VISITATIONS OF RULED SUMS⁽¹⁾

BY

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ABSTRACT. Let $\{X_i\}$ be a sequence of independent identically distributed random variables and for $D \subseteq I^+$ let $S_D = \sum_{i \in D} X_i$. A rule $(\cdot)$ is a mapping $I^+ \rightarrow 2^{I^+}$: $\forall n \mid (n) \mid = n$ and $S_{(\cdot)} = \{S_{(n)}\}$ is its associated ruled sum.

Ruled sums generalize ordinary sums S_n . Indeed, all a.e. results for S_n can be investigated for $S_{(n)}$ frequently requiring different methods and sometimes yielding different conclusions. In a previous paper we studied strong laws of large numbers and the law of the iterated logarithm.

In this paper we study infinite visitation. Under suitable hypotheses on the basic distribution function F of the X_i we show that, for all rules $(\cdot)$, $S_{(n)}$ visits each integer infinitely often a.e. in the lattice case (or has all points of the real line as accumulation points in the nonlattice case). In fact we obtain a "rate of visitation." There follows extensions of the Pólya theorem on encounters in the plane and 3-space from random walks to these ruled sums.

Some equivalence relations and partial orderings on rules are defined. For normal variables this leads to an extension of the previously mentioned result for ruled sums of the type of the iterated logarithm law.

I. Introduction and statement of visitation results. Let $\{X_i\}$ be a sequence of independent, identically distributed random variables (i.i.d.), and for $D \subseteq I^+$ ($\equiv$ set of positive integers) let $S_D \equiv \sum_{i \in D} X_i$. Let $\mathbf{R} \equiv$ set of all functions $(\cdot): I^+ \rightarrow 2^{I^+}$ where (n) contains exactly n elements for each $n \in I^+$. $(\cdot) \in \mathbf{R}$ is called a rule and $S_{(n)}$ is called its ruled sum. Intuitively there are two rules that are in some sense the "extreme rules":

(1) $(n) \equiv \{1, 2, \dots, n\}$ which we denote by n ,

(2) any rule $(\cdot)$ that satisfies $(n) \cap (m) = \emptyset$ for $n \neq m$; we denote such a rule by $\langle \cdot \rangle$.

Certainly n and $\langle n \rangle$ are at the top and bottom respectively in the two partial orderings given by

$$(\cdot) \leq (\cdot)' \text{ iff } \left| \bigcap_{i=1}^m (n_i) \right| \leq \left| \bigcap_{i=1}^m (n_i)' \right| \text{ for } n_1, \dots, n_m, m \in I^+$$

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and

$$(\cdot) \leq (\cdot)' \text{ iff } |(n_1) \cap (n_2)| \leq |(n_1)' \cap (n_2)'| \text{ for } n_1, n_2 \in I^+.$$

($|A|$ means the cardinality of the set A .) $\leq$ is the natural ordering on rules since $R \leq R'$ and $R' \leq R$ iff $R \cong R'$ under the natural equivalence relation:

$$\begin{aligned} (\cdot) \cong (\cdot)' \text{ iff there exists a 1-1 map } \sigma \text{ from} \\ \bigcup_{n=1}^{\infty} (n) \text{ onto } \bigcup_{n=1}^{\infty} (n)' \text{ such that } \bigcup_{i \in (n)} \{\sigma(i)\} = (n)' \\ \text{for } n \in I^+. \end{aligned}$$

(See Theorem 6 of § III.)

The weaker ordering $\leq$ is more natural when X_i are normally distributed since a Gaussian sequence $S_{(n)}$ is determined by its covariance $E(S_{(n)}S_{(j)}) = |(n) \cap (j)|$ assuming X_1 is $\mathcal{N}(0, 1)$. Moreover, the following two results of § 3 help to confirm that the intuition and this ordering are in accord.

Proposition 2. For $X_1 \mathcal{N}(0, 1)$, $(\cdot) \leq (\cdot)'$ iff $P[S_{(n)} \in (u, v) | S_{(j)} \in (u, v)] \leq P[S_{(n)'} \in (u, v) | S_{(j)'} \in (u, v)]$ for all $n, j \in I^+$ and $-\infty \leq u \leq v \leq \infty$.

Theorem 7. For $X_1 \mathcal{N}(0, 1)$, $(\cdot) \leq (\cdot)'$ implies $\overline{\lim} S_{(n)}/a_n \leq \overline{\lim} S_{(n)'} / a_n$ for all sequences $\{a_n\}$: $a_n \rightarrow \infty$.

When X_1 is a normal variable, Theorem 7 extends the result

$$\overline{\lim} S_n/a_n \leq \overline{\lim} S_{(n)}/a_n \leq \overline{\lim} S_{(n)'} / a_n$$

obtained in [1] for symmetric X_1 and "well-behaved" sequences $\{a_n\}$.

Thus a conjecture like

(C) A theorem which holds for n and $(\cdot)$ holds for all $(\cdot) \in R$

arises rather naturally. An unfortunate instance where (C) fails is that although the tail sigma field of $\{S_{(n)}\}$ for $(\cdot) = n$ or $(\cdot)$ is trivial (the Hewitt-Savage theorem [3, p. 63]), it need not be trivial for more general rules. Indeed the tail event $[|S_{(2n)} - S_{(2n-1)}| < \epsilon \text{ i.o.}]$ will never be a 0-1 event for certain rules if $1 > P[|X_1| < \epsilon] > 0$ (see § 4). Nevertheless, one hopes that (C) holds for certain classes of theorems. This paper is largely devoted to (C) for the class of theorems dealing with the visitations and distribution of values of the sequence of ruled sums of a rule.

In the case of visitation theorems, we take X_1 to be a lattice random variable on I ($\equiv$ set of integers). Without loss of generality we can assume that X_1 satisfies

$$(\omega) \quad \gcd_{i,j \in I} \{ |j - i| : P[X_1 = j] > 0 \text{ and } P[X_1 = i] > 0 \} = 1 \quad [10, \text{p. 232}].$$

Hence if $P[X_1 < 0] > 0$ and $P[X_1 > 0] > 0$ as we assume throughout then $\forall i \in I$, $\exists n(i) \in I^+ : P[S_{n(i)} = i] > 0$.

For $\{B_n\}$ a sequence of subsets of I ,

$$m_n(B_n) \equiv P[S_n \in B_n] \quad \text{and} \quad M_n(B_n) \equiv \sum_{i=1}^n m_i(B_i) + 1.$$

If $B_n = \{a\}$, then we write $m_n(a)$ and $M_n(a)$ respectively. If $a = 0$, then we suppress the a 's in the symbols altogether as we will do whenever B_n is understood. $\phi(A)$ denotes the indicator function of the set A , $\hat{f}(\theta)$ the characteristic function of X_1 .

The results in this paper on visitation theorems will now be discussed. In the following K denotes differing positive constants and $(*)$ or $(**)$ will be used for differing displayed items used only *locally*.

Theorem 1. Let $() \in \mathbb{R}$, $EX^+ = EX^-$, and let $M_n \xrightarrow{n} \infty$.

(1) Let $a \in I$, and $g: R^+$ (set of positive reals) $\rightarrow R^+$, where $g(x) \uparrow \infty$ and its derivative $g'(x) > 0$ and $\downarrow$. If g satisfies for some $\epsilon_0 \in (0, 1)$

$$(\dagger) \quad \sum_{k=j}^n g'(M_k(a))(P[S_{(k)} = a | S_{(j)} = a] - P[S_{(k)} = a]) \\ \leq K(g^{\epsilon_0}(M_n(a))) \quad \text{for all } 1 \leq j \leq n \in I^+,$$

then

$$\lim_n (g(M_n(a)))^{-1} \sum_{i=1}^n g'(M_i(a)) \phi[S_{(i)} = a] = 1 \quad a.e.$$

(2) Let $m_n \geq n^{-\alpha}$ for all $n \in I^+$ with $0 < \alpha < 1$. If $\exists \epsilon < 1/2: \forall a_1, a_2 \in I$, $\forall 1 \leq j \leq n \in I^+$,

$$(\dagger\dagger) \quad \sum_{k=j}^n P[S_{(k)} = a_1 | S_{(j)} = a_2] = O(M_n^{1+2\epsilon})$$

then, for $a \in I$,

$$P \left[\left| \sum_{k=1}^n \phi[S_{(k)} = 0] - \sum_{k=1}^n \phi[S_{(k)} = a] \right| > M_n^{3/4 + \epsilon/2 + \delta} \text{ i.o.} \right] = 0$$

for all $\delta > 0$.

(3) If $(\dagger)$ holds for $g(x) = \lg x$, $m_n \geq n^{-\alpha}$ and $(\dagger\dagger)$ is satisfied, then

$$\sum_{k=1}^n \phi[S_{(k)} = 0] / \sum_{k=1}^n \phi[S_{(k)} = a] \xrightarrow{n} 1 \quad a.e.$$

Chung and Erdős [4] essentially showed that for $() = n$ and $EX^- = EX^+$, $(\dagger)$ holds for $g(x) = \lg x$, and $\epsilon = 0$ and $(\dagger\dagger)$ holds with $\epsilon = 0$, i.e., they showed that, for all $a \in I$, $\lim_n (\lg(M_n))^{-1} \sum_{k=1}^n \phi[S_k = a] / M_k \rightarrow 1$,

$$P \left[\left| \sum_{k=1}^n \phi[S_k = a] - \sum_{k=1}^n \phi[S_k = 0] \right| > M_n^{3/4+\delta} \text{ i.o.} \right] = 0$$

for all $\delta > 0$, and $\sum_{k=1}^n \phi[S_k = a] / \sum_{k=1}^n \phi[S_k = 0] \xrightarrow{n} 1$ a.e. The proofs for (1)–(3) of Theorem 1 are adaptations of the proofs for $() = n$ that appear in [4]. The main difference in proof arises in finding a replacement for their repeated appeal to the Markov property of S_n which is, of course, lacking for general rules.

We note that since $S_{(n)}$ is a sequence of independent random variables, $(\dagger)$ and $(\dagger\dagger)$ always hold. Thus (1)–(3) of Theorem 1 are candidates for $(\mathcal{C})$. In this direction, we will show

Proposition 1. *If X_1 is symmetric, then for $() \in \mathbf{R}$,*

- (1) *$(\dagger)$ holds for all $a \in I$ and $g(x)$ which satisfies, for some $1 > \epsilon_0 > 0$, $xg'(x) = O(g^{\epsilon_0}(x))$, e.g., $g(x) = O(\lg^k(x))$ for $k \in I$.*
 (2) *$(\dagger\dagger)$ holds.*

Because $P[S_{n+1} = a] \geq m_1(a - a')P[S_n = a']$, we see for $M_n \rightarrow \infty$

$$(\rho) \quad M_n(a)/M_n(a') = O(1) \quad \text{for } a, a' \in I.$$

(Note (ρ) also holds if S_n were the sum of n i.i.d. k -dimensional random vectors.)

If X_1 is in the domain of attraction of a stable law of exponent $1/\alpha > 1$ (e.g., for $EX_1^2 < \infty$, $\alpha = 1/2$), then $P[S_n = 0] \sim Kn^{-\alpha}$. Thus (ρ) combined with Theorem 1 and Proposition 1 yields

Corollary 1. *Let X_1 be symmetric and $M_n \rightarrow \infty$. For $() \in \mathbf{R}$,*

- (a) *$(\lg M_n(a))^{-1} \sum_{k=1}^n \phi[S_{(k)} = a] / M_k(a) \xrightarrow{n} 1$ a.e. for all $a \in I$.*
 (b) *If X_1 is, moreover, in the domain of attraction of a stable law with exponent > 1 , then for all $a \in I$*

$$\sum_{k=1}^n \phi[S_{(k)} = a] / \sum_{k=1}^n \phi[S_{(k)} = 0] \xrightarrow{n} 1 \quad a.e.$$

Now (ρ) , the Borel-Cantelli lemma, (a) of Corollary 1 and the fact that

$$(\lg M_n(a))^{-1} \sum_{k=1}^n \phi[S_{(k)} = a] / M_k(a) = O \left((\lg M_n(a))^{-1} \sum_{k=1}^n \phi[S_{(k)} = a] \right)$$

gives

Corollary 2. For X_1 symmetric, $a \in I$ and $(\cdot) \in R$

$$P[S_{(n)} = a \text{ i.o.}] = \begin{cases} 1 \\ 0 \end{cases} \text{ iff } \sum_{n=1}^{\infty} P[S_n = 0] \begin{cases} = \infty \\ < \infty \end{cases} \text{ iff } \int_{-\pi}^{\pi} \frac{d\theta}{1 - \hat{f}(\theta)} \begin{cases} = \infty \\ < \infty \end{cases}.$$

(The characteristic function part of Corollary 2 is well known and is proved easily from (C2) and (C5) in the first part of § II. Corollary 2 has its genesis in Chung and Fuchs [5] for $(\cdot) = n$ and general X_1 .)

Theorem 2. Let X_1 be a lattice random variable. Then the following nine conditions are equivalent.

- (1) $\int_{-\pi}^{\pi} (1 - |\hat{f}(\theta)|)^{-1} d\theta = \infty$.
- (2) $\sum_n \sum_a P^2[S_n = a] = \infty$.
- (3) $\exists d_n \in I: \sum P[S_n = d_n] = \infty$.
- (4a) $\forall (\cdot) \in R: \sum m_n(S_{(n)}(\omega)) = \infty \text{ a.s.}$
- (4b) $\exists (\cdot) \in R: P[\sum m_n(S_{(n)}(\omega)) = \infty] > 0$.
- (5a) $\forall (\cdot) \in R \exists d_n \in I: \forall a \in R P[S_{(n)} - d_n = a \text{ i.o.}] = 1$.
- (5b) $\exists (\cdot) \in R, d_n \in I \text{ and } a \in R: P[S_{(n)} - d_n = a \text{ i.o.}] > 0$.
- (6a) $\forall (\cdot)_1, (\cdot)_2 \in R: (n)_1 \cap (m)_2 = \emptyset \forall n, m \in I^+ P[S_{(n)_1} = S_{(n)_2} + a \text{ i.o.}] = 1, \forall a \in I$.
- (6b) $\exists (\cdot)_1, (\cdot)_2 \in R: (n)_1 \cap (m)_2 = \emptyset \forall n, m \in I^+; \text{ and } \exists a \in I: P[S_{(n)_1} = S_{(n)_2} + a \text{ i.o.}] > 0$.

These nine equivalent conditions above generalize the notion of recurrence to essential recurrence in the spirit of [10, p. 250]. Even for S_n , this is a true extension of recurrence. In particular, if $E|X_1| < \infty$ and $EX_1 \neq 0$ then $P[S_n = a \text{ i.o.}] = 0$ for all a , while $P[S_n - nEX_1 = a \text{ i.o.}] = 1$ for all a [3, p. 56]. Of course, if X_1 is symmetric, then (1) is equivalent to $\int_{-\pi}^{\pi} (1 - \hat{f}(\theta))^{-1} d\theta = \infty$, i.e., essential recurrence becomes ordinary recurrence.

Theorem 2 seems fairly complete since condition (1) is a hypothesis on the basic datum, the distribution of X_1 , and moreover, no extraneous hypothesis is needed. The same is true for Theorem 3.

Theorem 3. Let (X_i) be i.i.d. lattice random variables. Let $(\cdot)_1, (\cdot)_2, (\cdot)_3, (\cdot)_4 \in R$ for which $(n)_i \cap (m)_j = \emptyset$ for $i \neq j = 1, 2, 3, 4$, and $n, m \in I^+$. The following five conditions are equivalent.

- (1) $\int_{-\pi}^{\pi} \int (1 - |\hat{f}(\theta)\hat{f}(\phi)|)^{-1} d\theta d\phi = \infty$.
- (2) $\sum_n (\sum_a P^2[S_n = a])^2 = \infty$.
- (3) $P[(S_{(n)_1}, S_{(n)_2}) = (S_{(n)_3}, S_{(n)_4}) \text{ i.o.}] = 1$.
- (4) $\sum_n \sum_a P^3[S_n = a] = \infty$.
- (5) $P[S_{(n)_1} = S_{(n)_2} = S_{(n)_3} \text{ i.o.}] = 1$.

(3) and (5) of Theorem 3 are, as was (6) of Theorem 2, extensions of Pólya-type theorems; e.g., (6) of Theorem 2 is an extension of two independent particles, each of which is performing a random walk in one dimension, meeting infinitely often (Feller [6, vol. 1, p. 329]).

Of course, one could consider similar questions for nonlattice random variables. For $EX_1^2 < \infty$, an application of a local limit theorem in the proof of Theorem 2 will yield Theorem 2 with points of infinite visitation replaced by accumulation points. An example of how one is to carry out these suggested steps is given in the proof of the following:

Theorem 4. *Let $EX_1^2 < \infty$ and let $\{a_n\}$ be a nondecreasing sequence of positive real numbers with $a_n = o(n^{1/2})$. Then the following are equivalent.*

- (1) $EX = 0$.
- (2) For some $b \in \text{Reals}$ and some $() \in \mathbf{R}$, $P[b \text{ is an accumulation point of } S_{(n)}/a_n] > 0$.
- (3) For all $b \in \text{Reals}$ and all $() \in \mathbf{R}$, $P[b \text{ is an accumulation point of } S_{(n)}/a_n] = 1$.
- (4) $\sum_n P[|S_n/a_n| < \epsilon] = \infty$ for all $\epsilon > 0$.

This result was obtained for $() = n$ and $a_n = n^{-\alpha}$, $0 < \alpha < 1/2$, without the assumption $EX_1^2 < \infty$ by Kesten [8, Theorem 4, p. 1173]. We note here that our proof of Theorem 4 actually shows that (3) $\Rightarrow$ (1) under the weaker assumption that $a_n = o(n)$. However Kesten shows that one cannot in general let $a_n = n$ and have (1) $\Rightarrow$ (3) [8, Theorem 7, p. 1196]. Finally we see that by taking $a_n = 1$ for all n , we obtain Corollary 2 with the following modifications. Replace the hypothesis X_1 symmetric with $E(X_1)^2 < \infty$ and insert Real part in front of the integral.

Theorem 5. *Let $a > 0$ be such that it is not a rational multiple of the span of X_1 when X_1 is a lattice variable. For $() \in \mathbf{R}$, $S_{(n)}$ is equidistributed mod $[0, a)$.*

This theorem generalizes a result of Robbins [16] for $() \equiv n$. Theorem 5 requires no extra distribution hypotheses and has a simple proof. Corollary 1 with its subsidiary hypotheses and a more difficult proof yields asymptotically equal number of visits to any pair of lattice points whereas Theorem 5 concludes only that the total number of visits in time N to all points in $\bigcup_{i \in I} \{(b, c) + ia\} \sim N(c - b)/a$ where $b < c < b + a$.

This completes our statements of our main visitation results. Their proofs will be given in § II. In § III we deal with two equivalence relations, and two partial orderings on rules, and prove related results some of which were stated earlier in this introduction. In § IV, which consists mainly of counterexamples,

we show some of the extent to which triviality of the tail σ -field fails for ruled sums. In § V we give extensions of some of our main visitation results to Gaussian processes more general than ruled sums of independent normal $(0, 1)$ variables.

II. Proofs of visitation results. We first gather together for easy reference a collection of basic results that will be essential to the proofs of Theorems 1–4 and Corollaries 1, 2.

Lemma 0. Let $\{Y_n\}$ be an arbitrary sequence of random variables with $E(Y_n^2) \leq c_n$ for all $n \in I^+$. If there exists a subsequence $\{n(k)\}$ of I^+ with

$$(a) \sum_k c_{n(k)} < \infty,$$

$$(b) \max_{n(k) \leq n \leq n(k+1)} |Y_n - Y_{n(k)}| \xrightarrow{k} 0 \text{ a.e.}$$

then $Y_n \rightarrow 0$ a.e.

Proof. Standard use of the Borel-Cantelli lemma and the Markov inequality.

In the rest of this collection of basic results we let $A, A', B, B' \subseteq I^+$ with $C \equiv A \cap B$, $C' \equiv A' \cap B'$ and $a \equiv |A|$, $a' \equiv |A'|$, etc. We assume without loss of generality that $a \geq b$ and $a' \geq b'$. For $D \subseteq I^+$, $S_D \equiv \sum_{i \in D} X_i$. Further, we let $d, f, b, i, l \in I$ and $k, m, n \in I^+$.

Symmetrization. An often used technique in probability theory is the symmetrization of a random variable Y [12, p. 245]. One considers a random variable Y' independent of but identically distributed to Y , and defines the symmetric random variable $Y^s = Y - Y'$, to be the symmetrization of Y . Similarly, if $Y_1, Y_2, \dots$ are random variables then $S_k^s \equiv \sum_{i=1}^k Y_i^s$ where the Y_i^s are independent of each other and of the $\{Y_j\}$ sequence. If $\hat{f}(\theta)$ is the characteristic function of Y then $|\hat{f}(\theta)|^2$ is the characteristic function of Y^s .

Needed results about the characteristic function $\hat{f}(\theta)$ of a symmetric random variable X . If X is the symmetrization of a random variable Y , then clearly $\hat{f}(\theta) \geq 0$ for all $\theta \in (-\pi, \pi]$, but this does not hold for a general symmetric X . Thus we define $\epsilon_0 \equiv \max\{\epsilon \in (0, \pi]: \hat{f}(\theta) \geq 0 \text{ for } |\theta| \leq \epsilon\}$ and $s_0 \equiv \sup\{|\hat{f}(\theta)|: \theta \in (-\pi, -\epsilon_0) \cup (\epsilon_0, \pi)\}$ where $s_0 = 0$ if $\epsilon_0 = \pi$.

That ϵ_0 and s_0 are well defined is contained in

(C1) $\hat{f}(\theta)$ is real valued, continuous at 0, $\hat{f}(0) = 1$ (i.e., $\epsilon_0 > 0$) and $s_0 < 1$ (because X_1 satisfies (ω)).

By the inversion formula for characteristic functions:

$$(C2) P[S_n = l] = (2\pi)^{-1} \int_{-\pi}^{\pi} \cos(l\theta) \hat{f}^n(\theta) d\theta.$$

By the definition of s_0 , we see $(2\pi)^{-1} \int_{-\pi}^{-\epsilon_0} |\hat{f}^n(\theta)| d\theta + \int_{\epsilon_0}^{\pi} |\hat{f}^n(\theta)| d\theta \leq s_0^n$ and so

$$P[S_n = 0] - P[S_n = l] \geq (2\pi)^{-1} \int_{-\epsilon_0}^{\epsilon_0} (1 - \cos(l\theta)) \hat{f}^n(\theta) d\theta - 2s_0^n \geq -2s_0^n.$$

Thus we have

(C3) $|P[S_n = l] - (2\pi)^{-1} \int_{-\epsilon_0}^{\epsilon_0} \cos(l\theta) \hat{f}^n(\theta) d\theta| < s_0^n$ and $P[S_n = 0] + 2s_0^n \geq P[S_n = l]$.

Let μ be the counting measure on I , then for $f, g \in \mathcal{L}^2(I, \mu)$

(C4) (Plancherel's theorem)

$$\sum_b f(b)g(b) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \hat{f}(\theta) \overline{\hat{g}(\theta)} d\theta$$

where $\hat{f}$ and $\hat{g}$ are the Fourier transforms of f and g .

Needed results from [4]. Let X_1 be a lattice variable satisfying (ω) and $E(X_1^+) = E(X_1^-)$ (where ∞ is allowed).

(CE1) $P[S_n = b]/P[S_{n-m} = l] \rightarrow 1$ (Theorem 4 of [4]).

(CE2) For $r > 0$, $\exists$ a constant K depending on r, ϵ , and $b - l$ but not on b or n : $\forall \epsilon > 0$

$$P[S_n = b] - P[S_n = l] \leq Kn^{-1/2+\epsilon} P[S_n = b] + n^{-r}.$$

In Theorem 3.2 of [4] Chung and Erdős do not claim K depends only on $b - l$ but their proof reveals this is so.

Needed results about ruled sums. We first state a corollary of Hölder's inequality:

(R0) Let $\{a_n\}$ be a sequence of positive numbers and let $\{i(l)\}$ and $\{j(l)\}$ be two 1-1 maps of I onto I . Then

$$(a) \sum_l a(l)a(i(l)) \leq \sum_l (a(l))^2,$$

$$(b) \sum_l a(l)a(i(l))a(j(l)) \leq \sum_l (a(l))^3.$$

(R1) If $a = b = a' = b'$ and $c \geq c'$ then $P[S_A = l, S_B = l] \geq P[S_{A'} = l, S_{B'} = l]$.

(In particular, $P[S_A = l, S_B = l] \geq P^2[S_a = l]$.)

Proof. Without loss of generality we assume $c = c' + 1$.

$$\begin{aligned} P[S_A = l, S_B = l] &= \sum_b P^2 \left[S_{|A|-|C|} = l - b \right] P[S_C = b] \\ &= \sum_b \sum_d P^2[S_{a-c} = l - b] P[X_1 = b - d] P[S_{c'} = d] \\ &= \sum_d P[S_{c'} = d] \sum_b P^2[S_{a-c} = l - b] P[X_1 = b - d] \\ &\geq \sum_d P[S_{c'} = d] \left[\sum_b P[S_{a-c} = l - b] P[X_1 = b - d] \right]^2 \\ &= \sum_d P[S_{c'} = d] P^2[S_{a-c'} = l - d] = P[S_{A'} = l, S_{B'} = l] \end{aligned}$$

where the inequality is the Schwarz inequality with respect to the probability

measure λ_d on I given by $\lambda_d(\{i\}) \equiv P[X_1 = i - d]$ for $i \in I$.

(R2) (a) Let $\mathcal{Q}$ be an event independent of $\{X_i\}$, $i \in A' \subseteq A$. If X_i is symmetric

$$P[S_A = d, \mathcal{Q}] \leq (P[S_{a'} = 0] + 2s_0^{a'})P[\mathcal{Q}].$$

(b) Let $\mathcal{Q}$ be an event independent of $\{X_i\}$, $i \in A' \cup B'$ where $A' \subseteq A$, $B' \subseteq B$, $A \cap B = \emptyset$ and $a' = b'$. Then

$$P[S_A - S_B = d, \mathcal{Q}] \leq P[S_{a'}^s = 0]P[\mathcal{Q}].$$

(c) Let $\mathcal{Q}$ be an event independent of $\{X_i\}$, $i \in A' \cup B' \cup C'$, where $A' \subseteq A$, $B' \subseteq B$, $C' \subseteq C$, A , B and C are pairwise disjoint and $a' = b' = c'$. Then

$$P[S_A - S_C = 0, S_B - S_C = 0, \mathcal{Q}] \leq P[S_{a'} - S_{c'} = 0, S_{b'} - S_{c'} = 0]P[\mathcal{Q}].$$

Proof. (a) By (C3)

$$\begin{aligned} P[S_A = d, \mathcal{Q}] &= \sum_b P[S_{a'} = d - b]P[S_{A-A'} = b, \mathcal{Q}] \\ &\leq (P[S_{a'} = 0] + 2s_0^{a'})P[\mathcal{Q}]. \end{aligned}$$

(b) Because $a' = b'$ and $A \cap B = \emptyset$, we see that $S_{A'} - S_{B'}$ is the sum of a' independent symmetrizations of X_1 . Hence by hypothesis and (C3)

$$\begin{aligned} P[S_A - S_B = d, \mathcal{Q}] &= \sum_b P[S_{a'}^s = d - b]P[S_A - S_B - (S_{A'} - S_{B'}) = b, \mathcal{Q}] \\ &\leq P[S_{a'}^s = 0]P[\mathcal{Q}]. \end{aligned}$$

(c)

$$\begin{aligned} &P[S_A - S_C = 0, S_B - S_C = 0, \mathcal{Q}] \\ &= \sum_{g, b} P[S_{A'} - S_{C'} = g, S_{B'} - S_{C'} = b] \\ &\quad \cdot P[S_A - S_C - (S_{A'} - S_{C'}) = -g, S_B - S_C - (S_{B'} - S_{C'}) = -b, \mathcal{Q}]. \\ &\leq \max_{g, b} P[S_{A'} - S_{C'} = g, S_{B'} - S_{C'} = b]P[\mathcal{Q}]. \end{aligned}$$

Since

$$P[S_{A'} - S_{C'} = g, S_{B'} - S_{C'} = b] = \sum_i P[S_{C'} = i]P[S_{A'} = g + i]P[S_{B'} = b + i],$$

(R2c) now follows from the hypothesis $a' = b' = c'$ and the case (b) of (R0).

(R3) Let $r > 0$ and $EX_1^+ = EX_1^-$. $\exists$ a constant K depending only on r and $l - d$: $\forall \epsilon > 0$

$$\begin{aligned}
& P[S_A = l, S_B = b] - P[S_A = d, S_B = b] \\
& \leq K(a - c)^{-1/2 + \epsilon} P[S_A = l, S_B = b] + (a - c)^{-r} P[S_B = b].
\end{aligned}$$

Proof. Recalling that in (CE2) the constant K depends only on the difference of the values, we have

$$\begin{aligned}
& P[S_A = l, S_B = b] - P[S_A = d, S_B = b] \\
& = \sum_f (P[S_{a-c} = l - f] - P[S_{a-c} = d - f]) P[S_C = f, S_B = b] \\
& \leq \sum_f (K(a - c)^{-1/2 + \epsilon} P[S_{a-c} = l - f] + (a - c)^{-r}) P[S_C = f, S_B = b] \\
& \leq K(a - c)^{-1/2 + \epsilon} P[S_A = l, S_B = b] + (a - c)^{-r} P[S_B = b].
\end{aligned}$$

This ends our list of basic results, and we begin to prove the visitation results; (1) of Theorem 1 is actually a special case of the following more general result:

Lemma 1. Let $\{Z_n\}$ be a sequence of d dimensional lattice random vectors, $\{B_n\}$ be a sequence of subsets of I^d , and $g: R^+ \rightarrow R^+$ with $g(x) \uparrow \infty$ and $g'(x)$ positive and $\downarrow$. If there is an $\epsilon \in (0, 1)$ so that

$$(\dagger') \quad \sum_{k=j}^n g'(M_k) (P[Z_k \in B_k | Z_j \in B_j] - P[Z_k \in B_k]) \leq K(g^\epsilon(M_n))$$

for all $j, n \in I^+$ and if $M_n \rightarrow \infty$, then

$$(1) \quad \lim g(M_n)^{-1} \sum_{i=1}^n g'(M_i) \phi[Z_i \in B_i] = 1 \quad a.s.$$

Furthermore, if

$$\overline{\lim} (1/g(M_n)) \sum_{k=j}^n g'(M_k) P[Z_k \in B_k | Z_j \in B_j] - P[Z_k \in B_k] \leq 0,$$

then

$$(2) \quad P[Z_j \in B_j \text{ i.o.}] = 1.$$

Proof. Let $L(n) \equiv \sum_{i=1}^n g'(M_i) \phi[Z_i \in B_i]$. Because $g'(x)$ is nonincreasing, we may use the standard approximation of sequences by the Riemann integral to obtain $E(L(n)) = O(1) + g(M_n)$, i.e., since $g(x) \uparrow \infty$ and $M_n \uparrow \infty$, $E(L(n)) \sim g(M_n)$.

Letting $V_n \equiv E[(g(M_n))^{-1} L(n) - 1]^2$, we have

$$\begin{aligned}
g^2(M_n)V_n &\leq O(g(M_n)) \\
&+ 2 \sum_{k>j=1}^n g'(M_j)g'(M_k)(P[Z_k \in B_k, Z_j \in B_j] - m_k m_j) \\
&= O(g(M_n)) \\
&+ 2 \sum_{j=1}^n g'(M_j)m_j \sum_{k=j+1}^n g'(M_k)(P[Z_k \in B_k | Z_j \in B_j] - m_k) \\
&= O(g^{1+\epsilon}(M_n)).
\end{aligned}$$

Thus $V_n = O(g^{-1+\epsilon}(M_n))$.

Proof of (1). We take $Y_n \equiv (g(M_n))^{-1}L(n) - 1$, $c_n \equiv (g(M_n))^{-1+\epsilon}$ and $n(k)$ so that $g(M_{n(k)})^{1-\epsilon} \sim k^2$ in Lemma 0. Then $\sum_k c_{n(k)} < \infty$ is clear and so $Y_{n(k)} \xrightarrow[k]{k} 0$ a.e. But the choice of $n(k)$ gives $g(M_{n(k)})/g(M_{n(k+1)}) \xrightarrow[k]{k} 1$ and so

$$\begin{aligned}
\max_{n(k) \leq n \leq n(k+1)} |Y_n - Y_{n(k)}| &\leq (g(M_{n(k)}))^{-1}(L_{n(k+1)} - L_{n(k)}) \\
&+ L_{n(k)}(g(M_{n(k+1)}) - g(M_{n(k)}))(g(M_{n(k)}))^{-2} \\
&\xrightarrow[k]{k} 0 \quad \text{a.e.}
\end{aligned}$$

Thus an application of Lemma 0 establishes (1).

Proof of (2). We have

$$\begin{aligned}
V_n &\leq O(g(M_n)^{-1}) \\
&+ \left(\frac{1}{g(M_n)} \sum_{j=1}^n g'(M_j)m_j \right) \frac{1}{g(M_n)} \sum_{k=j+1}^n g'(M_k)(P[Z_k \in B_k | Z_j \in B_j] - m_k) \rightarrow 0.
\end{aligned}$$

Choose $n(k)$ so that $V_{n(k)} \leq k^{-2}$. Then $(g(M_{n(k)}))^{-1}L(n(k)) - 1 = \xrightarrow[k]{k} 0$ a.e. by the Borel-Cantelli lemma. But $g(x) \uparrow \infty$ and so continuing as in the remarks preceding Corollary 2 in § I, the proof of Lemma 1 is completed.

In this paper we are primarily interested, as in (1) of Theorem 1, in the case where $B_n \equiv \{a\}$ for all n and some $a \in I^k$. (1) of Proposition 1 is a special case of

Lemma 2. Let $\{Z_k\}$ be a sequence of random variables and $g \in R^+ \rightarrow R^+$ satisfy $g' > 0$, $g' \downarrow$ and $xg'(x) = O(g^{\epsilon_0}(x))$ for some $\epsilon_0 \in (0, 1)$.

Let $M_n \rightarrow \infty$. Assume

(a) $\exists c > 0$ and $v < 1$: $\forall j, k \in I^+$ with $j \leq k$, $P[Z_k = a | Z_j = a] \leq P[Z_{k-j} = 0] + cv^{k-j}$.

(b) *There is a sequence of i.i.d. random variables $\{Y_j\}$ satisfying (ω) with $E(Y_1^{-1})$ such that $P[Z_k = 0] = P[\sum_{j=1}^k Y_j = 0]$ for all k .*

Then $(\dagger')$ holds for $B_n = \{a\}$.

If the Z_k are d -dimensional random vectors and a is the d -dimensional 0 vector, then $(\dagger')$ holds under the weaker hypothesis obtained by deleting (b).

Proof. By hypothesis (a) and the fact that $g'(x)$ is bounded,

$$\begin{aligned} Q_n &\equiv \sum_{k=j+1}^n g'(M_k)(P[Z_k = a | Z_j = a] - m_k) \\ &\leq \sum_{k=1}^{n-j} g'(M_{k+j})m_k(0) - \sum_{k=j+1}^n g'(M_k)m_k + O(1). \end{aligned}$$

By the hypothesis on $g(x)$, hypothesis (b) and (ρ) (i.e., $M_n(a)/M_n(a') = O(1)$) we see

$$(*) \quad \sum_{k=1}^j g'(M_{k+j})m_k(0) \leq g'(M_j)m_j(0) = O(g^{\epsilon_0}(M_j)).$$

Now let $\{b_k\}$ be a sequence in R^+ with $b_k \leq 1$ for all k , let $r > \delta > 0$, then by Hölder's inequality,

$$\begin{aligned} (H) \quad \sum_{k=1}^n k^{-r} b_k &\leq \left(\sum_{k=1}^n k^{-r/(r-\delta)} \right)^{r-\delta} \left(\sum_{k=1}^n b_k^{1/(1-r+\delta)} \right)^{1-r+\delta} \\ &= O \left(\left(\sum_{k=1}^n b_k \right)^{1-r+\delta} \right). \end{aligned}$$

We then have

$$\begin{aligned} (**) \quad \sum_{k=j+1}^{n-j} g'(M_{k+j})m_k(0) - \sum_{k=j+1}^n g'(M_k)m_k &\leq \sum_{k=j+1}^n g'(M_k)(m_k(0) - m_k) \quad \text{since } g'(x) \downarrow, \\ &\leq A \sum_{k=j+1}^n g'(M_k)m_k(0)k^{-1/2+\epsilon'} \quad \text{for any } \epsilon' > 0 \text{ by (CE2)} \\ &\quad \text{and hypothesis (b),} \\ &= O \left(\sum_{k=j+1}^n g'(M_k)m_k(0) \right)^{1/2+\epsilon''} \quad \text{for any } \epsilon'' > \epsilon' \text{ by (H),} \\ &\leq O(g^{1/2+\epsilon''}(M_n)). \end{aligned}$$

For the last inequality we have used the standard discrete approximation to the Riemann integral and the fact g' is nonincreasing. In the application of (H) we have used the fact that g' is bounded and $m_k(0) \rightarrow 0$ so for all large k $g'(M_k)m_k(0) \leq 1$.

Since ϵ' may be arbitrarily close to 0 and ϵ'' arbitrarily close to ϵ' , any $\epsilon > \max(\frac{1}{2}, \epsilon_0)$ suffices for $(\dagger')$.

Finally, when the Z_k are d -dimensional but a is the d -dimensional 0 vector, the proof becomes much simpler since only the hypothesis on g implies $(*)$ and that the expression $(**)$ is ≤ 0 .

Proof of (2) of Theorem 1. Let $D(l, n) \equiv \sum_{k=l}^n \phi[S_{(k)} = a] - \sum_{k=l}^n \phi[S_{(k)} = a']$ with $D_n \equiv D(1, n)$, and let $M_{l,n} \equiv \sum_{i=l}^n m_i$. By (R3) with $r = 3$

$$E[D^2(l, n)] = O(M_{l,n})$$

$$\begin{aligned} & + 2 \sum_{l \leq j < k \leq n} \{ (P[S_{(k)} = a, S_{(j)} = a] - P[S_{(k)} = a', S_{(j)} = a]) \\ & \quad + (P[S_{(k)} = a', S_{(j)} = a'] - P[S_{(k)} = a, S_{(j)} = a']) \} \\ & = O(M_{l,n}) + O \left[\sum_{l \leq j < k \leq n} \{ (k-j)^{-\frac{1}{2}+\epsilon'} (P[S_{(k)} = a, S_{(j)} = a] \right. \\ & \quad \left. + P[S_{(k)} = a', S_{(j)} = a']) \right. \\ & \quad \left. + (k-j)^{-3} P[S_j \in \{a, a'\}] \} \right] \quad \forall \epsilon' > 0 \\ & = O(M_{l,n}) + O(M_{l,n}) + \sum_{j=l}^n m_j(a) \sum_{k=l}^n k^{-\frac{1}{2}+\epsilon'} P[S_{(k)} = a | S_{(j)} = a] \\ & \quad + \sum_{j=l}^n m_j(a') \sum_{k=l}^n k^{-\frac{1}{2}+\epsilon'} P[S_{(k)} = a' | S_{(j)} = a'] \quad \forall \epsilon' > 0. \end{aligned}$$

Applying (H) and the hypothesis $(\dagger\dagger)$ of (2) to the last inequality gives

$$\begin{aligned} (*) \quad E[D^2(l, n)] & \leq K(M_{l,n} + M_{l,n}(M_n^{1+2\epsilon})^{\frac{1}{2}+\epsilon''}) \quad (\forall \epsilon'' > 0) \\ & \leq K(M_{l,n} M_n^{\frac{1}{2}+\epsilon+\epsilon''}) \quad (\forall \epsilon'' > 0). \end{aligned}$$

An inequality that appears in Billingsley [2, p. 102] says that if $\{Y_i\}$ is a sequence of random variables for which $\exists$ a positive sequence $\{\mu_i\}$ and $\beta, \nu \geq 1$:

$$(**) \quad E \left(\left| \sum_{i=l}^n Y_i \right|^\nu \right) \leq \left(\sum_{i=l}^n \mu_i \right)^\beta \quad \forall l < n$$

then

$$E \left(\max_{l \leq k \leq n} \sum_{i=l}^k Y_i \right)^\nu \leq \lg_2^\nu(4n) \left(\sum_{i=l}^n \mu_i \right)^\beta.$$

If we take $\mu_i \equiv Km_i M_n^{1/2 + \epsilon + \epsilon'}$, $\nu = 2$ and $\beta = 1$ then (*) shows that (**) holds for $Y_i = \phi[S_i = a] - \phi[S_i = a']$. Noting that the hypothesis on X_1 in (2) implies $n^{1-\alpha} = O(M_n)$ we obtain

$$\begin{aligned} E \left(\max_{l \leq k \leq n} D_k \right)^2 &= O(\lg_2^2(4n) M_n^{3/2 + \epsilon + \epsilon''}) \quad (\forall \epsilon'' > 0) \\ &= O(M_n^{3/2 + \epsilon + \epsilon''}) \quad (\forall \epsilon'' > 0). \end{aligned}$$

Fix ϵ'' and choose $n(k)$ so that $M_{n(k)} \sim k^\lambda$ where $\lambda > (\epsilon'')^{-1}$. Then $M_{n(k+1)}/M_{n(k)} \sim 1$ and so

$$E \left(\max_{m(k) \leq n \leq n(k+1)} D_n / M_n^{(3/4) + (\epsilon/2) + \epsilon''} \right)^2 = O(M_{n(k)}^{-\epsilon''}) = O(k^{-\lambda \epsilon''}).$$

Since $\sum_k k^{-\lambda \epsilon''} < \infty$, the Markov inequality and the Borel-Cantelli lemma imply

$$\max_{n(k) \leq n \leq n(k+1)} (D_n / M_n^{(3/4) + (\epsilon/2) + \epsilon''}) \xrightarrow[k \rightarrow \infty]{} 0 \quad \text{a.e.}$$

Proof of (2) of Proposition 1. By (R2) we see

$$\sum_{k=j+1}^n P[S_{(k)} = a | S_{(j)} = a'] \leq \sum_{k=j+1}^n m_{k-j}(0) + 2s_0^{k-j} = O(M_n)$$

which clearly means (††) holds with $\epsilon = 0$.

Proof of (3) of Theorem 1. Let $\delta \in (0, 1)$. For $k \in I^+$ define $n(k) = \max\{j: M_j \leq M_k^\delta\}$. Because $m_j < 1$, we see $M_{n(k)} \sim M_k^\delta$. By hypothesis, (†) holds for $g(x) = \lg x$ and thus by (1) of Theorem 1

$$\lim_k (\lg M_k)^{-1} \sum_{i=1}^{n(k)} \phi[S_{(i)} = a] / M_i = \delta.$$

Therefore

$$\begin{aligned} 1 - \delta &= \lim_k (\lg M_k)^{-1} \sum_{i=n(k)+1}^k \phi[S_{(i)} = a] / M_i \\ &\leq \varliminf_k \left(\sum_{i=1}^n \phi[S_{(i)} = a] \right) / M_k^\delta \lg M_k, \end{aligned}$$

i.e.,

$$P \left[\sum_{i=1}^k \phi[S_{(i)} = a] < M_k^{\delta'} \text{ i.o.} \right] = 0 \quad \text{for all } \delta' \in (0, 1).$$

This, combined with (2) of Theorem 1 yields (3) of Theorem 1.

Proof of Theorem 2. We will first prove (1) $\Rightarrow$ (2), (3) $\Rightarrow$ (1) and (2) $\Rightarrow$ (3) establishing the equivalence of (1), (2) and (3) which do not involve statements about sample sequences of ruled sums. (4a) $\Rightarrow$ (4b), (5a) $\Rightarrow$ (5b) and (6a) $\Rightarrow$ (6b) are obvious since the (b) statements are weakened forms of the (a) statements in which $\forall$ is replaced by $\exists$. We then show that (4b) $\Rightarrow$ (2), (5b) $\Rightarrow$ (3), (6b) $\Rightarrow$ (2) follow from the monotone convergence theorem and the Borel-Cantelli lemma. Next we show that (6a) $\Rightarrow$ (4a) and (5a) by a Fubini's theorem argument. Finally we prove that (1) $\Rightarrow$ (6a) which completes the demonstration of equivalence of the nine conditions of Theorem 2. The proof of (1) $\Rightarrow$ (6a) relies on tools developed earlier, in particular Lemma 1 and Lemma 2.

(1) $\Leftrightarrow$ (2). By (C4),

$$\sum_n \sum_a P^2[S_n = a] = (2\pi)^{-1} \sum_n \int_{-\pi}^{\pi} |\hat{f}(\theta)|^{2n} d\theta.$$

(3) $\Rightarrow$ (1). By (C2),

$$\sum_n P[S_n = d_n] \leq (2\pi)^{-1} \sum_n \int_{-\pi}^{\pi} |\hat{f}(\theta)|^n d\theta = (2\pi)^{-1} \int_{-\pi}^{\pi} (1 - |\hat{f}(\theta)|)^{-1} d\theta.$$

(2) $\Rightarrow$ (3). $\forall n$ let $d_n: P[S_n = d_n] = \max_a P[S_n = a]$. Then

$$\sum_n P[S_n = d_n] \geq \sum_n \sum_a P^2[S_n = a] = \infty.$$

(4a) $\Rightarrow$ (4b), (5a) $\Rightarrow$ (5b), (6a) $\Rightarrow$ (6b). Obvious.

(4b) $\Rightarrow$ (2). By the monotone convergence theorem

$$\infty = E \left(\sum_n m_n(S_{(n)}) \right) = \sum_n \sum_a P^2[S_n = a].$$

(5b) $\Rightarrow$ (3). By the Borel-Cantelli lemma

$$P[S_{(n)} - d_n = a \text{ i.o.}] > 0 \Rightarrow \sum_n P[S_{(n)} = d_n + a] = \infty.$$

(6b) $\Rightarrow$ (2). By the Borel-Cantelli lemma

$$P[S_{(n)_1} = S_{(n)_2} + a \text{ i.o.}] > 0 \Rightarrow \sum_n P[S_{(n)_1} = S_{(n)_2} + a] = \infty.$$

Since the variables $S_{(n)_1}$ and $S_{(n)_2}$ are independent

$$\infty = \sum_n P[S_{(n)_1} = S_{(n)_2} + a] = \sum_n \sum_b m_n(a+b)m_n(b) \leq \sum_n \sum_b m_n^2(b)$$

by (a) of (R0).

(6a) $\Rightarrow$ (4a) and (5a). Consider the maps σ_1 and σ_2 defined by $\sigma_1(n) \equiv 2n + 1$ and $\sigma_2(n) \equiv 2n$ for all $n \in I^+$. Then by Theorem 6 in § III (which is independent of § II), we see the rules $(\cdot)_1$ and $(\cdot)_2$ gotten by $(n)_i = \bigcup_{j \in (n)} \sigma_i(j)$, $i = 1, 2$, define two sequences of ruled sums equivalent to the sequence of ruled sums defined by $(\cdot)$. These two rules, $(\cdot)_1$ and $(\cdot)_2$ have the property demanded in (6), i.e., $(n)_1 \cap (m)_2 = \emptyset$ for $n, m \in I^+$, and thus

$$(*) \quad P[S_{(n)_1} = S_{(n)_2} + a, \text{ i.o.}] = 1 \quad \text{for all } a \in I.$$

We now use a standard Fubini argument often associated with symmetrization ([12, p. 245] or [7, p. 7]). Consider the sample spaces Ω_1 and Ω_2 generated by $\bigcup_{i \in \bigcup_{n \in (n)_1} \{X_i\}}$ and $\bigcup_{i \in \bigcup_{n \in (n)_2} \{X_i\}}$ respectively. Then the total sample space Ω could be taken as $\Omega_1 \times \Omega_2$ and the probability P on Ω could be considered to be $P_1 \times P_2$ where P_1 and P_2 are P restricted to Ω_1 and Ω_2 respectively. For $a \in I$, we define

$$A_a \equiv [\omega \in \Omega_2: P_1[S_{(n)_1} = S_{(n)_2}(\omega) + a \text{ i.o.}] = 1].$$

By (*) and the Fubini theorem we see that $P[A_a] = P_2[A_a] = 1$ for all $a \in I$.

For $a = 0$, (4) follows for $\omega \in A_a$ by the easy half of the Borel-Cantelli lemma.

I is a countable and so $P[\bigcap_a A_a] = 1$. Letting $\omega \in \bigcap_a A_a$ and $d_n \equiv S_{(n)_2}(\omega)$, (5) follows.

(1) $\Rightarrow$ (6a). In Lemma 2 we let $Z_j \equiv S_{(j)_2} - S_{(j)_1}$, i.e., Z_j is the sum of j independent symmetrizations of X_1 . Thus (b) of Lemma 2 is satisfied.

Letting $A \equiv (k)_1, B \equiv (k)_2, A' \subseteq (k)_1 - (k)_1 \cap (j)_1, B' \subseteq (k)_2 - (k)_2 \cap (j)_2$ with $|A'| = |B'| = k - j$ and letting $\mathcal{A} \equiv [Z_j = a]$ in (R2b) we see

$$P[Z_k = a, Z_j = a] \leq P[Z_{k-j} = 0]P[Z_j = a]$$

i.e., $P[Z_k = a | Z_j = a] \leq P[Z_{k-j} = 0]$ which shows (a) of Lemma 2 is satisfied.

(1) implies $\int_{-\pi}^{\pi} (1 - |\hat{f}(\theta)|^2)^{-1} d\theta = \infty$ and so by (C2), $M_n \xrightarrow[n]{\infty}$.

So letting $g(x) = \lg x$ in Lemmas 1 and 2 we see (1) implies (6a). In fact by (1) of Lemma 1 we have a "rate of occurrence" for these meetings $S_{(n)_1} = S_{(n)_2} + a$.

Proof of Theorem 3. In Lemma 2, we let $Z_n \equiv (Y_n, V_n) \equiv (S_{(n)_1} - S_{(n)_3}, S_{(n)_2} - S_{(n)_4})$. Thus $P[(S_{(n)_1}, S_{(n)_2}) = (S_{(n)_3}, S_{(n)_4}) \text{ i.o.}] = P[Z_n = (0, 0) \text{ i.o.}]$.

Using (R2b),

$$\begin{aligned} P[Z_n = (0, 0), Z_j = (0, 0)] &= P[Y_n = 0, Y_j = 0]P[V_n = 0, V_j = 0] \\ &\leq P[Y_{n-j} = 0]P[Y_j = 0]P[V_{n-j} = 0]P[V_j = 0] \\ &= P[Z_j = (0, 0)]P[Z_{n-j} = (0, 0)]. \end{aligned}$$

Thus (a) of Lemma 2 is satisfied and so for $g(x) = \lg x$, Lemma 1 and Lemma 2 show that $P[Z_n = (0, 0) \text{ i.o.}] = 1$ iff $\sum_n P[Z_n = (0, 0)] = \infty$. However, by (C4),

$$\begin{aligned} \sum_n P[Z_n = (0, 0)] &= \sum_n \left(\sum_a P^2[S_n = a] \right)^2 \\ &= (2\pi)^{-2} \sum_n \int_{-\pi}^{\pi} \int |\hat{f}(\theta)|^{2n} |\hat{f}(\phi)|^{2n} d\theta d\phi = \infty \\ &\text{iff } \int_{-\pi}^{\pi} \int (1 - |\hat{f}(\theta)| |\hat{f}(\phi)|)^{-1} d\theta d\phi = \infty. \end{aligned}$$

We have proved (1), (2) and (3) of Theorem 3 are equivalent.

We now show (4) is equivalent to (1) and (2). First, by the Schwarz inequality,

$$\sum_n \left(\sum_a m_n^2(a) \right)^2 \leq \sum_n \left(\sum_a m_n^3(a) \right) \left(\sum_a m_n(a) \right) = \sum_a \sum_n m_n^3(a).$$

On the other hand, since the Fourier transform of a product of two functions is the convolution of their Fourier transforms,

$$\sum_a m_n^3(a) = (2\pi)^{-2} \iint \hat{f}^n(\theta) \overline{\hat{f}^n(\theta - \phi)} \overline{\hat{f}^n(\phi)} d\theta d\phi$$

by Plancherel's theorem (C4). Since $|\overline{\hat{f}^n(\theta)} \overline{\hat{f}^n(\theta - \phi)} \overline{\hat{f}^n(\phi)}| \leq |\hat{f}^n(\theta) \hat{f}^n(\phi)|$,

$$\begin{aligned} &\left| \sum_n \iint \overline{\hat{f}^n(\theta)} \overline{\hat{f}^n(\theta - \phi)} \overline{\hat{f}^n(\phi)} d\theta d\phi \right| \\ &\leq \sum_n \iint |\hat{f}^n(\theta)| |\hat{f}^n(\phi)| d\theta d\phi = \iint (1 - |\hat{f}(\theta)| |\hat{f}(\phi)|)^{-1} d\theta d\phi. \end{aligned}$$

Hence

$$\sum_n \sum_a m_n^3(a), \quad \sum_n \left(\sum_a m_n^2(a) \right)^2 \quad \text{and} \quad \iint (1 - |\hat{f}(\theta) \hat{f}(\phi)|)^{-1} d\theta d\phi$$

are infinite together completing the proof that (1)–(4) are equivalent.

We now show (5) is equivalent to (4) to complete the proof of Theorem 3. In Lemma 2 we let $Z_n \equiv (S_{(n)_1} - S_{(n)_3}, S_{(n)_2} - S_{(n)_3})$. Thus

$$\begin{aligned}
& P[(S_{(n)_1} = S_{(n)_2} = S_{(n)_3} \text{ i.o.}] \\
& = P[(S_{(n)_1} - S_{(n)_3} = 0, S_{(n)_2} - S_{(n)_3} = 0) \text{ i.o.}] = P[Z_n = (0, 0) \text{ i.o.}].
\end{aligned}$$

Let X, X', X'' be independent and distributed like X_1 . To show (a) of Lemma 2 is satisfied, we take, in (R2c), $\mathcal{Q} = [Z_j = (0, 0)]$, $A = (n)_1$, $B = (n)_2$, $C = (n)_3$, $A' \subseteq (n)_1 - (j)_1$, $B' \subseteq (n)_2 - (j)_2$, $C' \subseteq (n)_3 - (j)_3$ with $|A'| = |B'| = |C'| = n - j$. By the hypothesis of Theorem 3, A, B and C are pairwise disjoint and A', B' and C' are independent of $\mathcal{Q}$. By (R2c), the definition of Z_n and the first sentence of this paragraph

$$P[Z_j = (0, 0), Z_n = (0, 0)] \leq P[Z_j = (0, 0)]P[Z_{n-j} = (0, 0)];$$

i.e., (a) of Lemma 2 is satisfied. For $g(x) = \lg x$, Lemma 1 and Lemma 2 now show that $P[Z_n = (0, 0) \text{ i.o.}] = 1$ iff $\sum_n P[Z_n = (0, 0)] = \infty$, i.e., iff $\sum_n \sum_a m_n^3(a) = \infty$ by definition of Z_n . This completes the proof of the equivalence of (1)–(5) of Theorem 3. (1) of Lemma 1 actually establishes a "rate of occurrence" for these triple meetings $S_{(n)_1} = S_{(n)_2} = S_{(n)_3}$.

The conditions in Lemma 2 yield a rate of visitation in Theorem 1. For Theorem 4, the variables are not necessarily lattice and we are only interested in infinite visitation so we use a somewhat different method based on Lemma 3.

Lemma 3. Let $\{X_n\}$ be a sequence of independent random variables, let $\{B_n\}$ be a sequence of open intervals in R , and let $\{f_n\}$ be a sequence of functions $\times_{i=1}^{\infty} R_i \rightarrow R$. Let $Z_n \equiv f_n(X_1, \dots, X_n, \dots)$ and for $x_1, \dots, x_m \in R$ let $Y_n \equiv Y_n(x_1, \dots, x_m) \equiv f_n(x_1, \dots, x_m, X_{m+1}, \dots)$. If $\exists A > 0: \forall m \in I^+, x_1, \dots, x_m \in R$

- (a) $\exists \Gamma > 0$ and $n_0: \forall n > n_0 \quad P[Y_n \in B_n] \geq \Gamma P[Z_n \in B_n]$,
 (b) $\exists$ a subsequence $n(k)$ of I^+ :

$$\max_{1 \leq j < n(k)} \sum_{l=j+1}^{n(k)} P[Y_l \in B_l | Y_j \in B_j] \leq A \sum_{l=1}^{n(k)} P[Y_l \in B_l]$$

then

$$P[Z_j \in B_j \text{ i.o.}] = \frac{1}{0} \text{ iff } \sum P[Z_j \in B_j] = \infty.$$

Proof. Since $\sum P[Z_j \in B_j] < \infty \Rightarrow P[Z_j \in B_j = 0 \text{ i.o.}] = 0$ by the Borel-Cantelli lemma we assume $\sum P[Z_j \in B_j] = \infty$. By (a) this implies $\sum P[Y_j \in B_j] = \infty$. Hence

$$\begin{aligned}
 (*) \quad & P \left[\overline{\lim} \left(\sum_{j=1}^n \phi[Y_j \in B_j] / \sum_{j=1}^n P[Y_j \in B_j] \right) > 0 \right] \\
 & \leq P[Y_j \in B_j \text{ i.o.}] = P[Z_j \in B_j \text{ i.o.} | X_1 = x_1, \dots, X_m = x_m]
 \end{aligned}$$

since the X_i are independent.

Kochen and Stone [9] showed that for any sequence of random variables W_j

$$\begin{aligned}
 & P \left[\overline{\lim} \sum_{j=1}^n \phi[W_j \in B_j] / \sum_{j=1}^n P[W_j \in B_j] > 0 \right] \\
 & \geq \overline{\lim} \left(\sum_{j=1}^n P[W_j \in B_j] \right)^2 / \sum_{i=1, j=1}^n P[W_i \in B_i, W_j \in B_j].
 \end{aligned}$$

Taking $W_j \equiv Y_j$ and letting $N_n = \sum_{j=1}^n P[Y_j \in B_j]$, (*) says

$$P[[Z_j \in B_j \text{ i.o.}] | X_1 = x_1, \dots, X_m = x_m] \geq \overline{\lim} N_n^2 / \sum_{i,j=1}^n P[Y_i \in B_i, Y_j \in B_j]$$

while (b) gives

$$\begin{aligned}
 & \sum_{i,j=1}^{n(k)} P[Y_i \in B_i, Y_j \in B_j] \\
 & = o(N_{n(k)}^2) + 2 \sum_{j=1}^{n(k)} P[Y_j \in B_j] \sum_{i=j+1}^{n(k)} P[Y_i \in B_i | Y_j \in B_j] \leq o(N_{n(k)}^2) + 2AN_{n(k)}^2.
 \end{aligned}$$

Thus $\forall m \in I^+, \forall x_1, \dots, x_m \in R$, $P[[Z_j \in B_j \text{ i.o.}] | X_1 = x_1, \dots, X_m = x_m] \geq (1/2A)$. By a corollary of the martingale convergence theorem [3, p. 93],

$$P[[Z_j \in B_j \text{ i.o.}] | X_1, \dots, X_m] \xrightarrow{m} \phi(Z_j \in B_j \text{ i.o.})$$

a.e. Now $\phi[Z_j \in B_j \text{ i.o.}]$ is an indicator function and thus takes only the values 0 or 1. However, we have shown ϕ to be $\geq (1/2A)$ a.e. and thus ϕ equals 1 a.e. The proof is complete.

Proof of Theorem 4. To prove this theorem, we first note the following:

Note 1. Since $a_n = o(n)$, we see that for $0 < \epsilon < |c|$, $\exists n_0 > 0$ such that $n > n_0 \Rightarrow [|S_{(n)}/a_n - d| < \epsilon] \subseteq [|S_{(n)} - nc|/n > \epsilon]$.

Note 2. A theorem of Katz [11] states: For $\alpha \geq 1$,

$$EX^\alpha < \infty \text{ iff } \sum_{k=1}^{\infty} k^{\alpha-2} P \left[\left| \frac{S_k - kEX_1}{k} \right| > \epsilon \right] < \infty.$$

Note 3. This note was pointed out to us by K. G. Binmore and M. Katz. A result of Stone [19] applied to nonlattice variables X_i with $E(X_i) = 0$, $E(X_i^2) = 1$ states $P[\sqrt{n}x \leq S_n \leq \sqrt{n}(x+b)] = (1/\sqrt{2\pi}) \int_x^{x+b} e^{-x^2/2} + o_n(1)(b + 1/\sqrt{n})$ where $o(1)$ is uniform in x, b . Define x, b by $\sqrt{n}x = b_n$, $\sqrt{n}(x+b) = c_n$. Then if $b(n), c(n) = o(n^{1/2})$ and $c(n) - b(n) \geq K > 0$ the Stone theorem yields $P[b(n) < S_n < c(n)] \sim (c(n) - b(n))(2\pi n)^{-1/2}$. (A similar result holds for lattice random variables (Gnedenko and Kolmogoroff [10, p. 233]).)

(4) $\Rightarrow$ (1) and (2) $\Rightarrow$ (1). In Note 1, let $c = EX_1$ and $d = 0$ if (4) holds, $d = b$ if (2) holds and then combine this with Note 2 and $\alpha = 2$ to get a contradiction.

(3) $\Rightarrow$ (2) and (3) $\Rightarrow$ (4) are easy. To complete the proof of equivalence we show (1) $\Rightarrow$ (3).

(1) $\Rightarrow$ (3). In Lemma 3, let $f_n(x_1, \dots, x_l, \dots) \equiv a_n^{-1} \sum_{i \in (n)} x_i$. Further let $m \in I^+$; $x_1, \dots, x_m \in R$; $Y_n \equiv Y_n(x_1, \dots, x_m)$, and $B_n \equiv (b - \epsilon, b + \epsilon)$ for all n . Let $A_n(b, \epsilon) \equiv [|Y_n(x_1, \dots, x_m) - b| < \epsilon] \equiv [Y_n \in B_n]$.

Now in Note 3, let $b(n) \equiv a_n(b - \epsilon) - \sum_{i \in (n) \cap m} x_i$ and $c(n) \equiv a_n(b + \epsilon) - \sum_{i \in (n) \cap m} x_i$. Since m is fixed, we see

$$P[Y_n \in B_n] = P[b(n) < S_{|(n) - m \cap (n)|} < c(n)] \sim 2 \epsilon a_n (2\pi n)^{-1/2}$$

(i.e., it is asymptotically independent of $x_1, \dots, x_m$ and b). Hence since a_n is nondecreasing we have

$$\begin{aligned} (*) \quad \sum_{k=j+1}^n P[A_k(b, \epsilon) | A_j(b, \epsilon)] &\leq \sum_{k=j+1}^n P[|S_{k-j} - b(a_k - a_j)| \leq (a_k + a_j)\epsilon] \\ &= O\left(\sum_{k=j+1}^n a_k(k-j)^{-1/2}\right) = O(a_n n^{1/2}), \end{aligned}$$

$$(**) \quad \sum_{k=1}^n P[A_k(b, \epsilon)] \geq \sum_{k=n/2}^n P[A_k(b, \epsilon)] \geq K \sum_{k=n/2}^n a_k k^{-1/2} \geq K a_{n/2} n^{1/2}.$$

Thus to finish the proof of Theorem 4, we combine (*) and (**) with Lemma 3 where $n(k)$ is a subsequence of I^+ with the property that $a_{n(k)} < C a_{n(k)/2}$ for some fixed positive constant C . Such a sequence $n(k)$ exists since even $\liminf a_{2k}/a_k > \sqrt{2} \Rightarrow \liminf a_{2^n}/(2^n)^{1/2} = \infty$ and this contradicts $a_n = o(n^{1/2})$.

Proof of Theorem 5. If $\hat{f}(2\pi t) \neq 1$,

$$\begin{aligned} E \left| \frac{1}{N} \sum_{n=1}^N e^{2\pi i t S_{(n)}} \right|^2 &= \frac{1}{N^2} \sum_{m=1}^N \sum_{n=1}^N E(e^{2\pi i t (S_{(n)} - S_{(m)})}) \\ &\leq \frac{1}{N^2} \sum_{m=1}^N \sum_{n=1}^N |f|^{n-m}(2\pi t) \leq \frac{N}{N^2} \frac{2}{1 - |f|(2\pi t)} = O\left(\frac{1}{N}\right). \end{aligned}$$

We conclude $(1/N) \sum_{n=1}^N e^{2\pi i t S(n)} \rightarrow 0$ a.e., by Lemma 0 with $N_k = k^2$, since

$$\begin{aligned} & \max_{N=N_k}^{N_{k+1}} \left| \frac{1}{N} \sum_{n=1}^N e^{2\pi i t S(n)} - \frac{1}{N_k} \sum_{n=1}^{N_k} e^{2\pi i t S(n)} \right| \\ &= \max_{N=N_k}^{N_{k+1}} \left| \frac{1}{N_k} \left(\sum_{n=1}^N - \sum_{n=1}^{N_k} \right) e^{2\pi i t S(n)} + \left(\frac{1}{N} - \frac{1}{N_k} \right) \sum_{n=1}^N e^{2\pi i t S(n)} \right| \\ &\leq (2k+1)/k^2 + 2/k = o(1). \end{aligned}$$

By the hypothesis, a.e., $\forall l \in I, N^{-1} \sum_{n=1}^N e^{2\pi i (l/a) S(n)} \rightarrow 0$. This is equivalent to $S(n)$ equidistributed mod $(0, a)$ a.e. [20].

III. Equivalence relations and partial orderings on rules. In this section we consider the question of when two rules may be considered equivalent from the viewpoint of ruled sums. In Theorem 6 we show that two natural definitions lead to the same equivalence relation. Theorem 6 is stated for lattice random variables for convenience. We then consider a weaker equivalence relation and associated partial ordering for Gaussian variables and prove in Theorem 7 an extension of a theorem of [1].

Theorem 6. Let $(\cdot)$ and $(\cdot)'$ be two rules. Then conditions (a) and (b) are equivalent.

(a) $\exists$ a 1-1 map σ from $\bigcup_{j=1}^{\infty} (j)$ onto $\bigcup_{i=1}^{\infty} (i)'$: $\forall n \in I^+ (n)' = \{\sigma(i), i \in (n)\}$,

(b) $\forall j, i_1, \dots, i_j \in I^+ |(i_1) \cap \dots \cap (i_j)| = |(i_1)' \cap \dots \cap (i_j)'|$.

$S(n), n = 1, 2, \dots$, and $S(n)', n = 1, 2, \dots$, define equivalent stochastic processes for all distributions $m(a) = P(X_1 = a)$ iff one and hence both of (a) and (b) hold.

Proof. That (a) $\Rightarrow$ (b) is clear since $\sigma\{(i_1) \cap \dots \cap (i_j)\} = \sigma((i_1)) \cap \dots \cap \sigma((i_j)) = (i_1)' \cap \dots \cap (i_j)'$.

To prove the converse, we first show that a mapping can always be defined if we restrict attention to any finite number n of sets $(i_1), \dots, (i_n)$.

Lemma 4. If $V_1, \dots, V_n$ are n finite sets, then the 2^n cardinalities $|V_{i_1} \cap \dots \cap V_{i_j} \cap \bigcup_{i=1}^n V_i|, j = 0, \dots, n$, and the 2^n cardinalities $|V_1^{\delta_1} \cap V_2^{\delta_2} \cap \dots \cap V_n^{\delta_n} \cap \bigcup_{i=1}^n V_i|$ where each δ_j varies over the two values complementation ($= 0$) or its absence ($= 1$) determine each other uniquely.

Proof. We denote $|V_{i_1} \cap \dots \cap V_{i_j}|$ by R_u where $u = \bar{u}^*$ is an n -long binary index $(u_1 \dots u_n)$ with $u_k = 1$ for $k = i_1, \dots, i_j$ and $u_k = 0$ otherwise. We

denote $|V_1^{\delta_1} \cap \dots \cap V_n^{\delta_n} \cap \bigcup_{i=1}^n V_i|$ by D_δ where $\delta = (\delta_1, \dots, \delta_n)$.

The 2^n vector R is clearly obtained from the 2^n vector D by a linear transformation A , $R = AD$, viz. $R_u = \sum_{u \leq \delta} D_\delta$, where $u \leq \delta$ is defined as $u_k \leq \delta_k$ for all $k = 1, \dots, n$. Thus D determines R uniquely. Conversely we show $\det A = 1$ so that $D = A^{-1}R$ and R determines D uniquely. For $n = 1$ $\det A = 1$. If the result is true for n and A_n is the $2^n \times 2^n$ matrix effecting the transformation from D to R , for $n + 1$ the $2^{n+1} \times 2^{n+1}$ matrix is

$$\begin{pmatrix} A_n & A_n \\ 0 & A_n \end{pmatrix}$$

whose determinant $= (\det A_n)^2 = 1$ completing the induction.

We now complete the proof that (b) $\Rightarrow$ (a) in Theorem 6. For any $m \in I^+$ the 2^m sets $(1)^{\delta_1} \cap \dots \cap (m)^{\delta_m}$, $\delta_i = 0, 1$, $i = 1, \dots, m$, are disjoint so by Lemma 4 we may define a 1-1 map τ from $\bigcup_{k=1}^m (k)$ onto $\bigcup_{k=1}^m (k)'$ preserving cardinalities of intersections of sets (i) , $i = 1, \dots, m$, by mapping the elements in $(1)^{\delta_1} \cap \dots \cap (m)^{\delta_m}$, 1-1 onto the elements in $(1)^{\delta_1'} \cap \dots \cap (m)^{\delta_m'}$ for $\delta_1, \dots, \delta_m = 0, 1$. In the set of all such mappings $\tau = \tau_m$ for any $m \in I^+$ we define $\tau_m < \tau_n^*$ if $m < n$ and $\tau_n^* = \tau_m$ on domain τ_m . For each $m = 1, 2, \dots$ there are only finitely many mappings τ_m and there exist mappings τ_n for arbitrarily large n . Hence we can choose an infinite chain $\tau_1 < \tau_2 < \dots$. For $i \in \bigcup_{k=1}^\infty (k)$ we may define $\tau(i) = \tau_k(i)$ where $i \in \text{dom } \tau_k$. τ necessarily maps $\bigcup_{k=1}^\infty (k)$ 1-1 onto $\bigcup_{k=1}^\infty (k)'$. Thus (b) $\Rightarrow$ (a).

If (a) and (b) hold for $()$ and $()'$, then for $P(X_1 = a) = m(a)$ for any j , $n_1 \dots n_j \in I^+$ and $a_1 \dots a_j \in I$ we have

$$\begin{aligned} P[S_{(n_1)} = a_1, \dots, S_{(n_j)} = a_j] &= P[S_{(n_1)'} = a_1, \dots, S_{(n_j)'} = a_j] \\ &= \sum_{(\mathbf{x}_{\delta} \geq v_i, x_{\delta} = a_i)_{i=1,2,\dots,j}} \prod_{\delta} m_{|\delta|}(x_{\delta}) \end{aligned}$$

where $\delta = \{\delta_k, k = 1, \dots, j\}$, $|\delta| = |\bigcap (n_k)^{\delta_k} \cap \bigcup_{k=1}^j (n_k)|$ and v_i is a j -vector with 1 in the i th place and 0's elsewhere. Since all finite-dimensional probabilities are equal, $S_{(n)}$ and $S_{(n)'}$ define equivalent processes. Conversely if (a) and (b) fail to hold it is clear we can find a distribution $m(a)$ and $n_1 \dots n_j$: $P[S_{(n_1)} = a_1, \dots, S_{(n_j)} = a_j] \neq P[S_{(n_1)'} = a_1, \dots, S_{(n_j)'} = a_j]$. This completes the proof of Theorem 6.

If either and hence both of (a) and (b) hold, we say $() \cong ()'$. A weaker equivalence relation $\simeq$ defined by $() \simeq ()'$ iff $\forall i, j \in I^+ | (i) \cap (j) | = | (i)' \cap (j)' |$ is sufficient when the underlying variable X_1 is Gaussian since the $S_{(j)}$

are jointly Gaussian and the covariances $E(S_{(i)}S_{(j)}) = |(i) \cap (j)|$ define the process.

An example of two non $\cong$ rules which are $\simeq$ is easy. Let $(2) = \{2, 3\}$, $(3) = \{3, 4, 5\}$, $(4) = \{2, 4, 6, 7\}$, $(2)' = \{2, 3\}$, $(3)' = \{3, 4, 5\}$, $(4)' = \{3, 6, 7, 8\}$ with $(n) = (n)'$ for $n \neq 2, 3, 4$ involving integers not appearing above. The pairwise intersections above have cardinality 1, but $|(2) \cap (3) \cap (4)| = 0 \neq 1 = |(2)' \cap (3)' \cap (4)'|$.

Corresponding respectively to the two equivalence relations $\cong$ and $\simeq$ on rules there are two natural partial orderings

$$(\) \leq (\)' \text{ iff } \forall j, i_1, \dots, i_j \in I^+ \quad |(i_1) \cap \dots \cap (i_j)| \leq |(i_1)' \cap \dots \cap (i_j)'|$$

convenient for general distribution $m(a)$ and

$$(\) \leq (\)' \text{ iff } \forall i, j \in I^+ \quad |(i) \cap (j)| \leq |(i)' \cap (j)'|.$$

We have at present no good theorem on $\leq$ although natural conjectures abound and so results should be forthcoming. The following results concern $\leq$.

Proof of Proposition (2) stated in introduction. If the 2×2 covariance matrix for $S_{(j)}$ and $S_{(n)}$ is $\leq$ the matrix for $S_{(j)'}$ and $S_{(n)'}$, then following Slepian's method [18]

$$\begin{aligned} (d/d\lambda)P_\lambda [S_{(j)} \in (u, v), S_{(n)} \in (u, v)] \\ = (g_\lambda(u, u) + g_\lambda(v, v) - g_\lambda(u, v) - g_\lambda(v, u))(ES_{(n)'}S_{(j)'} - ES_{(n)}S_{(j)}) \end{aligned}$$

(where P_λ and g_λ denote respectively probabilities and densities computed according to a covariance matrix with off diagonal entry $= (1 - \lambda)E(S_{(j)}S_{(n)}) + \lambda E(S_{(n)'}S_{(j)'})$). The proposition now follows from Lemma 5.

Lemma 5. Let $0 \leq c \leq \min(a, b)$. Then for all real u, v , $0 \leq e^{-(a+b-2c)u^2} + e^{-(a+b-2c)v^2} - e^{-(au^2+bu^2-2cuv)} - e^{-(av^2+bu^2-2cuv)}$.

Proof.

$$\begin{aligned} 0 &\leq [e^{-(a-c)u^2} - e^{-(a-c)v^2}][e^{-(b-c)u^2} - e^{-(b-c)v^2}] \\ &= e^{-(a+b-2c)u^2} + e^{-(a+b-2c)v^2} - e^{-((a-c)u^2+(b-c)v^2)} - e^{-((a-c)v^2+(b-c)u^2)}, \end{aligned}$$

because the two factors on the right of the inequality are both positive or both negative since $a - c \geq 0$, $b - c \geq 0$. Hence it is sufficient to show that

$$\begin{aligned} e^{-((a-c)u^2+(b-c)v^2)} + e^{-((a-c)v^2+(b-c)u^2)} \\ \geq e^{-(au^2+bu^2-2cuv)} + e^{-(av^2+bu^2-2cuv)}. \end{aligned}$$

However, the first term on the left is $\geq$ the first term on the right because

$$((a-c)u^2 + b-c)v^2 - (au^2 + bv^2 - 2cuv) = -c(u^2 + v^2 - 2uv) = -c(u-v)^2 \leq 0$$

since $c \geq 0$ and the same relationship holds between the second terms on the left and right completing the proof.

An extension of Proposition 2 to the joint distribution of $S_{(n)}$'s taken 3 at a time is false. In fact, let $S_{(1)} = X_1$, $S_{(2)} = X_2 + X_3$, $S_{(3)} = X_1 + X_2 + X_3$. Then for small positive ϵ

$$\begin{aligned} P[|S_{(1)}| < \epsilon, |S_{(2)}| < \epsilon, |S_{(3)}| < \epsilon] \\ \geq P[0 \leq X_1 < \epsilon/2]P[0 \leq X_2 + X_3 < \epsilon/2] \sim (1/2\pi)(\epsilon/2)^2(1/\sqrt{2}) \end{aligned}$$

while

$$\begin{aligned} P[|S_1| < \epsilon, |S_2| < \epsilon, |S_3| < \epsilon] \\ \leq P[|X_1| < \epsilon]P[|X_2| < 2\epsilon]P[|X_3| < 2\epsilon] \sim (1\sqrt{2\pi})^3(2\epsilon)(4\epsilon)(4\epsilon) \\ = (32/\sqrt{2\pi}^3)\epsilon^3 < (\epsilon/2)^2/(2\pi\sqrt{2}) \quad \text{for } \epsilon \text{ sufficiently small.} \end{aligned}$$

In fact this same argument gives the same inequality for general X_1 with continuous distribution and the above rules.

Proof of Theorem 7 stated in the introduction. By hypothesis the $(N-n+1) \times (N-n+1)$ covariance matrix of $S_{(n)}, \dots, S_{(N)}$, dominates the covariance matrix of $S_{(n)}, \dots, S_{(N)}$. Hence by Slepian's theorem [18, p. 468] for any $c_n, \dots, c_N$, $P[S_{(n)} \leq c_n, \dots, S_{(N)} \leq c_N] \geq P[S_{(n)} \leq c_n, \dots, S_{(N)} \leq c_N]$. Thus, choosing $c_n = a_n \lambda$,

$$P[\overline{\lim} S_{(n)}/a_n \leq \lambda] \geq P[\overline{\lim} S_{(n)}/a_n \leq \lambda].$$

Since $\overline{\lim} S_{(n)}/a_n$ is a constant by the Kolmogoroff 0-1 law, Theorem 7 is proved.

IV. Tail σ -fields for ruled sums. Let $\mathfrak{L}(\cdot)$ denote the tail σ -field for $S_{(n)}$; i.e., $\mathfrak{L}(\cdot) = \bigcap_n \mathfrak{B}(S_{(n)}, S_{(n+1)}, \dots)$ where $\mathfrak{B}(S_{(n)}, S_{(n+1)}, \dots)$ is the Borel σ -field generated by $S_{(n)}, S_{(n+1)}, \dots$. For both $S_{(n)}$ and S_n this tail σ -field is trivial by the Hewitt-Savage theorem. However, this is not the case for other rules, even rules closely related to $\langle n \rangle$.

Let Y be a variable independent of the X_i , all variables having the same distribution and $EX_1 = 0$, $EX_1^2 = 1$ and define $S_{(n)} = Y + S_{(n-1)}$, $n = 1, 2, \dots$, so that $S_{(n)}$ differs from the rule $S_{\langle n \rangle}$ only in the presence of a single permanent variable Y . We show that $Y \in \mathfrak{L}(\cdot)$ by showing that $Y \in \mathfrak{B}(S_{(n)}, S_{(n+1)}, \dots)$ for all n . In fact, consider

$$Z_N = \sum_{j=n}^N \frac{1}{j-1} S_{(j)} \in \mathfrak{B}(S_{(n)}, S_{(n+1)}, \dots).$$

Now $E[(Z_n/\log N - Y)^2] = o(1)$, and so by Chebyshev's inequality, $Z_n/\log N \xrightarrow{P} Y$. But then a subsequence converges a.s. to Y which establishes that $Y \in \mathfrak{S}$.

Since for m fixed $E(S_{(n)}S_{(m)}/\sqrt{nm}) = (1/\sqrt{nm}) \rightarrow 0$ as $n \rightarrow \infty$, this same example shows that $\forall m \text{ cov}(S_{(m)}S_{(n)}) \xrightarrow{n} 0$ is not sufficient to insure that $\mathfrak{S}(\cdot)$ is trivial even if the X_i 's were $\mathcal{N}(0, 1)$. However, for normal variables X_i , $\mathfrak{S}$ will be trivial if $\forall m \max_{Y_m, Z_n} \text{cov}(Y_m, Z_n) \rightarrow 0$ as $n \rightarrow \infty$, where $Y_m = \sum_{i=1}^m \lambda_i S_{(i)}$ and $Z_n = \sum_{j=n}^{\infty} \mu_j S_{(j)}$ with λ_i, μ_j real. For this condition implies $\max_{Y_m, Z_n} \text{cov}(Y'_m, Z'_n) \rightarrow 0$ where $Y'_m \in \mathcal{B}(S_{(1)} \cdots S_{(m)})$ and $Z'_n \in \mathcal{B}(S_{(n)}, S_{(n+1)}, \dots)$, (see [13, p. 134]), and this latter condition of asymptotic separation of Borel fields implies $\mathfrak{S} = 0$ (see [15]).

Of course, one can obtain less interesting rules (n) with $\mathfrak{S}(\cdot) \neq 0$, e.g., by having $1 \in (2n+1)$, $1 \notin (2n)$ so that $X_1 = S_{(2n+1)} - S_{(2n)} \in \mathcal{B}(S_{2n}, \dots)$ for all n . Similarly, we can algebraically arrange that $\mathfrak{S}(\cdot) = \mathcal{B}(X_1, X_2, \dots)$ by having each $j \in (n)$, and $\notin (m)$ for arbitrarily large m, n .

The rules $S_{(n)}$ and $S_{(n-1)} + Y$ are intuitively close. However, since their tail σ -fields are different, the respective measures they induce on their sample spaces are not equivalent and are thus orthogonal since we are dealing with normal processes [17]. It is also possible to find a rule intuitively close to S_n which has a trivial tail σ -field like S_n but induces a measure orthogonal to that induced by S_n . Such an example is given by $S_{(n)} = S_{n-1} + Y_n$ where $\{Y_n\}$ is a sequence of independent random variables all distributed as X_1 and independent of the sequence X_j . The orthogonality of $S_{(n)}$ and S_n is shown by a standard computation with the Hellinger integral [14].

V. More general normal processes.

Theorem 1N. *Let Y_n be a centered normal process:*

$$(a) \quad \forall j \leq n, \quad 0 \leq E(Y_n Y_j) \leq E(Y_j^2) = \sigma_j^2.$$

$$(b) \quad \forall j \leq n, \quad \sigma_n^2 - \sigma_j^2 \geq \sigma_{n-j}^2.$$

$$(c) \quad \sum_{j=1}^N (1/\sigma_j) \rightarrow \infty.$$

Then $\forall a \in R, \forall \epsilon > 0, \forall \delta > 0$

$$(1) \quad \lim_{n \rightarrow \infty} \frac{1}{\lg(\sum_{i=1}^n (1/\sigma_i))} \sum_{k=1}^n \frac{1}{\sum_{j=1}^k (1/\sigma_j)} \phi[|Y_k - a| < \epsilon] \rightarrow \sqrt{2/\pi} \epsilon \text{ a.e.,}$$

$$(2) \quad P \left[\left| \sum_{k=1}^n \phi[|Y_k| < \epsilon] - \sum_{k=1}^n \phi[|Y_k - a| < \epsilon] \right| \right. \\ \left. > (\lg n)^{1+\delta} \left(\sum_{i=1}^n (1/\sigma_i) \right)^{\delta+1/2} \text{ i.o.} \right] = 0,$$

$$(3) \quad \sum_{k=1}^n \phi[|Y_k| < \epsilon] / \sum_{k=1}^n \phi[|Y_k - a| < \epsilon] \rightarrow 1 \text{ a.e.}$$

We note that in typical cases such as $\sigma_j \sim \sqrt{j}$, $\sum_{j=1}^n 1/\sigma_j \sim \frac{1}{2}\sqrt{j}$, so that if $0 < \delta < \delta'$, $(\lg n)^{1+\delta}(\sum_{j=1}^n (1/\sigma_j))^{\delta+1/2} < [\sum_{j=1}^n (1/\sigma_j)]^{\delta'+1/2}$ for all large n , and so the $(\lg n)^{1+\delta}$ factor can be suppressed in the statement of (2).

Corollary 2N. Under hypotheses (a) and (b) of Theorem 1N $\forall a$ $P[a \text{ is an accumulation point of } Y_n] = \begin{cases} 1 & \text{iff } \sum_{j=1}^n (1/\sigma_j) = \infty \\ 0 & \text{iff } \sum_{j=1}^n (1/\sigma_j) < \infty \end{cases}$.

In order to prove Theorem 1N and its associated Corollary 2N, we will need analogs of (R2), (R3).

Under hypotheses (a) and (b) of Theorem 1N,

$$(R2N) \quad P[|Y_n - a| < \epsilon, |Y_j - a| < \epsilon] \leq P[|Y_{n-j}| < \epsilon]P[|Y_j - a| < \epsilon],$$

$$(R3N) \quad P[|Y_n| < \epsilon, |Y_j - a| < \epsilon] - P[|Y_n - b| < \epsilon, |Y_j - a| < \epsilon] \leq K(1/(\sigma_{n-j}^3 \sigma_j))$$

for $n > j$.

Proof of (R2N). Let Y'_j, Y'_n be jointly normal centered variables with covariance matrix

$$\begin{pmatrix} \sigma_j^2 & \sigma_j^2 \\ \sigma_j^2 & \sigma_n^2 \end{pmatrix}$$

Thus $Y'_n = Y'_j + (Y'_n - Y'_j)$ where $(Y'_n - Y'_j)$ is orthogonal to and thus independent of Y'_j . By the proof given in § 3 of Proposition 2 stated in the introduction

$$\begin{aligned} P[|Y_j - a| < \epsilon, |Y_n - a| < \epsilon] \\ &\leq P[|Y'_j - a| < \epsilon, |Y'_n - a| < \epsilon] = \int_{a-\epsilon}^{a+\epsilon} F_{Y'_n - Y'_j}(a - y - \epsilon, a - y + \epsilon) dF_{Y'_j}(y) \\ &\leq F_{Y'_n - Y'_j}(-\epsilon, \epsilon) F_{Y'_j}(a - \epsilon, a + \epsilon) \leq P[|Y_{n-j}| < \epsilon] P[|Y_j - a| < \epsilon] \end{aligned}$$

since $\sigma_{n-j}^2 \leq \sigma_n^2 - \sigma_j^2$ by hypothesis.

Proof of (R3N). $Y_n = (E(Y_j Y_n)/\sigma_j^2)Y_j + [Y_n - (E(Y_j Y_n)/\sigma_j^2)Y_j] = Z_{nj} + W_{nj}$. We note $E(Z_{nj} W_{nj}) = 0$ and so since Z_{nj} and W_{nj} are jointly normal, they are independent. $d_{nj}^2 \equiv E(Y_j Y_n)/\sigma_j^2 \leq 1$ (by hypothesis).

$$\begin{aligned}
Q &\equiv |P[|Y_n - b| < \epsilon, |Y_j - a| < \epsilon] - P[|Y_n| < \epsilon, |Y_j - a| < \epsilon]| \\
&= \left| \int_{(a-\epsilon)d_{nj}^2}^{(a+\epsilon)d_{nj}^2} [F_{W_{nj}}(b - \epsilon - z, b + \epsilon - z) - F_{W_{nj}}(-\epsilon - z, \epsilon - z)] dF_{Z_{nj}}(z) \right| \\
&\leq \frac{1}{2\pi} \left(\max_R \frac{1}{w_{nj}} e^{-x^2/2w_{nj}^2} - \min_R \frac{1}{w_{nj}} e^{-x^2/2w_{nj}^2} \right) 2\epsilon \frac{2\epsilon}{\sigma_j}
\end{aligned}$$

where $w_{nj}^2 = \text{var } W_{nj} = \sigma_n^2 - \text{var } Z_{nj} = \sigma_n^2 - E^2(Y_j Y_n) \sigma_j^2 / \sigma_j^4 \geq \sigma_n^2 - \sigma_j^2 \geq \sigma_{n-j}^2$ and $R = [-|a| - |b| - 2\epsilon, |a| + |b| + 2\epsilon]$. Since for $\lambda \geq 0$, $1 - e^{-\lambda} \leq \lambda$, $Q \leq (|a| + |b| + 2\epsilon)^2 (\epsilon^2/\pi) (\sigma_{n-j}^3 \sigma_j)^{-1}$.

We now mimic the proof of (2) of Theorem 1 to prove (2) of Theorem 1N. Let $D(l, n) \equiv \sum_{k=l}^n \phi[|Y_k - a| < \epsilon] - \sum_{k=l}^n \phi[|Y_k - a'| < \epsilon]$ with $D_n \equiv D(1, n)$, and let $M_{l,n} \equiv \sum_{i=l}^n P[Y_i \in (-\epsilon, \epsilon)]$. By (R3N),

$$\begin{aligned}
E[D^2(l, n)] &= O(M_{l,n}) \\
&\quad + 2 \sum_{l \leq j < k \leq n} \{ (P[|Y_k - a| < \epsilon, |Y_j - a| < \epsilon] \\
&\quad - P[|Y_k - a'| < \epsilon, |Y_j - a| < \epsilon]) \\
&\quad + (P[|Y_k - a'| < \epsilon, |Y_j - a'| < \epsilon] \\
&\quad - P[|Y_k - a| < \epsilon, |Y_j - a'| < \epsilon]) \} \\
&= O(M_{l,n}) + O\left(\sum_{l \leq j < k \leq n} \frac{1}{\sigma_{k-j}^3} \frac{1}{\sigma_j} \right) = O\left(\sum_{j=l}^n \frac{1}{\sigma_j} \right)
\end{aligned}$$

since $\sigma_n^2 \geq n\sigma_1^2$ by hypothesis. We now follow the proof of (2) of Theorem 1 with the modifications $\mu_i = (1/\sigma_i)$ in the application of the theorem from Billingsley [2] and $(\sum_{j=1}^n 1/\sigma_j)(\log^2 4n)$ replacing $M(n)^{3/2+\epsilon} \epsilon^n$ from the part dealing with $\sim k^\lambda$ till the end of the proof.

The proofs of (1) and (3) of Theorem 1N go through as previously.

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