

Spitzer's Test for the Cauchy Process on the Line

By

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DVORETZKY-ERDÖS [1] and F. SPITZER [4] obtained the exact estimates of the order of the inferior limits at time points ∞ and 0 for the oscillation of Brownian motion paths. One of the present authors, in [5], has extended the result of [1] to the symmetric stable processes in R^N of index α with $N > \alpha$. The purpose of this note is to give the criteria for the remaining case $N = \alpha = 1$ which correspond to the result of [4] for the case of $N = \alpha = 2$.

Let $\{X(t, w), P_x\}^*$ be a symmetric Cauchy process in R^1 , that is, a Markov process in R^1 homogeneous in time and space with the transition probability $P(t, x, E)$ defined by

$$P(t, x, E) = P_x\{X(t, w) \in E\} = \int_E p(t, x - y) dy,$$

where

$$p(t, x) = \frac{1}{\pi} \frac{t}{t^2 + x^2} \quad t > 0, \quad x \in R^1.$$

We may assume as usual that its sample paths are right continuous and have left-hand limits everywhere. Let $g(t)$ be a function defined for all large t such that $g(t) > 0$ and $g(t) \downarrow 0$ as $t \uparrow +\infty$.

Definition. (i) We define $g(t) \in \mathcal{U}_\infty$ or $g(t) \in \mathcal{L}_\infty$ according as

$$P_0\{w; \text{there exists some } M > 0 \text{ such that } |X(t, w)| \geq tg(t) \text{ for all } t \geq M\}^{**}$$

= 0 or 1.

(ii) $g(t) \in \mathcal{U}_0$ or $g(t) \in \mathcal{L}_0$ according as

$$P_0\{w; \text{there exists some } \delta > 0 \text{ such that } |X(t, w)| \geq tg(1/t) \text{ for all } 0 < t < \delta\}^{***}$$

= 0 or 1.

Let E be a set on (t, x) -plane defined by

$$E = \left\{ (t, x); |x| < t \cdot g\left(\frac{1}{t}\right), t \geq 0 \right\},$$

then $g(t) \in \mathcal{U}_0$ if and only if the point $(0, 0)$ is a regular point of E for the space-time Cauchy process.

Theorem. (i) $g(t) \in \mathcal{U}_\infty$ or \mathcal{L}_∞ according as

* P_x denotes the probability law governing paths starting at $x \in R^1$.

** This probability is always 0 or 1 by the 0—1 law of HEWITT and SAVAGE [2, pp. 493 to 494].

*** This probability is always 0 or 1 by the 0—1 law of BLUMENTHAL.

$$\int_0^\infty \frac{dt}{t|\log g(t)|} = +\infty \quad \text{or} \quad < +\infty.$$

(ii) $g(t) \in \mathcal{U}_0$ or \mathcal{Q}_0 according as

$$\int_0^\infty \frac{dt}{t|\log g(t)|} = +\infty \quad \text{or} \quad < +\infty.$$

Thus, for example, we have

Corollary.

$$P_0 \left\{ \lim_{t \rightarrow \infty} \frac{|X(t)|}{t^{-n}} = 0 \right\} = 1,$$

$$P_0 \left\{ \lim_{t \downarrow 0} \frac{|X(t)|}{t^n} = 0 \right\} = 1 \quad \text{for every } n > 0$$

and

$$P_0 \left\{ \lim_{t \rightarrow \infty} \frac{|X(t)|}{t^{-(\log t)^\alpha}} = +\infty \right\} = 1,$$

$$P_0 \left\{ \lim_{t \downarrow 0} \frac{|X(t)|}{t(\log t^{-1})^\alpha} = +\infty \right\} = 1 \quad \text{for every } \alpha > 0.$$

The proof is based on the following lemma which corresponds to lemma 1 of [4]. Its proof runs essentially on the same line as SPITZER's.

Lemma 1. *Let*

$$H(t_1, t_2; \varepsilon) = P_0 \left\{ \inf_{t_1 \leq s \leq t_2} |X(s)| < \varepsilon \right\}.$$

Then

$$H(t_1, t_2; \varepsilon) \sim \frac{\log t_2 - \log t_1}{\log \varepsilon^{-1}}$$

when $\varepsilon \rightarrow 0$.*

Proof. Let $\sigma_\varepsilon = \inf\{t; |X(t)| < \varepsilon\}$, then by the well-known first passage time relation

$$P_x \{X(t) \in (-\varepsilon, \varepsilon)\} = \int_0^t \int_{-\varepsilon}^\varepsilon P_y \{X(t-s) \in (-\varepsilon, \varepsilon)\} \cdot P_x \{\sigma_\varepsilon \in ds, X(\sigma_\varepsilon) \in dy\}.$$

It follows from the shape of the Cauchy distribution that

$$P_0 \{X(t-s) \in (0, 2\varepsilon)\} \leq P_y \{X(t-s) \in (-\varepsilon, \varepsilon)\} \leq P_0 \{X(t-s) \in (-\varepsilon, \varepsilon)\}$$

provided $|y| < \varepsilon$. Hence

$$\begin{aligned} \int_0^t P_0 \{X(t-s) \in (0, 2\varepsilon)\} P_x \{\sigma_\varepsilon \in ds\} &\leq P_x \{X(t) \in (-\varepsilon, \varepsilon)\} \\ &\leq \int_0^t P_0 \{X(t-s) \in (-\varepsilon, \varepsilon)\} P_x \{\sigma_\varepsilon \in ds\}. \end{aligned}$$

Taking Laplace transforms, we have

$$G_\lambda \{0, (0, 2\varepsilon)\} E_x(e^{-\lambda\sigma_\varepsilon}) \leq G_\lambda \{x, (-\varepsilon, \varepsilon)\} \leq G_\lambda \{0, (-\varepsilon, \varepsilon)\} E_x(e^{-\lambda\sigma_\varepsilon}),$$

* The symbol $f(\varepsilon) \sim g(\varepsilon)$ ($\varepsilon \downarrow 0$) means $\lim_{\varepsilon \downarrow 0} \frac{f(\varepsilon)}{g(\varepsilon)} = 1$.

where

$$G_\lambda\{x, E\} = \int_0^\infty e^{-\lambda t} P(t, x, E) dt = \int_E g_\lambda(x - y) dy$$

with

$$g_\lambda(x) = \int_0^\infty e^{-\lambda t} p(t, x) dt = \frac{1}{\pi} \int_0^\infty \frac{\cos x\xi}{\lambda + \xi} d\xi.$$

Therefore

$$\frac{G_\lambda\{x, (-\varepsilon, \varepsilon)\}}{G_\lambda\{0, (-\varepsilon, \varepsilon)\}} \leq \mathbb{E}x(e^{-\lambda\sigma_\varepsilon}) \leq \frac{G_\lambda\{x, (-\varepsilon, \varepsilon)\}}{G_\lambda\{0, (0, 2\varepsilon)\}}. \quad (1)$$

Fix $t_1 > 0$, then $H(t_1, t_1 + t; \varepsilon)$ is an increasing function of $t \geq 0$ whose Stieltjes measure is denoted by $H(t_1, t_1 + dt; \varepsilon)$. By the Markov property,

$$\begin{aligned} H(t_1, t_1 + t; \varepsilon) &= P_0 \left\{ \inf_{t_1 \leq s \leq t_1 + t} |X(s)| < \varepsilon \right\} = \int_{-\infty}^\infty P_x \left\{ \inf_{0 \leq s \leq t} |X(s)| < \varepsilon \right\} p(t_1, x) dx \\ &= \int_{-\infty}^\infty P_x \{ \sigma_\varepsilon < t \} p(t_1, x) dx \end{aligned}$$

and so

$$\int_0^\infty e^{-\lambda t} H(t_1, t_1 + dt; \varepsilon) = \int_{-\infty}^\infty \mathbb{E}x(e^{-\lambda\sigma_\varepsilon}) p(t_1, x) dx.$$

By (1)

$$\frac{\int_{-\infty}^\infty p(t_1, x) G_\lambda\{x, (-\varepsilon, \varepsilon)\} dx}{G_\lambda\{0, (-\varepsilon, \varepsilon)\}} \leq \int_0^\infty e^{-\lambda t} H(t_1, t_1 + dt; \varepsilon) \leq \frac{\int_{-\infty}^\infty p(t_1, x) G_\lambda\{x, (-\varepsilon, \varepsilon)\} dx}{G_\lambda\{0, (0, 2\varepsilon)\}}.$$

Since

$$\begin{aligned} \int_{-\infty}^\infty p(t_1, x) G_\lambda\{x, (-\varepsilon, \varepsilon)\} dx &= \int_0^\infty e^{-\lambda t} dt \int_{-\varepsilon}^\varepsilon \int_{-\infty}^\infty p(t_1, x) p(t, x - y) dx \\ &= \int_0^\infty e^{-\lambda t} dt \int_{-\varepsilon}^\varepsilon p(t_1 + t, y) dy \end{aligned}$$

we have

$$\frac{\int_0^\infty e^{-\lambda t} dt \int_{-\varepsilon}^\varepsilon p(t_1 + t, y) dy}{G_\lambda\{0, (-\varepsilon, \varepsilon)\}} \leq \int_0^\infty e^{-\lambda t} H(t_1, t_1 + dt; \varepsilon) \leq \frac{\int_0^\infty e^{-\lambda t} dt \int_{-\varepsilon}^\varepsilon p(t_1 + t, y) dy}{G_\lambda\{0, (0, 2\varepsilon)\}}.$$

Now, noting

$$p(t_1 + t, y) \leq p(t_1 + t, 0) = \frac{1}{\pi} \frac{1}{t_1 + t} \leq \frac{1}{\pi} \frac{1}{t_1},$$

we have

$$\int_0^\infty e^{-\lambda t} dt \int_{-\varepsilon}^\varepsilon p(t_1 + t, y) dy \sim \frac{2\varepsilon}{\pi} \int_0^\infty e^{-\lambda t} \frac{dt}{t_1 + t} \quad (\varepsilon \downarrow 0)$$

and

$$\begin{aligned} G_\lambda\{0, (-\varepsilon, \varepsilon)\} &= \int_{-\varepsilon}^\varepsilon g_\lambda(x) dx = \frac{1}{\pi} \int_0^\infty d\xi \int_{-\varepsilon}^\varepsilon \frac{\cos x\xi}{\lambda + \xi} dx = \frac{2}{\pi} \int_0^\infty \frac{\sin \varepsilon\xi}{(\lambda + \xi)\xi} d\xi \\ &= \frac{2\varepsilon}{\pi} \int_0^\infty \frac{\sin \xi}{(\lambda\varepsilon + \xi)\xi} d\xi \sim \frac{2\varepsilon}{\pi} \int_0^1 \frac{d\xi}{\lambda\varepsilon + \xi} \sim \frac{2\varepsilon}{\pi} \log \frac{1}{\varepsilon}. \end{aligned}$$

Similarly, we have

$$G_\lambda\{0, (0, 2\varepsilon)\} = \frac{1}{\pi} \int_0^\infty \frac{\sin 2\varepsilon\xi}{(\lambda + \xi)\xi} d\xi = \frac{2\varepsilon}{\pi} \int_0^\infty \frac{\sin \xi}{(2\lambda\varepsilon + \xi)\xi} d\xi \sim \frac{2\varepsilon}{\pi} \int_0^1 \frac{d\xi}{2\lambda\varepsilon + \xi} \sim \frac{2\varepsilon}{\pi} \log \frac{1}{\varepsilon}$$

and so

$$\lim_{\varepsilon \downarrow 0} \left(\log \frac{1}{\varepsilon} \right) \int_0^\infty e^{-\lambda t} H(t_1, t_1 + dt; \varepsilon) = \int_0^\infty e^{-\lambda t} \frac{dt}{t_1 + t}. \tag{2}$$

From (2) we have

$$\lim_{\varepsilon \downarrow 0} \left(\log \frac{1}{\varepsilon} \right) H(t_1, t_1 + t; \varepsilon) = \int_0^t \frac{dt}{t_1 + t} = \log \frac{t_1 + t}{t_1}. \tag{3}$$

The implication “(2) \rightarrow (3)” can be verified, for example, in the following way: let, for some $\lambda_0 > 0$, $\mu_\varepsilon(dt)$ be a measure on $[0, +\infty]$ defined by

$$\mu_\varepsilon(dt) = e^{-\lambda_0 t} \left(\log \frac{1}{\varepsilon} \right) H(t_1, t_1 + dt; \varepsilon), \quad \mu_\varepsilon(\{+\infty\}) = 0,$$

then by (2) $\{\mu_\varepsilon\}$ is uniformly bounded and so for any sequence $\varepsilon_m \downarrow 0$ we can select some subsequence $\varepsilon_n \downarrow 0$ such that

$$\mu_{\varepsilon_n} \rightarrow \mu \quad (\text{weakly})$$

where μ is a measure on $[0, +\infty]$. Then

$$\int_0^\infty e^{-\lambda t} \mu_{\varepsilon_n}(dt) \rightarrow \int_0^\infty e^{-\lambda t} \mu(dt)$$

and by (2)

$$\int_0^\infty e^{-\lambda t} \mu(dt) = \int_0^\infty e^{-(\lambda + \lambda_0)t} \frac{dt}{t_1 + t},$$

and so by the uniqueness theorem for Laplace transforms,

$$\mu(dt) = \frac{e^{-\lambda_0 t}}{t_1 + t} dt.$$

Thus

$$\int_0^t e^{\lambda_0 t} \mu_{\varepsilon_n}(dt) \rightarrow \int_0^t e^{\lambda_0 t} \mu(dt),$$

that is

$$\left(\log \frac{1}{\varepsilon_n} \right) H(t_1, t_1 + t; \varepsilon_n) \rightarrow \int_0^t \frac{dt}{t_1 + t}.$$

Proof of the theorem. We prove (i) first. Note that, by the space-time relation of the Cauchy process, we have

$$H(t_1, t_2; \varepsilon) = H(ct_1, ct_2; c\varepsilon), \quad c > 0.$$

Upon setting

$$E_k = \{w; |X(t)| < tg(t) \quad \text{for some } 2^k \leq t \leq 2^{k+1}\},$$

$$\tilde{E}_k = \{w; |X(t)| < 2^{k+1}g(2^k) \quad \text{for some } 2^k \leq t \leq 2^{k+1}\}$$

and

$$\underline{E}_k = \{w; |X(t)| < 2^k g(2^{k+1}) \text{ for some } 2^k \leq t \leq 2^{k+1}\}$$

for every positive integer k , we see

$$\underline{E}_k \subset E_k \subset \bar{E}_k$$

and by lemma 1

$$P_0(\underline{E}_k) = H(2^k, 2^{k+1}; 2^k g(2^{k+1})) = H(1, 2; g(2^{k+1})) \sim \frac{\log 2}{|\log g(2^{k+1})|}, \quad k \rightarrow \infty.$$

$$P_0(\bar{E}_k) = H(2^k, 2^{k+1}; 2^{k+1} g(2^k)) \sim \frac{\log 2}{|\log g(2^k)|}.$$

Now we shall assume the convergence of the integral in our theorem. Then we have

$$\sum_{k=1}^{\infty} P_0(\bar{E}_k) = O(1) \sum_{k=1}^{\infty} \frac{1}{|\log g(2^k)|} = O(1) \int_1^{\infty} \frac{dt}{t |\log g(t)|} < +\infty,$$

a fortiori $\sum_{k=1}^{\infty} P_0(E_k) < +\infty$ and so $g \in \mathcal{L}_\infty$ by the Borel-Cantelli lemma.

Next we consider the divergent case. We shall then prove $g \in \mathcal{U}_\infty$ by using the lemma in J. LAMPERTI [3]. In the same way as above we get

$$\sum_{k=1}^{\infty} P_0(\underline{E}_k) = O(1) \int_1^{\infty} \frac{dt}{t |\log g(t)|} = \infty.$$

Now define

$$\sigma_j(w) = \begin{cases} \inf\{t; |X(t)| < 2^j g(2^{j+1}), 2^j \leq t \leq 2^{j+1}\} \\ +\infty & \text{if there is no such } t. \end{cases}$$

For any $j < k$, it follows from the strong Markov property that

$$\begin{aligned} P_0(\underline{E}_j \cap \underline{E}_k) &= \int_{2^j}^{2^{j+1}} \int_{R^1} P_y\{|X(t)| < 2^k g(2^{k+1}) \text{ for some } 2^k - s \leq t \leq 2^{k+1} - s\} \\ &\quad \times P_0\{\sigma_j \in ds, X(\sigma_j) \in dy\}. \\ &\leq \int_{R^1} P_y\{|X(t)| < 2^k g(2^{k+1}) \text{ for some } 2^k - 2^{j+1} \leq t \leq 2^{k+1} - 2^j\} \\ &\quad \times P_0\{X(\sigma_j) \in dy; \sigma_j < \infty\}. \end{aligned}$$

Now we have

$$\begin{aligned} &P_y\{|X(t)| < 2^k g(2^{k+1}) \text{ for some } 2^k - 2^{j+1} \leq t \leq 2^{k+1} - 2^j\} \\ &\leq P_0\{|X(t)| < 2^k g(2^{k+1}) \text{ for some } 2^k - 2^{j+1} \leq t \leq 2^{k+1} - 2^j\}. \end{aligned}$$

This fact seems intuitively clear, but we shall prove it in the following lemma 2. Further by lemma 1 there exist positive constants c_1, c_2 and K such that

$$\frac{c_1}{|\log g(2^{k+1})|} \leq P_0(\underline{E}_k) = H(1, 2; g(2^{k+1})) \leq H\left(\frac{1}{2}, 2; g(2^{k+1})\right) \leq \frac{c_2}{|\log g(2^{k+1})|}$$

for all $k \geq K$.

Then, if $k > j \geq K$ and $k \geq j + 2$,

$$\begin{aligned} P_0(\underline{E}_j \cap \underline{E}_k) &\leq P_0(\underline{E}_j) \cdot H(2^k - 2^{j+1}, 2^{k+1} - 2^j; 2^k g(2^{k+1})) \\ &\leq P_0(\underline{E}_j) \cdot H(1/2, 2; g(2^{k+1})) \\ &\leq P_0(\underline{E}_j) \frac{c_2}{|\log g(2^{k+1})|} \leq \frac{c_2}{c_1} P(\underline{E}_j) P(\underline{E}_k). \end{aligned}$$

Then if we set $A_m = \underline{E}_{2^m}$ and $A'_m = \underline{E}_{2^{m+1}}$, the lemma of [3] is applicable to either $\{A_m\}$ or $\{A'_m\}$ and so we have $P_0(\lim_k \underline{E}_k) > 0$, a fortiori $P_0(\lim_k \overline{E}_k) > 0$. This proves $g \in \mathcal{U}_\infty$.

As for the proof of (ii), set for instance

$$\bar{E}_k = \{w; |X(t)| < 2^{-k}g(2^k) \text{ for some } 2^{-k-1} \leq t \leq 2^{-k}\}$$

and note

$$P_0(\bar{E}_k) = H(1/2, 1; g(2^k)) \sim \frac{\log 2}{|\log g(2^k)|}, \quad k \rightarrow \infty.$$

The above proof may be carried over in the same way.

Lemma 2. For fixed $0 < t_1 < t_2$ and $r > 0$,

$$\begin{aligned} & P_x\{|X(t)| < r \text{ for some } t_1 \leq t \leq t_2\} \\ & \leq P_0\{|X(t)| < r \text{ for some } t_1 \leq t \leq t_2\}. \end{aligned}$$

Proof. Let

$$F(x) = P_x\{|X(t)| < r \text{ for some } t_1 \leq t \leq t_2\} = \int \Phi(y) p(t_1, y - x) dy,$$

where

$$\Phi(y) = P_y\{|X(t)| < r \text{ for some } 0 \leq t \leq t_2 - t_1\}.$$

Then $\Phi(y)$ is a monotone decreasing function of $|y|$. In fact, if $0 \leq x \leq y$,

$$\begin{aligned} \Phi(x) &= P_x\{|X(t)| < r \text{ for some } 0 \leq t \leq t_2 - t_1\} \\ &= P_y\left\{|X(t)| < \frac{y}{x}r \text{ for some } 0 \leq t \leq \frac{y}{x}(t_2 - t_1)\right\} \geq \Phi(y) \end{aligned}$$

in view of the space-time relation of the Cauchy process. Therefore

$$\begin{aligned} F(0) - F(x) &= \int_{-\infty}^{\infty} p(t_1, y) (\Phi(y) - \Phi(y - |x|)) dy \\ &= \int_{-\infty}^0 \left[p\left(t_1, \frac{|x|}{2} + y\right) - p\left(t_1, \frac{|x|}{2} - y\right) \right] \left[\Phi\left(\frac{|x|}{2} + y\right) - \Phi\left(\frac{|x|}{2} - y\right) \right] dy \geq 0. \end{aligned}$$

Remark. Applying the above method to the case $\alpha < N$, we find that for fixed $0 < t_1 < t_2$ there exist positive constants K_1, K_2 such that

$$K_1 \varepsilon^{N-\alpha} \leq H(t_1, t_2; \varepsilon) \leq K_2 \varepsilon^{N-\alpha}$$

for sufficiently small ε . In this case we see

$$\lim_{\varepsilon \downarrow 0} \frac{G_\lambda\{0, I'_\varepsilon\}}{G_\lambda\{0, I_\varepsilon\}} < 1$$

where $I_\varepsilon = \{x; |x| < \varepsilon\}$, $I'_\varepsilon = \{x; |x - a| < \varepsilon\}$ and a is a point with $|a| = \varepsilon$ and so we could not find the exact estimate of the form

$$H(t_1, t_2; \varepsilon) \sim K_3 \varepsilon^{N-\alpha} (t_1^{1-N/\alpha} - t_2^{1-N/\alpha}), \quad \varepsilon \downarrow 0.$$

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