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ON A FUNCTIONAL CENTRAL LIMIT THEOREM
FOR RANDOM WALKS CONDITIONED
TO STAY POSITIVE

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Let $\{X_k : k \geq 1\}$ be a sequence of i.i.d.rv with $E(X_i) = 0$ and $E(X_i^2) = \sigma^2$, $0 < \sigma^2 < \infty$. Set $S_n = X_1 + \dots + X_n$. Let $Y_n(t)$ be $S_k/\sigma n^{1/2}$ for $t = k/n$ and suitably interpolated elsewhere. This paper gives a generalization of a theorem of Iglehart which states weak convergence of $Y_n(t)$, conditioned to stay positive, to a suitable limiting process.

1. Introduction. Let $\{X_i\}_{i \in N}$ be a sequence of i.i.d.rv with $E(X_i) = 0$ and $E(X_i^2) = \sigma^2$ where $0 < \sigma^2 < \infty$. Let $S_k = X_1 + \dots + X_k$ and $Y_n(t)$ be the continuous process on $[0, 1]$ for which $Y_n(k/n) = S_k/\sigma n^{1/2}$ and which is linearly interpolated elsewhere.

It is well known (see e.g., [2]) that $Y_n(t)$ converges weakly in $(C[0, 1], \rho)$ to the Brownian motion process, where $C[0, 1]$ is the set of continuous functions on $[0, 1]$ and ρ the supremum metric.

Let now $C^+ = \{f \in C : f(t) \geq 0 \text{ for } t \in [0, 1]\}$. We have $P(Y_n \in C^+) > 0$ for each n . So the definition of conditional probabilities is elementary. Let Y_n^+ be the Y_n -process conditioned to stay positive. That is for all Borel-sets $A \subset C[0, 1]$ we set $P(Y_n^+ \in A) = P(Y_n \in A | Y_n \in C^+)$. We remark that C^+ is a null set for the measure of the Brownian motion. Iglehart proved [3] weak convergence of the Y_n^+ process to the Brownian meander process W^+ which is defined by

$$(1.1) \quad W^+(t) = \left| \frac{1}{(1-\tau)^{1/2}} W(\tau + (1-\tau)t) \right|, \quad 0 \leq t \leq 1$$

with W the Brownian process and $\tau = \sup \{t \in [0, 1] : W(t) = 0\}$. (Notice that $\tau < 1$ a.s.)

Iglehart assumed $E|X_i|^3 < \infty$ and X_i nonlattice or integer valued with span 1. It is shown in this paper that these extra assumptions are superfluous. Iglehart calculates the finite-dimensional distributions and proves tightness. Then he identifies the process with (1.1) for which Belkin [1] calculated the finite dimensional distributions. The proof given here requires no computation. It is based on identifying $\lim_{n \rightarrow \infty} Y_n^+(t) = W(T+t) - W(T) = W^+(t)$ for an appropriate random time T and uses only the continuous mapping theorem (Theorem 5.1 in [2]).

2. Notations and preliminary lemmas. For $s \in (0, \infty]$ let C^s be the set of

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continuous functions on $[0, s]$ (or $[0, \infty)$ for $s = \infty$) and \mathcal{B}^s the smallest σ -algebra such that the mappings $C^s \ni f \rightarrow f(t) \in \mathbb{R}$ are measurable.

Let P^s be the measure of the Brownian motion on (C^s, \mathcal{B}^s) .

$T^s: C^s \rightarrow \bar{\mathbb{R}}^+ = [0, \infty]$ is the mapping with

$$(2.1) \quad T^s(f) = \inf \{t: f(u) \geq f(t) \text{ for } t \leq u \leq t + 1 \leq s\}, \quad (\inf \emptyset = \infty).$$

We set $T = T^\infty$ and $P = P^\infty$ for simplicity.

LEMMA 2.1. For all $s \in (0, \infty]$ T^s is \mathcal{B}^s -measurable.

PROOF. If $v = s - (u + 1) > 0$ then $\{T^s \leq u\} = \bigcap_{n \geq 1/v} \{f \in C^s: \text{there exists a rational } r \leq u + 1/n \text{ with } f(r) < \min_{1 \leq i \leq n-1} f(r + i/n) + 1/n\}$, which is easily seen to belong to \mathcal{B}^s .

LEMMA 2.2. $P(T < \infty) = 1$.

PROOF. Let $A_\varepsilon = \{f \in C^1: \text{ex. } s \leq 1 - \varepsilon \text{ with } f(s) \leq f(u) \text{ for } s \leq u \leq s + \varepsilon\}$. Now we have $A_\varepsilon \downarrow \{f \in C^1: f \text{ nonincreasing}\}$ as $\varepsilon \downarrow 0$. We infer $P(A_\varepsilon) \uparrow 1$ for $\varepsilon \downarrow 0$. If $\varphi: C^\infty \rightarrow C^\infty$ is defined by $\varphi(f)(t) = \varepsilon^{-1}f(\varepsilon t)$ then φ is measure preserving (see [5] page 246) and $\varphi(A_\varepsilon) \subset \{T < \infty\}$ so $P(T < \infty) \geq P(A_\varepsilon)$ for all $\varepsilon > 0$.

LEMMA 2.3. The following three statements are true for all $s \in (0, \infty]$.

$$(2.2) \quad P^s(f(T^s) = f(T^s + 1)) = 0;$$

$$(2.3) \quad P^s(T^s = s - 1) = 0;$$

$$(2.4) \quad P^s(\text{ex. } u \in (0, 1) \text{ with } f(T^s) = f(T^s + u)) = 0.$$

PROOF. We set $m(t) = \min_{0 \leq s \leq t} W(t)$. $D(t) = W(t) - m(t)$ has the same finite-dimensional distributions as $|W(t)|$ (see [5] page 193). Observe now that $T^s = \inf \{t \leq s - 1: m(t) = m(t + 1)\}$. Now $T^s = s - 1$ implies $D(s - 1) = 0$ which has P measure 0. This proves (2.3).

Let $U = \{\text{ex. } u < v < w \text{ with } m(u) = m(v) = m(w) \text{ and } D(u) = D(v) = D(w) = 0\}$. Then $U \subset \bigcup_{r, s \in \mathbb{Q}} \{\min_{0 \leq t \leq r} W(t) = \min_{r \leq t \leq r+s} W(t)\}$ and the last has P measure 0. This proves (2.4).

It suffices to prove (2.2) for $s = \infty$. With probability one, the hitting time process $\{T_{-x}: x \geq 0\}$ ($T_{-x} = \inf \{t: W(t) = -x\}$) has no jumps of length one. This follows from its Lévy decomposition (see Section 1.7 of [4]). Together with $P(U) = 0$ this yields (2.2).

LEMMA 2.4. For each $s \in (0, \infty]$ T^s is a continuous P^s a.e. on (C^s, ρ) .

PROOF. By (2.3) it suffices to consider the case $s = \infty$. Let f be such that $T(f) < \infty$ and f does not belong to the null sets defined in (2.2)—(2.4).

(I) We first prove that for all $\delta > 0$ there exists an $\varepsilon > 0$ with

$$T(f') \leq T(f) + \delta \quad \text{when } \rho(f, f') < \varepsilon.$$

By (2.2) there is as $\tau < \delta$ so that

$$\inf_{T+1 \leq u \leq T+1+\tau} f(u) > f(T).$$

Now (2.4) gives $\varepsilon = \frac{1}{3}(\inf_{T+\tau \leq u \leq T+\tau+1} f(u) - f(T)) > 0$.

If $\rho(f, f') < \varepsilon$ and γ' is such that $T(f) \leq \gamma' \leq T(f) + \tau$ and $f'(\gamma') = \inf_{T \leq u \leq T+\tau} f'(u)$ then $T(f') \leq \gamma' \leq T(f) + \delta$.

(II) To show the other inequality note that

$$\lim_{n \rightarrow \infty} (\inf \{T(f') : \rho(f, f') < 1/n\}) = \lambda \leq T(f).$$

Let $\{f_n\}_{n \in \mathbb{N}}$ be a sequence with $\rho(f, f_n) \leq 1/n$ and $\lim_{n \rightarrow \infty} T(f_n) = \lambda$. Let $\varepsilon > 0$. By the continuity of f and the uniform convergence of f_n , there exists n_0 such that for $n \geq n_0$ we have:

$$\begin{aligned} \inf_{\lambda \leq u \leq \lambda+1} f(u) &\geq \inf_{T(f_n) \leq u \leq T(f_n)+1} f(u) - \varepsilon \\ &\geq \inf_{T(f_n) \leq u \leq T(f_n)+1} f_n(u) - 2\varepsilon \\ &\geq f_n(T(f_n)) - 2\varepsilon \geq f(T(f_n)) - 3\varepsilon \geq f(\lambda) - 4\varepsilon. \end{aligned}$$

So $\inf_{\lambda \leq u \leq \lambda+1} f(u) \geq f(\lambda)$ which implies $T(f) \leq \lambda$ completing the proof of Lemma 2.4.

Let u be the function in C^1 which is everywhere equal -1 . We define a map $\Phi_s : C^s \rightarrow C^1$

$$\begin{aligned} \Phi_s(f)(t) &= f(T^s(f) + t) && \text{for } T^s(f) < \infty \\ &= u && \text{for } T^s(f) = \infty. \end{aligned}$$

We write $\Phi = \Phi_\infty$ for simplicity.

A straightforward conclusion of Lemma 2.4 is

LEMMA 2.5. For each $s \in (0, \infty]$ Φ_s is continuous P^s a.s. on (C^s, ρ) .

3. Sums of independent random variables conditioned to stay positive. Let X_1, X_2, \dots , be i.i.d.rv with $E(X_i) = 0$; $E(X_i^2) = \sigma^2 < \infty$ ($\sigma^2 > 0$) and $S_k = \sum_{j=1}^k X_j$. $T_n = \inf \{k : S_{k+i} \geq S_k \text{ for } i = 1, \dots, n\}$. Clearly $T_n < \infty$ holds a.s. We set $Z_k = S_{T_n+k} - S_{T_n}$.

LEMMA 3.1. For each sequence of real numbers a_1, \dots, a_n

$$(3.1) \quad \begin{aligned} P(S_k \leq a_k, k = 1, \dots, n | S_k \geq 0, k = 1, \dots, n) \\ = P(Z_k \leq a_k, k = 1, \dots, n). \end{aligned}$$

PROOF. This is an easy consequence of the independence and identical distribution of the X_i :

If $B_j = \bigcup_{s=0}^{j-1} \{S_s \leq S_r \text{ for } s+1 \leq r \leq \min(j, s+n)\}$ we have

$$\begin{aligned} P(S_{T_n+k} - S_{T_n} \leq a_k \text{ for } k = 1, \dots, n) \\ = \sum_{j=0}^{\infty} P(S_{j+k} - S_j \leq a_k \text{ for } k = 1, \dots, n | T_n = j) P(T_n = j) \end{aligned}$$

$$\begin{aligned}
 &= \sum_{j=0}^{\infty} P(S_{j+k} - S_j \leq a_k \text{ for } k = 1, \dots, n | S_{j+k} \geq S_j \\
 &\quad \text{for } k = 1, \dots, n \text{ and } B_j^c) P(T_n = j) \\
 &= P(S_k \leq a_k, k = 1, \dots, n | S_k \geq 0, k = 1, \dots, n) \\
 &\hspace{15em} \text{since } T_n < \infty \text{ a.s.}
 \end{aligned}$$

We set $Y_n(k/n) = (1/n^{\frac{1}{2}}\sigma)S_k$ for $k \geq 0$ and $Y_n(t)$ linearly interpolated.

Let Q_n be the probability measure defined on $(C^\infty, \mathcal{B}^\infty)$ by this process. Let $\Pi_s: C^\infty \rightarrow C^s$ be the projection map and Φ, C^+ defined as above. We remark that $P^s = P\Pi_s^{-1}$.

Let $Q_n \Pi_1^{-1}(dx | C^+)$ be the probability measure on C^1 which is defined by

$$Q_n \Pi_1^{-1}(A | C^+) = Q_n(\Pi_1^{-1}(A \cap C^+)) / Q_n(\Pi_1^{-1}(C^+))$$

for $A \in \mathcal{B}^1$.

THEOREM 3.2. *The probability measures $Q_n \Pi_1^{-1}(dx | C^+)$ converge weakly to $P\Phi^{-1}$ (on (C^1, ρ)).*

PROOF. We have proved in Lemma 3.1 that

$$(3.2) \quad Q_n \Pi_1^{-1}(dx | C^+) = Q_n \Phi^{-1}(dx) \text{ holds.}$$

Now by Donsker's theorem (see [2]), $Q_n \Pi_s^{-1}$ converges weakly to P^s for $s < \infty$. With regard to Lemma 2.5 we have for $s < \infty$

$$(3.3) \quad Q_n(\Phi_s \Pi_s)^{-1} \rightarrow P^s \Phi_s^{-1} \text{ weakly.}$$

(Theorem 5.1 in [2].)

Let A be a continuity set in \mathcal{B}^1 , that is $P\Phi^{-1}(\partial A) = 0$. We are going to show that

$$(3.4) \quad \lim_{n \rightarrow \infty} Q_n \Phi^{-1}(A) = P\Phi^{-1}(A).$$

The theorem then follows. (3.4) doesn't follow directly from (3.3) because we have there the assumption $s < \infty$. Set

$$D = \{f \in C^1 : \min_{0 \leq t \leq 1} f(t) \geq -\frac{1}{2}\}.$$

Without loss of generality we can assume $A \subset D$. (If not: replace A by $A \cap D$ noticing $Q_n \Phi^{-1}(D^c) = P\Phi^{-1}(D^c) = P\Phi^{-1}(\partial D) = 0$).

Let $\varepsilon > 0$ be given. According to Lemma 2.2 we have $P(T < \infty) = 1$. So there exists a real number $c > 0$ such that $P(T \leq c - 1) \geq 1 - \varepsilon$.

We choose n_0 such that for $n \geq n_0$

$$(3.5) \quad |Q_n \Pi_c^{-1}(T^c < \infty) - P^c(T^c < \infty)| \leq \varepsilon.$$

(According to Lemma 2.4 $\{T^c < \infty\}$ is a continuity set with respect to P^c . (3.5) then follows by Donsker's theorem.)

We infer from (3.5) and the setting of c :

$$(3.6) \quad P(\Phi_c \Pi_c \neq \Phi) \leq \varepsilon,$$

$$(3.7) \quad Q_n(\Phi_c \Pi_c \neq \Phi) \leq 2\varepsilon .$$

(We have $\{\Phi_c \Pi_c = \Phi\} \cap \{T < \infty\} = \{T^c \Pi_c < \infty\} = \{T \leq c - 1\}$.)

We choose $n_1 \geq n_0$ such that for $n \geq n_1$

$$(3.8) \quad |Q_n(\Phi_c \Pi_c)^{-1}(A) - P^c \Phi_c^{-1}(A)| \leq \varepsilon .$$

(The element u doesn't belong to ∂A because we assumed $A \subset D$. It is easily seen that $(\Phi_c \Pi_c)^{-1}(\partial A) \subset \Phi^{-1}(\partial A)$ holds, so we infer that $P(\Phi_c \Pi_c)^{-1}(\partial A) = P^c \Phi_c^{-1}(\partial A) = 0$ and the existence of an n_1 , such that (3.8) holds then follows from (3.3).)

For $n \geq n_1$ we have:

$$\begin{aligned} |Q_n \Phi^{-1}(A) - P \Phi^{-1}(A)| &\leq |Q_n \Phi^{-1}(A) - Q_n(\Phi_c \Pi_c)^{-1}(A)| \\ &\quad + |Q_n(\Phi_c \Pi_c)^{-1}(A) - P^c \Phi_c^{-1}(A)| \\ &\quad + |P(\Phi_c \Pi_c)^{-1}(A) - P \Phi^{-1}(A)| \\ &\leq Q_n(\Phi \neq \Phi_c \Pi_c) + \varepsilon + P(\Phi \neq \Phi_c \Pi_c) \leq 4\varepsilon . \end{aligned}$$

So $\lim_{n \rightarrow \infty} Q_n \Phi^{-1}(A) = P \phi^{-1}(A)$ which is (3.4) and the proof is complete.

So far we have proved that Y_n^+ converges weakly to $P \Phi^{-1}$ which is $W(T+t) - W(T)$ $0 \leq t \leq 1$. It remains to identify $W(T+\cdot) - W(T)$ with the Brownian meander W^+ . But this clearly follows from Iglehart's result. We give a sketch of a proof using the methods of the present paper: Let $X_i = \pm 1$ each with probability $\frac{1}{2}$. Set $\mu_n = \inf \{k \leq n : \text{the sequence } S_k, \dots, S_n \text{ does not change sign}\}$ and let $\nu_n = n - \mu_n$ (remark that $\nu_n \geq 1$). We define $\tilde{Y}_n(t)$ as follows: $\tilde{Y}_n(k/\nu_n) = (1/\nu_n)^{\frac{1}{2}} |S_{\mu_n+k}|$ for $0 \leq k \leq \nu_n$ and linearly interpolated elsewhere. $\tilde{Y}_n(\cdot)$ has the same distribution as $Y_{\nu_n}^+(\cdot)$ where $\{Y_k^+\}_{k \in \mathbb{N}}$ and ν_n are independent. Define $\tau' : C^1 \rightarrow [0, 1]$ by $\tau'(f) = \inf \{t \in [0, 1] : f(s) \text{ does not change sign for } s \in [t, 1]\}$. Further, define $\Psi : C^1 \rightarrow C^1$ by $\Psi(f)(t) = |(1 - \tau')^{-\frac{1}{2}} f(\tau' + (1 - \tau')t)|$ for $\tau' \in [0, 1)$, and $\Psi(f)$ identically zero for $\tau' = 1$. We then have $\tilde{Y}_n = \Psi(Y_n)$, which is identical in law to $Y_{\nu_n}^+$. Now $\tau' = \tau = \sup \{t \in [0, 1] : f(t) = 0\}$ P^1 -a.s. (This can be proved in the same way as the statements of Lemma 2.3). So W^+ has the same distribution as $\Psi(W)$. It can be shown by the same methods as in Lemma 2.4 and 2.5 that Ψ is P^1 -a.s. continuous on (C^1, ρ) . The continuous mapping theorem implies $\tilde{Y}_n \rightarrow W^+$ and so $Y_{\nu_n}^+ \rightarrow W^+$ in distribution. By Theorem 3.2 $Y_n^+ \rightarrow W(T+\cdot) - W(T)$. Clearly $\nu_n \rightarrow \infty$ in distribution. This is sufficient for $Y_{\nu_n}^+ \rightarrow W(T+\cdot) - W(T)$ because $\{Y_n^+\}$ and ν_n are independent. It follows that W^+ and $W(T+\cdot) - W(T)$ have the same distribution.

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