

The Variation of a Stable Path is Stable

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Summary. Let $X(t)$ be a separable symmetric stable process of index α . Let P be a finite partition of $[0, 1]$, and \mathcal{P} a collection of partitions. The variation of a path $X(t)$ is defined in three ways in terms of the sum $\sum_{t_i \in P} |X(t_i) - X(t_{i-1})|^\beta$ and the collection \mathcal{P} . Under certain conditions on \mathcal{P} and on the parameters α and β , the distribution of the variation is shown to be a stable law. Under other conditions the distribution of the variational sum converges to a stable distribution.

1. Introduction

Let $X(t)$ be a separable symmetric stable process on R^1 of index α , $0 < \alpha \leq 2$. $X(t)$ has characteristic function $e^{-t|\lambda|^\alpha}$. Let P be a finite partition, $\{t_1 < \dots < t_k\}$ of $[0, 1]$, and \mathcal{P} a collection of such partitions. Let $V(P, \beta)$ denote the variational sum $\sum_{t_i \in P} |X(t_i) - X(t_{i-1})|^\beta$, where $\beta > 0$. The variation of the stable path $X(t)$ can be calculated in at least three ways: as the supremum over \mathcal{P} of $V(P, \beta)$; as the limit as $n \rightarrow \infty$ of the supremum over \mathcal{P}_n of $V(P, \beta)$ where $\mathcal{P}_n = \{P \in \mathcal{P} : \text{mesh of } P \leq 1/n\}$; if \mathcal{P} is a sequence $\{P_n\}$, as the \limsup as $n \rightarrow \infty$ of $V(P, \beta)$. We give sufficient conditions on \mathcal{P} that the variation, calculated in each of these ways, be a stable random variable of index α/β if $\beta \leq 1$. The case $\beta > 1$ presents special problems, some of which are resolved in Section 5. We use and refine results of Blumenthal and Gettoor, and of Bochner [1, 2, 3].

2. Definitions

We impose closure conditions on \mathcal{P} which are satisfied, for example, by the collection of all finite partitions and by the collection of all finite rational partitions.

If $P = \{t_1, \dots, t_k\} \subset [0, 1]$, let $rP + s$ denote the partition

$$\{rt_1 + s, \dots, rt_k + s\} \cap [0, 1].$$

We say that \mathcal{P} is closed under translation and multiplication by $r = \frac{1}{3}, \frac{1}{2}, \frac{2}{3}$ if for each $P \in \mathcal{P}$ some refinement of each of $rP, (1/r)P, P+r, P-r$ is in \mathcal{P} .

We define \mathcal{P} to be closed under piecing at $r = \frac{1}{2}, \frac{1}{3}$ if whenever P_1 and P_2 are in \mathcal{P} some refinement of the partition $(P_1 \cap [0, r]) \cup (P_2 \cap [r, 1]) \cup r$ is also in \mathcal{P} .

When we use these conditions with $\beta > 1$, we intend that the words “some refinement of” should be omitted.

If \mathcal{P} is a sequence of partitions we require that the closure conditions hold in the forward direction of the sequence, and define \mathcal{P} to be closed with respect to translation, multiplication, and piecing for $r = \frac{1}{3}, \frac{1}{2}, \frac{2}{3}$ if whenever P_n and P_m are in \mathcal{P} there is a refinement further along in the sequence for each of $rP_n, 1/rP_n, P_n+r, P_n-r$, and $(P_n \cap [0, r]) \cup (P_m \cap [r, 1]) \cup r$. An example of such a sequence is $P_n = \{i/n, i=0, \dots, n\}$.

Let $X \stackrel{d}{=} Y$ mean that the random variables X and Y have the same distribution. It follows from the form of the characteristic function that for $a > 0$, $X(t) \stackrel{d}{=} a^{1/\alpha} X(t/a)$.

The mesh μ of a partition P will be the maximum length of the interval between successive points of P .

Finally, we introduce distinguishing notation for the three types of variation which we consider. The supremum over P in \mathcal{P} of the variational sum $V(P, \beta) = \sum_{t_i \in P} |X(t_i) - X(t_{i-1})|^\beta$ is denoted by β -var X .

The limit as $n \rightarrow \infty$ of the supremum over \mathcal{P}_n of $V(P, \beta)$ where $\mathcal{P}_n = \{P \in \mathcal{P} : \text{mesh of } P \leq 1/n\}$ is denoted by $\lim \beta$ -var $X(\mathcal{P}_n)$.

In case \mathcal{P} is a sequence of partitions P_n , the $\lim \sup$ as $n \rightarrow \infty$ of $V(P_n, \beta)$ is denoted by $\lim \sup \beta$ -var $X(P_n)$.

β -var $X[a, b]$ denotes the supremum over P in \mathcal{P} of $V(P \cap [a, b], \beta)$.

3. A Method Using the Definition of Stable Random Variable

Theorem 1. *Let $X(t)$ be the symmetric stable process on R^1 of index α , $0 < \alpha \leq 2$, let $0 < \beta \leq 1$, and let \mathcal{P} be a collection of finite partitions of $[0, 1]$, closed under translation and multiplication by $\frac{1}{3}, \frac{1}{2}, \frac{2}{3}$ and under piecing at $\frac{1}{3}, \frac{1}{2}$. Then the distribution of β -var X is a stable law of index α/β and*

- i) if $0 < \alpha/\beta < 1$ then the distribution is a proper stable law on $[0, \infty)$;
- ii) if $\alpha/\beta = 1$ then β -var X is either a constant with probability one, or is ∞ with probability one;
- iii) if $\alpha/\beta > 1$, $P\{\beta\text{-var } X = \infty\} = 1$.

Proof. That β -var X is a random variable is established by Blumenthal and Gettoor [1]. We will show that $Y = \beta$ -var X satisfies the defining relation for a stable random variable,

$$S_n \stackrel{d}{=} c_n Y, \quad \text{where } S_n = \sum_{i=1}^n Y_i,$$

the Y_i are independent and distributed like Y , and $c_n = n^{\beta/\alpha}$. It suffices that the relation hold for $n = 2, 3$ [4, p. 215]. An equivalent relation is $r^{\beta/\alpha} Y_1 + (1-r)^{\beta/\alpha} Y_2 \stackrel{d}{=} Y$, for $r = \frac{1}{2}, \frac{1}{3}$, where $Y_1 \stackrel{d}{=} Y_2 \stackrel{d}{=} Y$ and Y_1, Y_2 are independent. The proof of this relation is given in three parts: For $r = \frac{1}{2}, \frac{1}{3}$,

$$\beta\text{-var } X = \beta \text{ var } X[0, r] + \beta\text{-var } X[r, 1], \tag{1}$$

where the summands are independent;

$$\beta\text{-var } [r, 1] \stackrel{d}{=} \beta\text{-var } [0, 1-r]; \tag{2}$$

$$\beta\text{-var } [0, r] \stackrel{d}{=} r^{\beta/\alpha} \beta\text{-var } [0, 1], \tag{3}$$

for $r = \frac{1}{2}, \frac{1}{3}, \frac{2}{3}$.

For fixed P and X , since $\beta \leq 1$, the variational sum is only increased if P is replaced by a refinement of $P \cup \{\frac{1}{3}, \frac{1}{2}\}$. Therefore we assume $r \in P$ and write

$$V(P, \beta) = V(P \cap [0, r], \beta) + V(P \cap [r, 1], \beta).$$

The condition that \mathcal{P} is closed under piecing at $r = \frac{1}{3}, \frac{1}{2}$ enables us to calculate β -var X , for fixed X , by taking the supremum of these two terms separately. Since the random variables β -var $X [0, r]$ and β -var $X [r, 1]$ depend on values of $X(t)$ in non-overlapping time intervals, they are independent, and (1) is proved.

Because the process is stationary,

$$V(P \cap [r, 1], \beta) \stackrel{d}{=} \sum_{t_{i-1}, t_i \in P \cap [r, 1]} |X(t_i - r) - X(t_{i-1} - r)|^\beta.$$

For each path the supremum over $P \in \mathcal{P}$ of the right side is the same as β -var $X [0, 1 - r]$, since \mathcal{P} is closed under translation by r . The supremum of the left side is β -var $X [r, 1]$, so we have (2).

For fixed $P, V(P \cap [0, r], \beta) \stackrel{d}{=}$

$$r^{\beta/\alpha} \sum_{t_{i-1}, t_i \in P \cap [0, r]} |X(t_i/r) - X(t_{i-1}/r)|^\beta. \tag{4}$$

The closure of \mathcal{P} under multiplication by r enables us to calculate the supremum of (4) over $P \in \mathcal{P}$ for fixed X by taking the supremum of $r^{\beta/\alpha} V(P, \beta)$. Statement (3) follows.

Now combining (1) and (2) and applying (3) to each term we have

$$\begin{aligned} \beta\text{-var } X &\stackrel{d}{=} \beta\text{-var } X [0, r] + \beta\text{-var } X [0, 1 - r] \\ &\stackrel{d}{=} r^{\beta/\alpha} \beta\text{-var } X + (1 - r)^{\beta/\alpha} \beta\text{-var } X, \end{aligned}$$

where the terms on the right are independent random variables. This is the relation we wished to prove. Note that only the proof of (1) requires that $\beta \leq 1$.

Let F be the distribution of $Y = \beta\text{-var } X$. Since Y is nonnegative, either F is concentrated at a single point, possibly ∞ , or the Laplace transform of F is $e^{-c\lambda^\gamma}$ where $0 < \gamma < 1$, [cf. 4, p. 424]. In the latter case $\gamma = \alpha/\beta$.

Suppose F is concentrated at a . The supremum is 0 only with probability 0, so $a > 0$, and $F * F$ is concentrated at $2a$. But $Y_1 + Y_2 \stackrel{d}{=} 2^{\beta/\alpha} Y$, so a is ∞ except possibly when $\beta/\alpha = 1$. We conclude that when $\beta/\alpha < 1$, F is concentrated at ∞ , whereas if $\beta/\alpha = 1$, F may be concentrated at any point of $[0, \infty]$. This concludes the proof of (ii) and (iii).

In case $0 < \alpha/\beta < 1$ the above argument does not tell us whether F is the proper stable law of index α/β on $[0, \infty)$ or whether F is concentrated at ∞ . The following theorem of Blumenthal and Gettoor [1] gives the answer.

Theorem A. Let $X(t)$ be the symmetric stable process in R^n of index $\alpha, 0 < \alpha \leq 2$, and let $\mathcal{P}' = \{\text{all finite partitions of } [0, 1]\}$. Then $P\{\beta\text{-var } X = \infty\} = 1$ or 0 according as $\beta \leq \alpha$ or $\beta > \alpha$.

We use only the second half of this theorem. (The above includes an independent proof of the result for $\beta < \alpha$.) If $\mathcal{P} \subset \mathcal{P}'$ then $\beta\text{-var } X$ over $\mathcal{P} \leq \beta\text{-var } X$ over \mathcal{P}' . Therefore if $\beta > \alpha, P\{\beta\text{-var } X = \infty\} = 0$ for any collection of finite partitions and F is not concentrated at ∞ . We conclude that F is the proper stable law on $[0, \infty)$.

If \mathcal{P} is a sequence of partitions, closed under translation and multiplication by $\frac{1}{3}, \frac{1}{2}, \frac{2}{3}$, and piecing at $\frac{1}{3}, \frac{1}{2}$, which proceeds by refinements, and $\beta \leq 1$, then $\lim \sup$

β -var $X(P_n) = \text{limit } \beta\text{-var } X = \beta\text{-var } X$, the supremum over the sequence \mathcal{P} regarded as a set, so that Theorem 1 applies. If \mathcal{P} does not proceed by refinements, the identity $\limsup \beta\text{-var } X(P_n) \stackrel{d}{=} \beta\text{-var } X$, for $\beta \leq 1$, $\beta \neq \alpha$ follows from Theorem 2, which can also be proved as an easy corollary to Theorem 1.

Theorem 2. *Let \mathcal{P} be a sequence of finite partitions of $[0, 1]$, closed with respect to translation and multiplication by $\frac{1}{3}, \frac{1}{2}, \frac{2}{3}$ and piecing at $\frac{1}{3}, \frac{1}{2}$. Let $\beta \leq 1$. Then Theorem 1 holds with $\beta\text{-var } X$ replaced by $\limsup \beta\text{-var } X(P_n)$.*

Proof. The proof parallels the proof of Theorem 1 with the obvious modifications. In place of Theorem A we need to know that $P\{\limsup \beta\text{-var } X = \infty\} = 0$ if $\beta > \alpha$. This follows easily from Theorem A. As a set, $\mathcal{P} \subset \mathcal{P}'$ so that

$$\limsup \beta\text{-var } X(P_n) \leq \beta\text{-var } X \text{ over } \mathcal{P} \leq \beta\text{-var } X \text{ over } \mathcal{P}'.$$

4. A Method Using Laplace Transforms

When \mathcal{P} is a sequence we expect that the distribution of the variation can sometimes be established by showing that the distributions of $Y_n = \sum_{t_i \in P_n} |X(t_i) - X(t_{i-1})|^\beta$ converge to a stable law, while the random variables Y_n converge almost surely. An example is given by Theorem 3, which differs from Theorem 2 in allowing $\beta > 1$ and in restricting the mesh of the partitions. We require the following lemma:

Lemma. *Let X_n be a sequence of random variables and let $\limsup_{n \rightarrow \infty} X_n = X$. Let F_n be the distribution of X_n , and suppose the F_n converge to a limit distribution F . Let G be the distribution of X . Then $G(y) \leq F(y)$ for every y . If X_n converges to X almost surely, then $G = F$.*

Theorem 3. *Let $X(t)$ be a symmetric stable process on R^1 with index $\alpha, 0 < \alpha < 2$. Let \mathcal{P} be a sequence of partitions P_n such that P_n contains n points and the mesh μ_n of P_n goes to 0 as n goes to ∞ . Let $Y_n = \sum_{t_i \in P_n} |X(t_i) - X(t_{i-1})|^\beta$. Then*

i) *if $\beta > \alpha$, the distribution of Y_n converges to the one-sided stable law of index α/β . If in addition $\beta \leq 1$ and P_n proceeds by refinements then $Y_n \rightarrow Y$ almost surely, where Y is distributed by the limiting distribution;*

ii) *if $\alpha/2 < \beta < \alpha$, the distribution of Y_n converges to 0 on $[0, \infty)$ and*

$$P\{\limsup_{n \rightarrow \infty} Y_n = \infty\} = 1;$$

iii) *if $\alpha = \beta$ then (ii) holds.*

Proof. Let H be the distribution of $X(1)$. That $x^\alpha(1 - H(x)) \rightarrow b$, a constant > 0 as $x \rightarrow \infty$ is a well-known property of stable laws [c. f. 4, p. 547]. Since H is symmetric, the distribution of $|X(1)|^\beta$ is $G(x) = 2H(x^{1/\beta}) - 1$ on $[0, \infty)$, and

$$x^{\alpha/\beta}(1 - G(x)) \rightarrow 2b \quad \text{as } x \rightarrow \infty. \tag{5}$$

The distribution of $|X(t_i) - X(t_{i-1})|^\beta$ is the same as that of $|X(1)|^\beta(t_i - t_{i-1})^{\beta/\alpha}$ and the terms of the sum Y_n are independent. Denoting $(t_i - t_{i-1})$ by Δt_i , the distribution of Y_n is

$$G(\Delta t_1^{-\beta/\alpha} x) * G(\Delta t_2^{-\beta/\alpha} x) * \dots * G(\Delta t_n^{-\beta/\alpha} x).$$

Let $\varphi = \mathfrak{L}G$. Then the Laplace transform of the convolution is $\prod_{i=1}^n \varphi(\lambda \Delta t_i^{\beta/\alpha})$, and that of the distribution with density $1 - G$ is $(1 - \varphi(\lambda))/\lambda$. Therefore (5) implies

$$(1 - \varphi(\lambda))/\lambda \sim c \lambda^{\alpha/\beta - 1} \quad \text{as } \lambda \rightarrow 0 \quad [4, \text{p. 424}]. \quad (6)$$

For fixed λ consider

$$-\log \prod_{i=1}^n \varphi(\lambda \Delta t_i^{\beta/\alpha}) = -\sum_{i=1}^n \log[\varphi(\lambda \Delta t_i^{\beta/\alpha}) - 1 + 1]. \quad (7)$$

We use the inequality $|\log(1+x) - x| \leq \frac{1}{2}x^2$, for small $|x|$, with $x = \varphi - 1$. Given $\varepsilon > 0$, let n be large enough so that $1 - \varphi(\lambda \Delta t_i^{\beta/\alpha}) \leq (c + \varepsilon)(\lambda \Delta t_i^{\beta/\alpha})^{\alpha/\beta}$ for $i = 1, \dots, n$. Then

$$\begin{aligned} & \sum_{i=1}^n |\log(\varphi(\lambda \Delta t_i^{\beta/\alpha}) - 1 + 1) - (\varphi(\lambda \Delta t_i^{\beta/\alpha}) - 1)| \\ & \leq \frac{1}{2} \sum_{i=1}^n |1 - \varphi(\lambda \Delta t_i^{\beta/\alpha})|^2 \leq \frac{1}{2} n \lambda^{2\alpha/\beta} \sum \Delta t_i^2 (c' + \varepsilon)^2 \\ & \rightarrow 0 \quad \text{as } n \rightarrow \infty \quad \text{since } \max_{i=1, \dots, n} \Delta t_i \rightarrow 0. \end{aligned}$$

It follows that

$$-\sum_{i=1}^n \log \varphi(\lambda \Delta t_i^{\beta/\alpha}) \sim \sum_{i=1}^n 1 - \varphi(\lambda \Delta t_i^{\beta/\alpha}).$$

Now choose n so large that

$$(c - \varepsilon) \lambda^{\alpha/\beta} \Delta t_i \leq 1 - \varphi(\lambda \Delta t_i^{\beta/\alpha}) \leq (c + \varepsilon) \lambda^{\alpha/\beta} \Delta t_i, \quad i = 1, \dots, n.$$

Summation over i yields

$$(c - \varepsilon) \lambda^{\alpha/\beta} \leq \sum_{i=1}^n 1 - \varphi(\lambda \Delta t_i^{\beta/\alpha}) \leq (c + \varepsilon) \lambda^{\alpha/\beta},$$

i.e.

$$\sum_{i=1}^n 1 - \varphi(\lambda \Delta t_i^{\beta/\alpha}) \rightarrow c \lambda^{\alpha/\beta} \quad \text{as } n \rightarrow \infty.$$

Combining this with (7) we have

$$-\log \prod_{i=1}^n \varphi(\lambda \Delta t_i^{\beta/\alpha}) \rightarrow c \lambda^{\alpha/\beta},$$

or

$$\prod_{i=1}^n \varphi(\lambda \Delta t_i^{\beta/\alpha}) \rightarrow \exp(-c \lambda^{\alpha/\beta}).$$

It follows from the continuity theorem for Laplace transforms that the distribution of Y_n converges to F where $\mathfrak{L}F = \exp(-c \lambda^{\alpha/\beta})$.

If $\beta \leq 1$ and \mathscr{P} proceeds by refinements then for each sample path the sequence Y_n is nondecreasing and therefore converges. The last statement of (i) now follows from the Lemma.

If $\beta < \alpha$ we use a similar argument. Since $1 - G(x) \sim 2b x^{-\alpha/\beta}$ and $\alpha/\beta > 1$, $F(x) = \int_0^x 1 - G(y) dy \rightarrow c$ as $x \rightarrow \infty$, where $0 < c < \infty$. It follows that $\Omega F(\lambda) = [1 - \varphi(\lambda)]/\lambda \rightarrow c$ as $\lambda \rightarrow 0$, i.e. $1 - \varphi(\lambda) \sim c\lambda$.

Proceeding as in case (i), given $\varepsilon > 0$ we have as $n \rightarrow \infty$,

$$\sum_{i=1}^n |1 - \varphi(\lambda \Delta t_i^{\beta/\alpha})|^2 \leq (c + \varepsilon)^2 \lambda^2 \sum \Delta t_i^{2\beta/\alpha} \rightarrow 0$$

if $2\beta/\alpha > 1$, so that again

$$-\sum_{i=1}^n \log \varphi(\lambda \Delta t_i^{\beta/\alpha}) \sim \sum_{i=1}^n 1 - \varphi(\lambda \Delta t_i^{\beta/\alpha}).$$

For large enough n ,

$$(c - \varepsilon) \lambda \sum_{i=1}^n \Delta t_i^{\beta/\alpha} \leq \sum_{i=1}^n 1 - \varphi(\lambda \Delta t_i^{\beta/\alpha}).$$

Since $\beta/\alpha < 1$ and $\max_{i=1, \dots, n} \Delta t_i \rightarrow 0$ the left side $\rightarrow \infty$, and so does $-\log \prod_{i=1}^n \varphi(\lambda \Delta t_i^{\beta/\alpha})$. Therefore $\prod_{i=1}^n \varphi(\lambda \Delta t_i^{\beta/\alpha}) \rightarrow 0$ as $n \rightarrow \infty$ for each λ , and the continuity theorem implies that the distribution of Y_n converges to 0 for each x in $[0, \infty)$. It follows from the Lemma that $P \{\limsup_{n \rightarrow \infty} Y_n = \infty\} = 1$.

If $\beta = \alpha$, then $y(1 - G(y)) \rightarrow b$ as $y \rightarrow \infty$, and $F(x) = \int_0^x 1 - G(y) dy \sim b \log x$. A Tauberian theorem [4, p. 421] implies that $[1 - \varphi(\lambda)]/\lambda \sim c \log(1/\lambda)$ as $\lambda \rightarrow 0$.

Again proceeding as in case (i), given $\varepsilon > 0$, as $n \rightarrow \infty$,

$$\begin{aligned} \sum_{i=1}^n |1 - \varphi(\lambda \Delta t_i)|^2 &\leq (c + \varepsilon)^2 \lambda^2 \sum_{i=1}^n \left(\Delta t_i \log \frac{1}{\lambda \Delta t_i} \right)^2 \\ &\leq \sum_{i=1}^n \Delta t_i^2 (\lambda \Delta t_i)^{-2\varepsilon} \rightarrow 0. \end{aligned}$$

For large enough n ,

$$\begin{aligned} (c - \varepsilon) \lambda \log \left(\frac{1}{\lambda \max_{i=1, \dots, n} \Delta t_i} \right) \sum \Delta t_i &\leq (c - \varepsilon) \lambda \sum \Delta t_i \log \left(\frac{1}{\lambda \Delta t_i} \right) \\ &\leq \sum_{i=1}^n 1 - \varphi(\lambda \Delta t_i). \end{aligned}$$

As $n \rightarrow \infty$, the left side and consequently also the right side $\rightarrow \infty$. As in case (ii) we conclude that the distribution of Y_n converges to 0, and $P \{\limsup_{n \rightarrow \infty} Y_n = \infty\} = 1$.

Brownian motion, the stable process with index $\alpha = 2$ can also be investigated by the method of Theorem 3. In the notation used there, $F(x) \rightarrow c > 0$ as $x \rightarrow \infty$ for every $\beta > 0$. This leads to the result that if $\mu_n \rightarrow 0$ as $n \rightarrow \infty$, the distribution of Y_n converges to a distribution concentrated at ∞ if $\beta < 2$, at 0 if $\beta > 2$, and at a number $d > 0$ if $\beta = 2$. We conclude that Y_n does not converge to a stable random variable in the case $\beta > \alpha = 2$.

Some information about the rate of growth of the variational sum if $\alpha = \beta < 2$ and $P_n = \{i/n, i=0, \dots, n\}$ is given by Theorem 4.

Theorem 4. Let $X(t)$ be a symmetric stable process of index $\alpha, 0 < \alpha < 2$. The distribution of $Z_n = \sum_{i=1}^n |X(i/n) - X((i-1)/n)|^\alpha / \log n$ converges to a distribution concentrated at $c, 0 < c < \infty$.

Proof. Let G be the distribution of $|X(1)|^\alpha$ and $F(x) = \int_0^x 1 - G(y) dy$. The distribution of Z_n is $G^{n^*}(x(n \log n))$. As in part (iii) of Theorem 3, $y(1 - G(y)) \rightarrow b$ as $y \rightarrow \infty, F(x) \sim b \log x$, and $[1 - \varphi(\lambda)]/\lambda \sim c \log 1/\lambda$, where $\Omega G = \varphi$. As $n \rightarrow \infty$,

$$-\log \varphi^n(\lambda/(n \log n)) \sim n [1 - \varphi(\lambda/(n \log n))] \sim n c (\lambda/(n \log n))$$

$$\cdot \log((n \log n)/\lambda) = c \lambda [\log n + \log \log n + \log 1/\lambda] / \log n \rightarrow c \lambda.$$

Therefore $\varphi^n(\lambda/(n \log n)) \rightarrow e^{-c\lambda}$. It follows from the continuity theorem for Laplace transforms that $G^{n^*}((n \log n) x)$ converges to the distribution concentrated at c .

5. Corollaries Using the Method of Section 3 and Results of Section 4

Corollary 1. Let $X(t)$ be a symmetric stable process of index $\alpha, 0 < \alpha < 2$, and let $\mathcal{P}_n = \{\text{all finite partitions of } [0, 1] \text{ of mesh } \mu \leq 1/n\}$. Let Y_n be the supremum over $P \in \mathcal{P}_n$ of $V(P, \beta), \beta > \alpha$. Then the variation $\lim \beta\text{-var } X(\mathcal{P}_n)$ defined by $Y = \lim_{n \rightarrow \infty} Y_n$ is a proper stable random variable of index α/β .

Proof. The proof follows that of Theorem 1. Statements (2) and (3) hold just as in Theorem 1. We must show that (1) holds.

Let A be the set of paths which are continuous at r . Since the stable process has no fixed discontinuities, $P\{A\} = 1$. For each path in $A, Y_n \leq Y_n[0, r] + Y_n[r, 1] + \sup_{P \in \mathcal{P}_n} |X(t_j) - X(t_{j-1})|^\beta$, where t_{j-1} and t_j are the points of P neighboring r , or $t_{j-1} = t_j = r$ if $r \in P$. As $n \rightarrow \infty, t_j - t_{j-1} \rightarrow 0$, and so does $\sup_{P \in \mathcal{P}_n} |X(t_j) - X(t_{j-1})|^\beta$.

Almost surely, then,

$$\lim Y_n \leq \lim Y_n[0, r] + \lim Y_n[r, 1].$$

To obtain the reverse inequality we write for each point in A ,

$$Y_n + \sup_{P \in \mathcal{P}_n} (|X(t_j) - X(r)|^\beta + |X(r) - X(t_{j-1})|^\beta) \geq Y_n[0, r] + Y_n[r, 1],$$

and apply a similar argument. We conclude that (1) holds.

As in Theorem 1, Y satisfies the defining relation for a stable random variable. Let F be the distribution of Y . F is either concentrated at a point, possibly ∞ , or is the proper one-sided stable law of index α/β . Since $\alpha < \beta$, as in Theorem 1 F cannot be concentrated at a point of $(0, \infty)$. Theorem A implies, again, that F is not concentrated at ∞ . That F is not concentrated at 0 follows from Theorem 3. Let $\{P_n\}$ be a sequence of partitions such that P_n has n points and the mesh of P_n decreases to 0 as $n \rightarrow \infty$. Then Theorem 3 part (i), together with the Lemma, implies that the distribution G of $\limsup \beta\text{-var } X(P_n)$ is bounded by a proper

one-sided stable law; in particular $G(0)=0$. For each sample path, $\lim \beta$ -var $X(\mathcal{P}_n) \geq \limsup \beta$ -var $X(P_n)$. We conclude that $F(0)=0$.

We note that this proof does not extend to Brownian motion since for $\alpha=2$, $\beta > \alpha$, and P_n becoming dense in $[0, 1]$, $\limsup \beta$ -var $X(P_n)=0$ almost surely.

Corollary 2. *Let $\mathcal{P} = \{\text{all finite partitions of } [0, 1]\}$, $0 < \alpha < 2$, $\beta > \alpha$, and β -var $X = \sup_{P \in \mathcal{P}} V(P, \beta)$. Let G be the distribution of β -var X . Then G is in the domain of normal attraction of the stable law F of Corollary 1, and $1 - G(x) \sim c x^{-\alpha/\beta}$ as $x \rightarrow \infty$.*

Proof. First we introduce some notation in addition to that of Corollary 1. If P is a finite partition, let $\mathcal{P}_n \cup P = \{\text{partitions in } \mathcal{P}_n \text{ which include the points of } P\}$. Let F_n denote the distribution of Y_n , $F_n(P)$ the distribution of the supremum of $V(Q, \beta)$ over Q in $\mathcal{P}_n \cup P$, and $F(P)$ the limit as $n \rightarrow \infty$ of $F_n(P)$. Since the collection of partitions $\mathcal{P}_n \cup P$ decreases with increasing n , the supremum of $V(Q, \beta)$ is nonincreasing, $F_n(P)$ is nondecreasing, and $F(P)$ is the distribution of the limiting random variable.

Let $r \in [0, 1]$. From Corollary 1 we have $\lim Y_n = \lim Y_n[0, r] + \lim Y_n[r, 1]$ almost surely. In terms of the new notation, if we regard $\{r\}$ as a partition, $F = F(r)$. More generally, if P is any finite partition $F = F(P) = \lim F_n(P)$. Since $\mathcal{P}_n \supset \mathcal{P}_n \cup P$ and $F_n(P)$ is nondecreasing as n increases, we have $F_n \leq F_n(P) \leq F$. This inequality holds, in particular, when P is P_n , the regular partition of mesh $1/n$. We conclude that $F_n(P_n)$ converges to F . But $\mathcal{P}_n \cup P_n = \mathcal{P} \cup P_n$, so $F(P_n)$ converges to F .

Now $F(P_n)$ is the n -fold convolution of the distribution of β -var $X[0, 1/n]$ which, as we saw in Section 1, is distributed like $n^{-\beta/\alpha}$ β -var X ; i.e. $F(P_n) = [G(n^{\beta/\alpha} x)]^{n*}$ converges to the stable law F . By definition, then, G is in the domain of normal attraction of F . The accompanying asymptotic relation is well known [4].

The next corollary is an extension of Theorem 3 part (i). Its proof uses Corollary 1 as well as previous results.

Corollary 3. *Let $X(t)$ be a symmetric stable process of index α , $0 < \alpha < 2$, and let \mathcal{P} be a sequence of partitions P_n of $[0, 1]$ such that P_n contains n points and $\mu_n = \text{mesh of } P_n$ decreases to 0 as $n \rightarrow \infty$. Then the distribution of $\limsup \beta$ -var $X(P_n)$ is bounded between two proper stable laws of index α/β , for every $\beta > \alpha$.*

Proof. With \mathcal{P}_n as in Corollary 1, for each path

$$\lim \beta\text{-var } X(\mathcal{P}_n) \geq \limsup \beta\text{-var } X(P_n).$$

If F and G are the distributions of these two random variables in the order written, then $F \leq G$. Corollary 1 says F is a proper stable law of index α/β . On the other hand Theorem 3 part (i), together with the Lemma, says G is bounded above by such a stable law.

6. Discussion

The restriction $\beta \leq 1$ in Theorem 1 was imposed in order to allow $\frac{1}{2}$ and $\frac{1}{3}$ to be included in all partitions. If \mathcal{P} is all finite partitions of $[0, 1]$, it is possible that the distribution of β -var X remains unchanged if $\frac{1}{2}$ or $\frac{1}{3}$ is added to every partition. If this is the case then the method of Theorem 1 will give a new proof of Theorem A whenever $\beta \leq \alpha$ and will specify that β -var X is distributed by a stable law when $\beta > \alpha$.

The scaling relation (3) can be shown to hold for each t in $[0, 1]$. When β -var X is distributed by a proper stable law with Laplace transform $\exp(-c\lambda^{\alpha/\beta})$ it follows from (3) that the transform of the law of β -var $X[0, t]$ is $\exp(-tc\lambda^{\alpha/\beta})$. It now follows from the stationary and Markov properties of the stable process $X(t)$ that β -var $X[0, t]$ can be viewed as a one-sided stable process or stable subordinator of index α/β . In fact an alternative approach to Theorem 1 is to show that β -var $X[0, t]$ is a differential process which satisfies (3) for each t . Blumenthal and Gettoor have observed this in a slightly different context [2, p. 509].

It is not clear what happens in Theorem 3 part (ii) if $\beta \leq \alpha/2$. If P_n is the sequence $P_n = \{i/n, i=0, \dots, n\}$, however, the method of Theorem 4 leads to the result $P\{\limsup \beta\text{-var } X = \infty\} = 1$ for every $\beta < \alpha < 2$.

The earliest results of this type appear to be those of P. Lévy [5] on the 2-variation of the Brownian path. Lévy showed that if \mathcal{P} is all finite partitions then $2\text{-var } X = \infty$ almost surely, whereas if \mathcal{P} is an increasing sequence of partitions which become dense in $[0, 1]$, i.e. the mesh goes to 0, then $\lim 2\text{-var } X = c > 0$ with probability 1. The method of Theorem 1 indicates that two such possibilities exist when $\alpha = \beta \leq 1$. However we can conclude from part (iii) of Theorem 3, together with Theorem A, that Brownian motion is the *only* stable process of the family for which the two types of α -variation have different distributions. If $0 < \alpha < 2$, then both the supremum of $\sum_{t_i \in \mathcal{P}} |X(t_i) - X(t_{i-1})|^\alpha$ over all finite partitions and the \limsup through a sequence of partitions which become dense, are almost surely infinite.

There remains the possibility that for α in $(0, 2]$, $\beta > 1$ and $\beta > \alpha$, the distribution of β -var X may differ from that of $\limsup \beta\text{-var } X(P_n)$ when the limit is taken through a sequence of partitions which become dense. However Corollaries 2 and 3 indicate that for $0 < \alpha < 2$, the two distributions have similar asymptotic behavior.

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