\nu_1} + h_2 n_{\nu_2} - h_3 n_{\mu_1} - h_4 n_{\mu_2}) \omega) d\omega \]
- \[ \int_0^1 e((h_1 n_{\nu_1} + h_2 n_{\nu_2}) \omega) d\omega \cdot \int_0^1 e(-(h_3 n_{\mu_1} + h_4 n_{\mu_2}) \omega) d\omega. \]

Notice that \( I(h, \nu, \mu) = 0 \) or 1. Consequently, for fixed \( h_1, h_2, h_3, h_4, \nu_1, \) and \( \nu_2 \)
\[
\sum_{j < p \leq q \leq k} \sum_{\mu_1, \mu_2 \in H_p} I(h, \nu, \mu)
\]
is less than the number of solutions of the Diophantine equation
\[ h_1 n_{\nu_1} + h_2 n_{\nu_2} - h_3 n_{\mu_1} - h_4 n_{\mu_2} = 0 \]
subject to \( h_1 n_{\nu_1} \neq -h_2 n_{\nu_2} \) and \( h_3 n_{\mu_1} \neq -h_4 n_{\mu_2} \). According to [12, Corollary 2, p. 129] this number is bounded by \( 6 \cdot 72^{r+3} \ll 1 \). Indeed, in [12, (5.8)] set \( c = -(h_1 n_{\nu_1} + h_2 n_{\nu_2}), \ z = 1, \ a = h_3 \) and \( b = h_4 \). Thus the left-hand side in (3.30) is
\[
\ll \sum_{j < q \leq k} (\text{card } H_q)^2 \sum_{\alpha \in H_q} \sum_{1 \leq i \leq q} |c_{h_i}|^2 |c_{h_i}|^2
\]
\[
\ll (t-s)^2 \sum_{j < q \leq k} m_q^2 (\text{card } H_q)^2.
\]
In summary (3.29) is bounded by
\[
(t-s)^2 \sum_{j < q \leq k} m_q^2 (\text{card } H_q)^2 \ll (t-s)^2 \sum_{j < q \leq k} q^{4r-\frac{1}{2}}.
\]
Hence by Lemma A.2 and (3.3) the quantity (3.28) is
\[
\ll h^{-2} n^{2e} n^{4r-\frac{1}{2}} \ll n^{-\frac{1}{k}}.
\]
The result follows now from the last estimate combined with (3.22) and (3.24)–(3.28).

Lemma 3.4 follows from Lemmas 3.5 and 3.6 and and \( n \)-fold application of (2.9).

We now truncate the martingale difference sequence \( \{Y_j, \mathcal{F}_j\} \) and recenter at conditional expectations. Choose \( \rho \) with \( \frac{1}{2} \rho < \rho < 1/(8t) \) and keep it fixed. We define the events
\[
A_j = \{ \|Y_j\| \leq h_j^{\frac{1}{2} - \rho} \}
\]
and
\[
B_j = \{ \|(t-s)^{-\frac{1}{4} + \varepsilon} \sum_{\nu \leq j} E(Y_{\nu}^2 | \mathcal{F}_{\nu-1}) \| \leq 4Ch_j \}
\]
where \( C \) is as in Lemma 3.4. We set
\[
Y_j^* = Y_j 1_{A_j} 1_{B_j}, \quad W_j = Y_j^* - E(Y_j^* | \mathcal{F}_{j-1}).
\]
Then \( \{W_j, \mathcal{F}_j, j \geq 1\} \) is a bounded martingale difference sequence close to \( \{Y_j, \mathcal{F}_j, j \geq 1\} \) as the following two lemmas show.
Lemma 3.7. With probability 1
\[ \sum_{j=1}^{\infty} \| Y_j^* - Y_j \| < \infty. \]

Proof. We have by (3.33)-(3.35)
(3.36) \[ \{ \| Y_j^* - Y_j \| > 0 \} \subset A_j^c \cup B_j^c. \]
Now by Lemmas 3.2, 2.1, and (3.33), (3.3), (3.4) and the conditional version of Jensen's inequality
\[
P(A_j^c) \leq P(\| Y_j \| > \frac{1}{2} h_j^{1-p}) + P(\| v_j \| > \frac{1}{2} h_j^{1-p})
\leq h_j^{-3+6\rho} E\| Y_j \|^{6} + 0 \leq h_j^{-3+6\rho} E\| E(w_j | F_j) \|
\leq h_j^{-3+6\rho} E(\| w_j \|^{6} | F_j) \leq h_j^{-3+6\rho}(\text{card } H_j)^3(\log j)^6
\leq j^{-3+12\rho}(\log j)^6.
\]
By the convergence part of the Borel Cantelli lemma only finitely many of the events \( A_j^c \) happen with probability 1.

To show that only finitely many events \( B_j^c \) happen with probability 1, we argue as follows. For given \( j \geq 1 \), let \( m \) be such that \( m^{100} \leq j < (m + 1)^{100} \). Since \( h_{(m+1)^{100}} / h_{m^{100}} \to 1 \) we have for \( j \geq j_0 \)
\[
B_j^c \subset D_m := \left\{ \max_{j \leq (m+1)^{100}} \left( (t-s)^{-i-t} \sum_{\nu \leq j} E(Y_j^2 | F_{\nu-1}) \right) > 3Ch_{(m+1)^{100}} \right\}.
\]
Hence by Lemmas 3.3, 3.4, and (3.18), \( P(D_m) \ll m^{-3} \). Thus with probability 1 only finitely many events \( D_m \) happen which proves the above claim.

The lemma follows now from (3.36).

Lemma 3.8. As \( n \to \infty \)
\[ \sum_{j \leq n} \| E(Y_j^* | F_{j-1}) \| = o(h_j^{\frac{1}{2}}) \quad \text{a.s.} \]

Proof. Since \( E(Y_j | F_{j-1}) = 0 \) and since \( B_j \) is \( F_{j-1} \)-measurable we have by (3.33), (3.34), and Lemma 3.2
\[
\| E(Y_j^* | F_{j-1}) \| = \| 1_{B_j} E(Y_j 1_{A_j} | F_{j-1}) \| \leq \| E(Y_j 1_{A_j} | F_{j-1}) \|
= \| E(Y_j 1_{A_j} | F_{j-1}) \|
\ll \| E(y_j 1_{\{ \| y_j \| \geq \frac{1}{2} h_j^{1/2} j^{-1/2} \} | F_{j-1}) \| + 1
\ll h_j^{-\frac{1}{2}+5\rho} \| E(y_j^6 | F_{j-1}) \|.
\]
Thus by (3.3) and the estimate for \( E(\| y_j \|^{6} \) from the proof of Lemma 3.7 we obtain
\[
h_j^{-\frac{1}{2}} E\| E(Y_j^* | F_{j-1}) \| \ll j^{-3+10\rho}(\log j)^6.
\]
From the Beppo Levi theorem we conclude that
\[
\sum_{j \geq 1} h_j^{-\frac{1}{2}} E\| E(Y_j^* | F_{j-1}) \| < \infty \quad \text{a.s.}
\]
The result follows now from the Kronecker lemma.

Finally we can prove the desired exponential bound.
Lemma 3.9. Let $0 < \sigma < 2\rho - \alpha$. Then for all $0 \leq s < t \leq 1$ with $t - s \geq \frac{1}{16} h_n^{-\frac{1}{4}}$

$$P \left( \max_{k \leq n} \left| \sum_{j \leq k} W_j \right| \geq 6(t - s)^\sigma (Ch_n \log \log h_n)^{\frac{1}{2}} \right) \ll \exp(-4(t - s)^{-\sigma} \log \log h_n).$$

Proof. We apply Lemma A.3 to

$$d_j := \begin{cases} \frac{1}{2} W_j h_n^{-\frac{1}{4}} (t - s)^{-\alpha - \frac{1}{2} \sigma} C^{-\frac{1}{4}} & \text{if } j \leq n, \\ 0 & \text{if } j > n. \end{cases}$$

Let

$$C_j = C_{jn} = \left\{ \left( (t - s)^{-\frac{1}{4} + \epsilon} \sum_{\nu \leq j} E(Y_\nu^2 | F_{\nu - 1}) \right) \leq 4Ch_n \right\}.$$ 

Then $B_j \subset C_j$ for $j \leq n$ and thus

$$E(W_j^2 | F_{j-1}) \leq 1_{B_j} E(Y_j^2 | F_{j-1}) \leq 1_{C_j} E(Y_j^2 | F_{j-1}).$$

Consequently, and since $2\alpha + \sigma < \frac{1}{4}$,

$$s^2 = \frac{1}{4} h_n^{-1} (t - s)^{-2\alpha - \sigma} C^{-1} \sum_{j \leq n} E(W_j^2 | F_{j-1}) \leq 1.$$

Moreover,

$$|d_j| \leq h_n^{-\rho} (t - s)^{-\alpha - \frac{1}{2} \sigma} C^{-\frac{1}{4}} =: c.$$

Thus the probability in question equals

$$P \left( \sup_{k \leq n} \left| \sum_{j \leq k} d_j \right| > 3(t - s)^{-\frac{1}{2} \sigma} (\log \log h_n)^{\frac{1}{2}} \right) \ll \exp(-4(t - s)^{-\sigma} \log \log h_n)$$

since

$$h_n^{-\rho} (t - s)^{-\alpha - \frac{1}{2} \sigma} C^{-\frac{1}{4}} 3(t - s)^{-\frac{1}{2} \sigma} (\log \log h_n)^{\frac{1}{2}} \ll h_n^{-\rho} (\log \log h_n)^{\frac{1}{4}} h_n^{\frac{1}{2}(\alpha + \sigma)} \rightarrow 0.$$ 

Lemma 3.10. With probability 1 there exists $n_0 = n_0(\omega)$ such that for all $n \geq n_0$ and all $s, t$ with $0 \leq s < t \leq 1$

$$\max_{k \leq n} \left| \sum_{j \leq k} W_j \right| \leq 12 \cdot 2^\rho (t - s)^\alpha (Ch_n \log \log h_n)^{\frac{1}{2}} + \frac{1}{4} h_n^{\frac{1}{2}}.$$

Proof. As in the proof of Lemma 3.2 we expand $s$ and $t$ in dyadic expansion so that (3.12) holds with $m$ and $M$ chosen suitably below. Instead of (3.13) we define

$$Z = Z(s, t) = Z(k; s, t) = \left| \sum_{j \leq k} W_j(s, t) \right|.$$
We note that (3.14) still holds. We apply (3.14) repeatedly and obtain instead of (3.16)
(3.37)
\[
Z(s,t) \leq Z(a2^{-m}, b2^{-m}) + \sum_{i=m+1}^{M} Z(a, (a_1 + i)2^{-1}) + Z(b, (b_1 + i)2^{-1}) + Z(a, a_1, b, b_1, (a_1 + i)2^{-1}, (b_1 + i)2^{-1}) + Z(a, a_1, b, b_1, (a_1 + i)2^{-1}, (b_1 + i)2^{-1}) + 2\phi(\log h_n)
\]
where \(a, b, a_i, b_i (m < i \leq M + 1)\) are integers with \(0 < a, b \leq 2^m, 0 < a_i, b_i < 2^i (m < i \leq M + 1)\). We set
(3.38) \(m = m(n) = \lfloor \log \log h_n \rfloor, \quad M = M(n) = \left[ \frac{1}{2\log 2} \log h_n \right] + 4\)
and
\[
\phi(x) = 12(C\log \log x)^{\frac{1}{4}}.
\]
We define the following events:
\[
E_n(a, b) = \left\{ \max_{k \leq n} Z(h_k; a2^{-m}, b2^{-m}) \geq ((b - a)2^{-m})^a\phi(h_n) \right\},
\]
\[
E_n = \bigcup_{0 < a, b < 2^m} E_n(a, b),
\]
\[
F_n(i, a) = \left\{ \max_{k \leq n} Z(h_k; a2^{-i}, (a_1 + 1)2^{-i}) \geq 2^{-ai}\phi(h_n) \right\},
\]
\[
F_n = \bigcup_{m < i \leq M} \bigcup_{0 < a < 2^i} F_n(i, a).
\]
By Lemma 3.9
\[
P(E_n(a, b)) \ll \exp(-4\log \log h_n) \ll (\log h_n)^{-4}
\]
and thus by (3.38)
(3.39) \(P(E_n) \ll 2^{2m}(\log h_n)^{-4} \ll (\log h_n)^{-3}\).
Similarly
\[
P(F_n(i, a)) \ll \exp(-4 \cdot 2^{ai}\log \log h_n)
\]
and so
\[
P(F_n) \ll \sum_{m < i} 2^i \exp(-4 \cdot 2^{ai}\log \log h_n) \ll (\log h_n)^{-3}.
\]
Consequently by (3.39)
\[
P(E_n \cup F_n) \ll (\log h_n)^{-3}
\]
and so
\[
\sum_{p=1}^{\infty} P(E_{2^p} \cup F_{2^p}) < \infty.
\]
The Borel Cantelli lemma implies that with probability 1 only finitely many of the events $E_{2^p}$ or $F_{2^p}$ occur. Let $n$ be sufficiently large and let $p$ be such that $2^{p-1} \leq n < 2^p$. Then by (3.37) we have with probability 1 for all $0 \leq s < t \leq 1$

$$\max_{k \leq n} Z(k, s, t) \leq \left((b - a)2^{-m(2^p)}\right)^{\alpha} + 2 \sum_{m(2^p) \leq M(2^p)} 2^{-ia} \phi(h_{2^p}) + \frac{1}{4}h_n^\frac{1}{2}$$

$$\leq (t - s)^{\alpha} \phi(h_{2^p}) + o(h_{2^p}) + \frac{1}{4}h_n^\frac{1}{2} \leq 2^a(t - s)^{\alpha} \phi(h_n) + \frac{1}{4}h_n^\frac{1}{2}.$$ 

This proves the lemma.

**Lemma 3.11.** Let $\epsilon < 1/(8\tau)$. Then with probability 1

$$\max_{k \leq h_n - h_{n-1}} \left| \sum_{\nu = h_{n-1}+1}^{h_{n-1}+k} x_\nu \right| \ll h_n^{\frac{1}{4}-\epsilon}.$$

**Proof.** We apply Lemma 2.1 with $N = (h._{n-1}, h_n)$ and obtain

$$P \left( \max_{k \leq h_n - h_{n-1}} \left| \sum_{\nu = h_{n-1}+1}^{h_{n-1}+k} x_\nu \right| \geq h_n^{\frac{1}{4}-\epsilon} \right)$$

$$\leq h_n^{-3+6\epsilon}(h_n - h_{n-1})^{3}(\log n)^{6} \ll n^{-3+12\epsilon}(\log n)^{6}.$$ 

The lemma follows from the convergence part of the Borel Cantelli lemma.

We finally can finish the proof of Theorem 4. Let $N$ be given. Find $n$ such that $h_{n-1} \leq N < h_n$. Then by Lemmas 3.1, 3.2, 3.3, 3.7, 3.8, 3.10, and 3.11 and by (3.3) there exists with probability 1 and $N_0 = N_0(\alpha, \omega)$ such that for all $N \geq N_0$ and all $s, t$ with $0 \leq s < t \leq 1$

$$\max_{k \leq N} k |F_k(t) - F_k(s)| \leq \max_{p \leq n} h_p|F_{h_p}(t) - F_{h_p}(s)| + \max_{q \leq h_n - h_{n-1}} \left| \sum_{\nu = h_{n-1}+1}^{h_{n-1}+q} x_\nu \right|$$

$$\leq 12 \cdot 2^a(t - s)^{\alpha}(Ch_n \log \log h_n)^{\frac{1}{4}} + \frac{1}{2}h_{n-1}^{\frac{1}{2}} + o(h_n^{\frac{1}{2}})$$

$$\leq 13 \cdot 2^a(t - s)^{\alpha}(CN \log \log N)^{\frac{1}{4}} + N^{\frac{1}{4}}.$$ 

4. PROOF OF THEOREM 2

We shall apply Lemma A.4. The random variable

$$U(\omega) = U(\omega_1, \omega_2) = \omega_2, \quad \omega \in [0, 1)^2,$$

has uniform distribution and is independent of the sequence $\{\cos 2\pi n_\nu \omega_1, \nu \geq 1\}$.

Let $H_j$, $I_j$ and $F_j$ be defined as in Section 3. let $F'_j$ be the set of all rectangles $A \times [0, 1)$ where $A \in F_j$. We define

$$\xi_\nu = \xi_\nu(\omega_1, \omega_2) = 2^{\nu+4}E(\cos 2\pi n_\nu \cdot |F'_j|), \quad \nu \in H_j.$$ 

Then $\xi_\nu$ is a random variable defined on $[0, 1)^2$ by

$$\xi_\nu(\omega_1, \omega_2) = 2^{\nu+4} \int_{U_{kh_j}} \cos 2\pi n_\nu u \, du, \quad \omega_1 \in U_{kh_j}, \quad \omega_2 \in [0, 1),$$
for \( \nu \in H_j \). Thus by (3.5) we have for all \((\omega_1, \omega_2) \in [0, 1)^2\)

\[
(4.1) \quad \left| \xi_\nu(\omega_1, \omega_2) - 2^{\frac{1}{4}} \cos 2\pi n_\nu \omega_1 \right| \leq 4\pi n_\nu 2^{-n_\nu} \ll h_n^{-12}, \quad \nu \in H_j.
\]

We set

\[
(4.2) \quad y_j := \sum_{\nu \in H_j} \xi_\nu
\]

and

\[
(4.3) \quad X_j := y_j - E(y_j \mid \mathcal{F}_{j-1}).
\]

Then \(\{X_j, \mathcal{F}_j', j \geq 1\}\) is a martingale difference sequence and this is the sequence to which we later apply Lemma A.4. We first show that it is an adequate approximation for the partial sums of the cosine series.

**Lemma 4.1.** For almost all \((\omega_1, \omega_2) \in [0, 1)^2\)

\[
\sum_{\nu \leq h_n} 2^{\frac{1}{4}} \cos 2\pi n_\nu \omega_1 - \sum_{j \leq n} X_j \ll h_n^{\frac{1}{2}(1-\alpha)}
\]

where \(\alpha < 1/(4\pi)\) is as in Theorem 4 and Section 3.

**Proof.** By (4.1)

\[
(4.4) \quad \sum_{\nu = 1}^{\infty} |\xi_\nu - 2^{\frac{1}{4}} \cos 2\pi n_\nu \cdot | < \infty.
\]

Since for \(0 < a < b \leq 1\)

\[
\left| \int_a^b \cos 2\pi n_\nu u \, du \right| \leq 2n_\nu^{-1}
\]

we have for \(\nu \in H_j\)

\[
(4.5) \quad |E(\xi_\nu \mid \mathcal{F}_{j-1})| = |E(\cos 2\pi n_\nu \cdot \mid \mathcal{F}_{j-1})|2^{\frac{1}{4}} \leq 2^{n_\nu + \frac{3}{4}} n_\nu^{-1} \ll \exp \left(-\frac{1}{2} j^4\right)
\]

by (3.5) and (3.1). Hence by (3.3)

\[
(4.6) \quad \sum_{j = 1}^{\infty} |X_j - y_j| \ll \sum_{j = 1}^{\infty} h_j \exp \left(-\frac{1}{2} j^4\right) < \infty.
\]

Finally by Koksma's inequality and Lemma 3.1 we have with probability 1

\[
\left\| \sum_{j \leq n} \sum_{\nu \in l_j} \cos 2\pi n_\nu \cdot \right\| \ll h_n^{\frac{1}{2}(1-\alpha)}.
\]

This with (4.4) and (4.6) proves the lemma.

**Lemma 4.2.** For some \(\rho > 0\), depending on \(\tau\) only

\[
V_n := \sum_{j \leq n} E(X_j^2 \mid \mathcal{F}_{j-1}) = h_n + O(h_n^{1-\rho}) \quad a.s.
\]

where the constant implied by \(O\) is nonrandom.
Proof. By (4.3), (4.5), and (4.1),

\[
E(X_j^2 \mid \mathcal{F}_{j-1}) = E(y_j^2 \mid \mathcal{F}_{j-1}) + O\left(h_j \exp\left(-\frac{1}{4}j^4\right)\right)
\]
\[
= \sum_{\mu, \nu \in H_j} E(\xi_{\mu} \xi_{\nu} \mid \mathcal{F}_{j-1}) + O\left(\exp\left(-\frac{1}{4}j^4\right)\right)
\]
\[
= 2 \sum_{\mu, \nu \in H_j} E(\cos 2\pi n_{\mu} \cdot \cos 2\pi n_{\nu} \cdot \mathcal{F}_{j-1}) + O(h_j^{-10})
\]
\[
= \text{card } H_j + O(h_j^{-10}).
\]

The last relation comes from the fact that on an interval \(U\) of the form (3.17) a typical conditional expectation equals

\[
2^x \cdot \frac{1}{2} \int_U \cos(2\pi(n_{\mu} + n_{\nu})u)du + 2^x \cdot \frac{1}{2} \int_U \cos(2\pi(n_{\mu} - n_{\nu})u)du.
\]

Thus for \(\mu = \nu\) the second term equals \(\frac{1}{2}\) and this explains the main term. Because of (3.3), (3.1), and (3.5) and because of Tijdeman's theorem [24] in all other cases the above terms are bounded by

\[
2^x |n_{\nu} - n_{\mu}|^{-1} \leq 2^x \max_{i \in H_j} \frac{(\log n_i)^{C_1}}{n_i} \ll \exp(j^2 - (j - 1)^4)h_j^{12} \cdot e^{-j^2} j^{2C_1} \ll \exp(-\frac{1}{4}j^4).
\]

Thus

\[
V_n = \sum_{j \leq n} \text{card } H_j + O(1)
\]
\[
= h_n - \sum_{j \leq n} \text{card } I_j + O(1) = h_n + O(h_n^{1-\rho}).
\]

We now apply Lemma A.4 with \(f(x) = x^{1-\varepsilon}, \varepsilon < 1/(4\tau)\). Using (2.8) and (3.3) we see that a typical term in (A3) is bounded by

\[
V_n^{-3+3\varepsilon} E|X_n|^6 \ll h_n^{-3+3\varepsilon}(\text{card } H_n)^3 \ll n^{-3+6\varepsilon}.
\]

Thus (A3) holds. Hence we obtain a sequence \(\{Y_j(\omega_1, \omega_2), j \geq 1\}\) of independent standard normal variables such that with probability 1

\[
\sum_{n \geq 1; h_n + O(h_n^{1-\rho}) \leq t} X_n - \sum_{j \leq t} Y_j \ll t^{1-\varepsilon/50}.
\]

In view of Lemma 4.1 this implies

\[
2^\frac{1}{2} \sum_{\nu \leq h_n} \cos 2\pi n_{\nu} \omega_1 - \sum_{j \leq h_n + O(h_n^{1-\rho})} Y_j(\omega_1, \omega_2) \ll h_n^{1-\varepsilon/50}.
\]

Koksma's inequality [13, p. 143] and Lemma 3.11 imply

\[
\max_{k \leq h_n - h_{n-1}} \left| \sum_{\nu = h_{n-1}+1}^{h_{n-1}+k} \cos 2\pi n_{\nu} \omega_1 \right| \ll h_n^{1-\varepsilon} \text{ a.s.}
\]
and Levy's maximal inequality implies that

\[ \max_{k \leq h_n - h_{n-1} + O(h_n^{1-\rho})} \left| \sum_{\nu = h_{n-1} + 1}^{h_{n-1} + k} Y_j(\omega_1, \omega_2) \right| \ll h_n^{\frac{1}{2} - \lambda} \quad \text{a.s.} \tag{4.9} \]

for some \( \lambda > 0 \), depending only on \( \rho \). Theorem 2 follows from (4.7), (4.8), and (4.9).

**Appendix**

Let \( g(i, j) \) be a superadditive function, i.e., a function satisfying

\[
g(i, j) = 0 \quad \text{for all } 1 \leq i \leq j \leq n
\]

\[
g(i, j) \leq g(i, j + 1) \quad \text{for all } 1 \leq i \leq j \leq n
\]

\[
g(i, j) + g(j + 1, k) \leq g(i, k) \quad \text{for all } 1 \leq i \leq j \leq k \leq n.
\]

The following lemma is a special case of [17, Theorem 3.1].

**Lemma A.1.** Let \( \alpha > 1 \) and \( \gamma \geq 1 \) be given. Let \( X_1, X_2, \ldots \) be a sequence of random variables with finite \( \gamma \)th moments. Suppose that there exists a superadditive function \( g(i, j) \) such that

\[ E \left| \sum_{1 \leq \nu \leq j} X_\nu \right|^\gamma \leq g^\alpha(i, j) \quad \text{for all } 1 \leq i \leq j \leq n. \tag{A1} \]

Then there exists a constant \( A \), depending only on \( \alpha \) and \( \gamma \) (but not on \( n \), \( X_k \), or otherwise on \( g \)) such that

\[ E \max_{k \leq n} \left| \sum_{\nu = 1}^{k} X_\nu \right|^\gamma \leq A g^\alpha(1, n). \]

There are similar results when (A1) is replaced by probability estimates [6, Theorem 12.2] and [17, Theorem 3.2]. A result of this type is needed in the proof of Lemma 3.6. Unfortunately, none of the two above results is easy to apply in this situation. The following lemma is not as sharp, but it applies easily and is almost trivial.

**Lemma A.2.** Let \( X_1, X_2, \ldots \) be a sequence of random variables with \( k \)th partial sum \( S_k \). Let \( n \geq 1 \) and \( t \geq 0 \) be fixed. Suppose that for some \( u_\nu \geq 0 \), \( 1 \leq \nu \leq n \),

\[ P(|S_k - S_j| \geq t) \leq \sum_{\nu = j+1}^{k} u_\nu \quad \text{for all } 1 \leq j < k \leq n. \tag{A2} \]

Then

\[ P \left( \max_{k \leq n} |S_k| \geq 2t \log n \right) \leq \sum_{\nu \leq n} u_\nu. \]

**Proof.** Write \( k \leq n \) in dyadic expansion

\[ k = 2^p + \epsilon_1 2^{p-1} + \cdots + \epsilon_p \]
where $\varepsilon_i = 0, 1$ ($1 \leq i \leq p$). Then $p < 2 \log k \leq 2 \log n$ and

$$|S_k| \leq |S_{2^p}| + |S_{2^p + 2^{p-1}} - S_{2^p}| + \ldots.$$ 

Thus applying (A2) $p$ times we obtain the result.

We quote a special case of a recent result of Pinelis [21].

Lemma A.3. Let $\{f_n, F_n, n \geq 0\}$ be a real-valued martingale $f_0 \equiv 0$, and let $d_n = f_n - f_{n-1}$, $n = 1, 2, \ldots, d_0 \equiv 0$ be its associated martingale difference sequence. Put

$$f^* = \sup \{|f_n|, n = 0, 1, 2, \ldots\},$$

$$d^* = \sup \{|d_n|, n = 0, 1, 2, \ldots\},$$

and

$$s = \left( \sum_{n=1}^{\infty} E(d_n^2 | F_{n-1}) \right)^{\frac{1}{2}}.$$

Suppose that $\text{ess sup} s \leq 1$ and $\text{ess sup} d^* \leq c$, for some $c > 0$. Then

$$P(f^* > r) < 2 \exp \left( \frac{r}{c} - \left( \frac{r}{c} + \frac{2}{c^2} \right) \log \left( 1 + \frac{rc}{2} \right) \right), \quad r \geq 0.$$

We also need a theorem of Strassen [23] which we restate in a form more convenient for our purposes (see [16, Theorem 7]).

Lemma A.4. Let $\{X_n, F_n, n \geq 1\}$ be a real-valued square-integrable martingale difference sequence, defined on some probability space $(\Omega, F, P)$. Let $f$ be a nondecreasing function with $f(x) \to \infty$ as $x \to \infty$ and such that $\frac{f(x) \log x^n}{x}$ is nonincreasing for some $\alpha > 50$. Suppose that

$$V_n := \sum_{j \leq n} E(X_j^2 | F_{j-1}) \to \infty \quad \text{a.s.}$$

and that

(A3) \quad $\sum_{n \geq 1} E(X_n^2 1\{X_n^2 > f(V_n)\}/f(V_n)) < \infty.$

Suppose there exists a random variable $U$, uniformly distributed over $[0, 1)$ and independent of the sequence $\{X_n\}$. Then there exists a sequence $\{Y_n, n \geq 1\}$ of independent standard normal random variables defined on $(\Omega, F, P)$ such that with probability 1

$$\sum_{n \geq 1} x_n 1\{V_n \leq t\} - \sum_{m \leq t} Y_m \ll t^{\frac{1}{50}}(f(t)/t)^{1/50}.$$

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In a recent paper, I. Berkes and W. Philipp, The size of trigonometric and Walsh series and uniform distribution mod 1, J. London Math. Soc. (to appear), a counterexample to the Erdős-Baker conjecture (1.1) is provided. In addition, we prove that for sequences $\{n_k\}$ satisfying (1.4) the upper half of the law of the iterated logarithm (1.3) is indeed false.
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