# Exercise set $1^*$

# Question 1

Suppose that X is a stable process in any dimension (including the case of a Brownian motion). Show that |X| is a positive self-similar Markov process.

# Solution Question 1

- Temporarily write  $(X_t^{(x)}, t \ge 0)$  in place of  $(X, \mathbb{P}_x)$
- Markov property of X tells us that, for  $s, t \ge 0$ ,

$$X_{t+s}^{(x)} = \tilde{X}_{s}^{(X_{t}^{(x)})},$$

where  $\tilde{X}^{(x)}$  is an independent copy of  $X^{(x)}$ .

• Isotropy implies that

$$|X_{t+s}^{(x)}| = |\tilde{X}_s^{(y)}|_{y=X_t^{(x)}} = {}^d |\tilde{X}_s^{(z)}|_{z=(|X_t^{(x)}|,0,0\cdots,0)}$$

- Hence Markov property holds, strong Markov property (and Feller property) can be developed from this argument
- Self-similarity of |X| follows directly from the self-similarity of X.

### Question 2

Suppose that B is a one-dimensional Brownian motion. Prove that

$$\frac{B_t}{x}\mathbf{1}_{(\underline{B}_t>0)}, \qquad t \ge 0,$$

is a martingale, where  $\underline{B}_t = \inf_{s < t} B_s$ .

# Solution Question 2

- Note that  $(B_t, t \ge 0)$  is a martingale.
- Optional stopping implies that  $(B_{t \wedge \tau_0^-}, t \ge 0)$  is a martingale, where  $\tau_0^- = \inf\{t > 0 : B_t < 0\}$
- Since  $B_{\tau_0^-} = 0$ , it follows that  $B_{t \wedge \tau_0^-} = B_t \mathbf{1}_{(t < \tau_0^-)}, t \ge 0$ .
- Finally note that  $\{\tau_0^- > t\} = \{\underline{B}_t > 0\}$

# Question 3<sup>\*</sup>

Suppose that X is a stable process with two-sided jumps

- Show that the range of the ascending ladder process H, say range(H) has the property that it is equal in law to  $c \times range(H)$ .
- Hence show that, up to a multiplicative constant, the Laplace exponent of H satisfies  $k(\lambda) = \lambda^{\alpha_1}$  for  $\alpha_1 \in (0, 1)$  (and hence the ascending ladder height process is a stable subordinator).
- Use the fact that, up to a multiplicative constant

$$\Psi(z) = |\theta|^{\alpha} (\mathrm{e}^{\pi \mathrm{i}\alpha(\frac{1}{2}-\rho)} \mathbf{1}_{(\theta>0)} + \mathrm{e}^{-\pi \mathrm{i}\alpha(\frac{1}{2}-\rho)} \mathbf{1}_{(\theta<0)}) = \hat{\kappa}(\mathrm{i}z)\kappa(-\mathrm{i}z)$$

to deduce that

$$\kappa(\theta) = \theta^{\alpha \rho} \text{ and } \hat{\kappa}(\theta) = \theta^{\alpha \rho}$$

and that  $0 < \alpha \rho, \alpha \hat{\rho} < 1$ 

• What kind of process corresponds to the case that  $\alpha \rho = 1$ ?

#### Solution Question 3

- Range of H agrees with the range of  $\overline{X}$ . Scaling of X applies to scaling of  $\overline{X}$ , in particular  $(c\overline{X}_{c^{-\alpha}t}, t \ge 0)$  is equal in law to  $(\overline{X}_t, t \ge 0)$ , so that  $c \times \operatorname{range} \overline{X} = \operatorname{range}(c\overline{X}) = \operatorname{range}(\overline{X})$ . And hence the same is true for H.
- The latter is equivalent to the condition that the Laplace exponent of H is proportional to that of cH, i.e  $\kappa(z) = k_c \kappa(cz)$ , for  $z \ge 0$ , where  $k_c > 0$  is a constant that only depends on c. Since  $\kappa(1)$  must be a constant, we see that  $\kappa(z) = \kappa(cz)/\kappa(c)$ . Hence, as  $\kappa$  is increasing, one can easily deduce that  $k(\lambda) = \kappa(1)\lambda^{\alpha_1}$  for some  $\alpha_1 \in [0, 1]$ . In other words, H is a stable subordinator with parameter  $\alpha_1$ . We exclude the case  $\alpha_1 = 0$  since it corresponds to the setting where the range of H is the empty set. A similar argument applied to -X shows that the descending ladder height process must also belong to the class of stable subordinators.
- We therefore assume that (up to multiplicative constants)  $\kappa(z) = z^{\alpha_1}, z \ge 0$ , and  $\hat{\kappa}(z) = z^{\alpha_2}, z \ge 0$ , for some  $\alpha_1, \alpha_2 \in (0, 1]$ . Appealing to the the stable process exponent, we must choose the parameters  $\alpha_1$  and  $\alpha_2$  such that, for example, when z > 0,

$$z^{\alpha} e^{\pi i \alpha (\frac{1}{2} - \rho)} = z^{\alpha_1} e^{-\frac{1}{2}\pi i \alpha_1} \times z^{\alpha_2} e^{\frac{1}{2}\pi i \alpha_2}.$$

Matching radial and angular parts, this is only possible if  $\alpha_1$  and  $\alpha_2$  satisfy

$$\begin{cases} \alpha_1 + \alpha_2 = \alpha, \\ \alpha_1 - \alpha_2 = -\alpha(1 - 2\rho), \end{cases}$$

which gives us  $\alpha_1 = \alpha \rho$  and  $\alpha_2 = \alpha \hat{\rho}$ . As X does not have monotone paths, it is necessarily the case that  $0 < \alpha \rho \leq 1$  and  $0 < \alpha \hat{\rho} \leq 1$ . In conclusion, for  $\theta \geq 0$ ,

$$\kappa(\theta) = \theta^{\alpha \rho} \text{ and } \hat{\kappa}(\theta) = \theta^{\alpha \hat{\rho}}.$$

• When  $\alpha \rho = 1$ , the ascending ladder height process is a pure linear drift. In that case, the range of the maximum process  $\overline{X}$  is  $[0, \infty)$ . The process is spectrally negative.

#### Question 4

Suppose that  $(X, P_x), x > 0$  is a positive self-similar Markov process and let  $\zeta = \inf\{t > 0 : X_t = 0\}$  be the lifetime of X. Show that  $P_x(\zeta < \infty)$  does not depend on x and is either 0 for all x > 0 or 1 for all x > 0.

## Solution Question 4

• We can appeal to the scaling property and write, for all c > 0,

$$\begin{split} \zeta^{(cx)} &= \inf\{t > 0 : X_t^{(cx)} = 0\} \\ &\stackrel{d}{=} c^{\alpha} \inf\{c^{-\alpha}t > 0 : cX_{c^{-\alpha}t}^{(x)} = 0\} \\ &= c^{\alpha}\zeta^{(x)}, \end{split}$$

showing that  $P_x(\zeta < \infty) = P_{cx}(\zeta < \infty)$ , for all x, c > 0, as claimed. Note that this also shows that  $x^{-\alpha}\zeta^{(x)}$  is independent of the value of x.

• Denote by  $p \in [0, 1]$  the common value of the probabilities  $P_x(\zeta < \infty)$ , x > 0. We shall now show that either p = 0 or p = 1. Thanks to the Markov property, we can now write, for all x, t > 0,

$$P_x(t < \zeta < \infty) = E_x(\mathbf{1}_{(t < \zeta)} P_{X_t}(\zeta < \infty)) = p P_x(t < \zeta),$$

and, hence,

$$p = P_x(\zeta \le t) + P_x(t < \zeta < \infty)$$
$$= P_x(\zeta \le t) + p(1 - P_x(\zeta \le t))$$
$$= p + (1 - p)P_x(\zeta < t).$$

This forces us to conclude that either p = 1 or  $P_x(\zeta \le t) = 0$ , for all x, t > 0. In other words, p = 1 or p = 0.

#### Question 5<sup>\*</sup>

Suppose that X is a symmetric stable process in dimension one (in particular  $\rho = 1/2$ ) and that the underlying Lévy process for  $|X_t|\mathbf{1}_{\{t<\tau^{\{0\}}\}}$ , where  $\tau^{\{0\}} = \inf\{t > 0 : X_t = 0\}$ , is written  $\xi$ . (Note the indicator is only needed when  $\alpha \in (1,2)$  as otherwise X does not hit the origin.) Show that (up to a multiplicative constant) its characteristic exponent is given by

$$\Psi(z) = 2^{\alpha} \frac{\Gamma(\frac{1}{2}(-iz+\alpha))}{\Gamma(-\frac{1}{2}iz)} \frac{\Gamma(\frac{1}{2}(iz+1))}{\Gamma(\frac{1}{2}(iz+1-\alpha))}, \qquad z \in \mathbb{R}.$$

[Hint!] Think about what happens after X first crosses the origin and apply the Markov property as well as symmetry. You will need to use the law of the overshoot of X below the origin given a few slides back.

#### Solution Question 5

• The trick is to go back and think about the pssMp  $(X_t \mathbf{1}_{\underline{X}_t > 0}, t \ge 0)$ . Previously we had identified the characteristics exponent of its underlying Lévy process (through the Lamperti transform) as killed Lévy process with exponent

$$\Psi^*(z) = \frac{\Gamma(\alpha - iz)}{\Gamma(\alpha\hat{\rho} - iz)} \frac{\Gamma(1 + iz)}{\Gamma(1 - \alpha\hat{\rho} + iz)}, \qquad z \in \mathbb{R},$$
(1)

which contained a killing rate

$$q^* = \Psi^*(0) = \frac{\Gamma(\alpha)}{\Gamma(\alpha\hat{\rho})\Gamma(1 - \alpha\hat{\rho})}.$$
(2)

(Note in our setting,  $\rho = \hat{\rho} = 1/2$ ).

• Instead of killing the underlying  $\xi$  at rate  $q^*$  (i.e. the stable process goes negative), we want our process to regenerate in  $(0, \infty)$ , corresponding to our underlying Lévy process undergoing a jump at rate  $q^*$ . In other words, the Lévy process corresponding to |X| can be written in the form

$$\xi = \xi^* + \xi^{**}$$

where  $\xi^{**}$  is a compound Poisson process with arrival rate  $q^*$  and jump distribution F, which we are to determine. In particular, if  $\Psi$  is the characteristic exponent of the Lévy process that underlies |X|, then

$$\Psi(\theta) = (\Psi^*(\theta) - q^*) + q^* \int_{\mathbb{R}} (1 - e^{i\theta x}) F(dx) = \Psi^*(\theta) - q^* \int_{\mathbb{R}} e^{i\theta x} F(dx)$$
(3)

• The point of regeneration must correspond to precisely  $-X_{\tau_0^-}$ . In particular, referring to the Lamperti transform, we must have

$$\frac{-X_{\tau_0^-}}{X_{\tau_0^--}} = e^{\Delta},$$

where  $\Delta$  has distribution F.

• We can use the previously given joint overshoot distribution

$$\mathbb{P}_{1}(-X_{\tau_{0}^{-}} \in \mathrm{d}u, X_{\tau_{0}^{-}-} \in \mathrm{d}v) = \frac{\sin(\alpha\pi/2)}{\pi} \frac{\Gamma(\alpha+1)}{\Gamma(\alpha/2)^{2}} \left( \int_{0}^{\infty} \mathbf{1}_{(y \leq 1 \wedge v)} \frac{(1-y)^{\frac{\alpha}{2}-1}(v-y)^{\frac{\alpha}{2}-1}}{(v+u)^{1+\alpha}} \mathrm{d}y \right) \mathrm{d}v \mathrm{d}u$$

and note that (the point of issue x does not matter by scaling)

$$\begin{split} &\int_{\mathbb{R}} \mathrm{e}^{\mathrm{i}\theta x} F(\mathrm{d}x) \\ &= \mathbb{E}_{x} \left[ \left( \frac{-X_{\tau_{0}^{-}}}{X_{\tau_{0}^{-}}} \right)^{\mathrm{i}\theta} \right] = \mathbb{E}_{1} \left[ \left( \frac{-X_{\tau_{0}^{-}}}{X_{\tau_{0}^{-}}} \right)^{\mathrm{i}\theta} \right] \\ &= \frac{\sin(\pi\alpha\hat{\rho})}{\pi} \frac{\Gamma(\alpha+1)}{\Gamma(\alpha\rho)\Gamma(\alpha\hat{\rho})} \int_{0}^{\infty} \int_{0}^{\infty} \int_{0}^{\infty} \mathbf{1}_{(y\leq 1\wedge v)} \frac{u^{\mathrm{i}\theta}(1-y)^{\alpha\hat{\rho}-1}(v-y)^{\alpha\rho-1}}{v^{\mathrm{i}\theta}(v+u)^{1+\alpha}} \mathrm{d}u \mathrm{d}v \mathrm{d}y \\ &= \frac{\sin(\pi\alpha\hat{\rho})}{\pi} \frac{\Gamma(\alpha+1)}{\Gamma(\alpha\rho)\Gamma(\alpha\hat{\rho})} \int_{0}^{1} \int_{y}^{\infty} \int_{0}^{\infty} \frac{u^{\mathrm{i}\theta}(1-y)^{\alpha\hat{\rho}-1}(v-y)^{\alpha\rho-1}}{v^{\mathrm{i}\theta}(v+u)^{1+\alpha}} \mathrm{d}u \mathrm{d}v \mathrm{d}y \end{split}$$
(4)

Note  $\rho = \hat{\rho} = 1/2$ .

• For the innermost integral in (4), substituting w = v/u and appealing to the integral representation of the beta function, we have

$$\int_0^\infty \frac{u^{\mathrm{i}\theta}}{(u+v)^{1+\alpha}} \,\mathrm{d}u = v^{\mathrm{i}\theta-\alpha} \int_0^\infty \frac{w^{\mathrm{i}\theta}}{(1+w)^{1+\alpha}} \,\mathrm{d}w = v^{\mathrm{i}\theta-\alpha} \frac{\Gamma(\mathrm{i}\theta+1)\Gamma(\alpha-\mathrm{i}\theta)}{\Gamma(\alpha+1)}.$$

Substituting z = v/y, the next iterated integral in (4) becomes

$$\int_{y}^{\infty} v^{-\alpha} (v-y)^{\alpha\rho-1} \, \mathrm{d}v = y^{-\alpha\hat{\rho}} \int_{0}^{\infty} \frac{z^{\alpha\rho-1}}{(1+z)^{\alpha}} \, \mathrm{d}z = y^{-\alpha\hat{\rho}} \frac{\Gamma(\alpha\rho)\Gamma(\alpha\hat{\rho})}{\Gamma(\alpha)}$$

Finally, it remains to calculate the resulting outer integral of (4),

$$\int_0^1 y^{-\alpha\hat{\rho}} (1-y)^{\alpha\hat{\rho}-1} \,\mathrm{d}y = \Gamma(1-\alpha\hat{\rho})\Gamma(\alpha\hat{\rho}).$$

Multiplying together these expressions and using the reflection identity for the gamma function  $^1$  we obtain

$$\mathbb{E}_1\left[\left(-\frac{X_{\tau_0^-}}{X_{\tau_0^--}}\right)^{\mathrm{i}\theta}\right] = \frac{\Gamma(\mathrm{i}\theta+1)\Gamma(\alpha-\mathrm{i}\theta)}{\Gamma(\alpha)}.$$
(5)

• Putting (1), (2), (3) and (5) together, together with the reflection formula

$$\Gamma(z)\Gamma(1-z) = \frac{\pi}{\sin(\pi z)},$$

we get (with  $\rho = \hat{\rho} = 1/2$ ),

$$\begin{split} \Psi(\theta) &= \frac{\Gamma(\alpha - \mathrm{i}\theta)}{\Gamma(\frac{\alpha}{2} - \mathrm{i}\theta)} \frac{\Gamma(1 + \mathrm{i}\theta)}{\Gamma(1 - \frac{\alpha}{2} + \mathrm{i}\theta)} - \frac{\Gamma(\alpha)}{\Gamma(\frac{\alpha}{2})\Gamma(1 - \frac{\alpha}{2})} \frac{\Gamma(\mathrm{i}\theta + 1)\Gamma(\alpha - \mathrm{i}\theta)}{\Gamma(\alpha)} \\ &= \frac{\Gamma(\alpha - \mathrm{i}\theta)\Gamma(1 + \mathrm{i}\theta)}{\pi} \left( \sin(\frac{\pi\alpha}{2} - \mathrm{i}\pi\theta) - \sin(\frac{\pi\alpha}{2}) \right). \end{split}$$

Next, recalling that

$$2\sin(\alpha)\sin(\beta) = \cos(\alpha - \beta) - \cos(\alpha + \beta)$$

we can write

$$2\sin(\frac{\pi(\alpha-1)}{2} - i\frac{\pi\theta}{2})\sin(-i\frac{\pi\theta}{2}) = \cos(\frac{\pi\alpha}{2} - \frac{\pi}{2}) - \cos(\frac{\pi\alpha}{2} - \frac{\pi}{2} - \pi i\theta)$$
$$= \sin(\frac{\pi\alpha}{2}) - \sin(\frac{\pi\alpha}{2} - \pi i\theta).$$

At this point, things become nasty! We need to used the Legendre duplication formula,

$$\Gamma(z)\Gamma(z+\frac{1}{2}) = 2^{1-2z}\Gamma(2z),$$

 $<sup>{}^{1}\</sup>Gamma(z)\Gamma(1-z) = \pi/\sin(\pi z)$ 

the reflection formula again and the recursion formula,  $\Gamma(1+z) = z\Gamma(z)$ , to get

$$\begin{split} \Psi(\theta) &= -2 \frac{\Gamma(\alpha - \mathrm{i}\theta)\Gamma(1 + \mathrm{i}\theta)}{\pi} \sin(\frac{\pi(\alpha - 1)}{2} - \mathrm{i}\frac{\pi\theta}{2}) \sin(-\mathrm{i}\frac{\pi\theta}{2}) \\ &= -2 \cdot 2^{-1 + \alpha - \mathrm{i}\theta} \Gamma(\frac{\alpha}{2} - \mathrm{i}\frac{\theta}{2}) \Gamma(\frac{\alpha}{2} + \frac{1}{2} - \mathrm{i}\frac{\theta}{2}) \cdot 2^{-1 + 1 + \mathrm{i}\theta} \Gamma(\frac{1}{2} + \mathrm{i}\frac{\theta}{2}) \Gamma(1 + \mathrm{i}\frac{\theta}{2}) \\ &\qquad \times \frac{1}{\Gamma(\frac{\alpha}{2} - \frac{1}{2} - \mathrm{i}\frac{\theta}{2})\Gamma(\frac{3}{2} - \frac{\alpha}{2} + \mathrm{i}\frac{\theta}{2})} \cdot \frac{\pi}{\Gamma(-\mathrm{i}\frac{\theta}{2})\Gamma(1 + \mathrm{i}\frac{\theta}{2})} \\ &= -\pi 2^{\alpha} \frac{\Gamma(\frac{1}{2}(-\mathrm{i}\theta + \alpha))}{\Gamma(-\frac{1}{2}\mathrm{i}\theta)} \frac{\Gamma(\frac{1}{2}(\mathrm{i}\theta + 1))}{\Gamma(\frac{1}{2}(\mathrm{i}\theta + 1 - \alpha))} \cdot \frac{\Gamma(\frac{\alpha}{2} + \frac{1}{2} - \mathrm{i}\frac{\theta}{2})\Gamma(1 + \mathrm{i}\frac{\theta}{2})}{\Gamma(\frac{\alpha}{2} - \frac{1}{2} - \mathrm{i}\frac{\theta}{2})(\frac{1}{2} - \frac{\alpha}{2} + \mathrm{i}\frac{\theta}{2})\Gamma(1 + \mathrm{i}\frac{\theta}{2})} \\ &= \pi 2^{\alpha} \frac{\Gamma(\frac{1}{2}(-\mathrm{i}\theta + \alpha))}{\Gamma(-\frac{1}{2}\mathrm{i}\theta)} \frac{\Gamma(\frac{1}{2}(\mathrm{i}\theta + 1))}{\Gamma(\frac{1}{2}(\mathrm{i}\theta + 1 - \alpha))}, \\ &= \pi 2^{\alpha} \frac{\Gamma(\frac{1}{2}(-\mathrm{i}\theta + \alpha))}{\Gamma(-\frac{1}{2}\mathrm{i}\theta)} \frac{\Gamma(\frac{1}{2}(\mathrm{i}\theta + 1))}{\Gamma(\frac{1}{2}(\mathrm{i}\theta + 1 - \alpha))}, \end{split}$$

which is the required exponent up to the multiplicative constant  $\pi$ .

## Question 6\*

Use the previous exercise to deduce that the MAP exponent underlying a stable process with two sided jumps is given by

$$\begin{bmatrix} -\frac{\Gamma(\alpha-z)\Gamma(1+z)}{\Gamma(\alpha\hat{\rho}-z)\Gamma(1-\alpha\hat{\rho}+z)} & \frac{\Gamma(\alpha-z)\Gamma(1+z)}{\Gamma(\alpha\hat{\rho})\Gamma(1-\alpha\hat{\rho})} \\ \frac{\Gamma(\alpha-z)\Gamma(1+z)}{\Gamma(\alpha\rho)\Gamma(1-\alpha\rho)} & -\frac{\Gamma(\alpha-z)\Gamma(1+z)}{\Gamma(\alpha\rho-z)\Gamma(1-\alpha\rho+z)} \end{bmatrix},$$

for  $\operatorname{Re}(z) \in (-1, \alpha)$ .

## Solution Question 6

First recall that the matrix exponent takes the specific form

$$\Psi(z) = \begin{bmatrix} \psi_1(z) & 0 \\ 0 & \psi_{-1}(z) \end{bmatrix} + \begin{bmatrix} -Q_{1,-1} & Q_{1,-1} \\ Q_{-1,1} & -Q_{-1,1} \end{bmatrix} \circ \begin{bmatrix} 1 & \mathbf{E}[e^{zU_{1,-1}}] \\ \mathbf{E}[e^{zU_{-1,1}}] & 1 \end{bmatrix}$$
(6)

We can borrow lots of ideas and calculations from Question 5. Until it first crosses the origin, X behaves like  $X \mathbf{1}_{(X>0)}$  and hence we can take

$$\psi_1(z) = -(\Psi^*(-iz) - q^*).$$

Moreover the rate at which the change from the positive to the negative half-line occurs is rate  $q^*$ .

When the process jumps from the positive to negative half line, the process the chain maps from 1 to -1 transferring the left-limit  $X_{\tau_0^-}$  to  $-X_{\tau_0^-}$ . However the actual positioning is  $X_{\tau_0^-}$ , so a multiplicative correction in the radial positioning of

$$\frac{|X_{\tau_0^-}|}{X_{\tau_0^-}} = e^{U_{1,-1}}.$$

From (5), we thus conclude that

$$\mathbf{E}[\mathrm{e}^{\mathrm{i}zU_{1,-1}}] = \frac{\Gamma(\mathrm{i}z+1)\Gamma(\alpha-\mathrm{i}z)}{\Gamma(\alpha)}$$

is needed.

By anti-symmetry (or otherwise noting that the behaviour on the negative half-line is that of -X with the roles of  $\rho$  and  $\hat{\rho}$  interchanged, written  $\rho \leftrightarrow \hat{\rho}$ ). In conclusion we have everything we need to populate the matrices in (6). And hence,

$$\begin{split} \Psi(z) &= \begin{bmatrix} -(\Psi^*(-\mathrm{i}z) - q^*) & 0 \\ & \bullet \end{bmatrix} + \begin{bmatrix} -q^* & q^* \\ & \bullet \end{bmatrix} \circ \begin{bmatrix} 1 & \frac{\Gamma(z+1)\Gamma(\alpha-z)}{\Gamma(\alpha)} \\ & \bullet \end{bmatrix} \\ &= \begin{bmatrix} -\frac{\Gamma(\alpha-z)}{\Gamma(\alpha\hat{\rho}-z)} \frac{\Gamma(1+z)}{\Gamma(1-\alpha\hat{\rho}+z)} & \frac{\Gamma(z+1)\Gamma(\alpha-z)}{\Gamma(\alpha\hat{\rho})\Gamma(1-\alpha\hat{\rho})} \\ & \bullet \end{bmatrix} \\ &= \begin{bmatrix} -\frac{\Gamma(\alpha-z)}{\Gamma(\alpha\hat{\rho}-z)} \frac{\Gamma(1+z)}{\Gamma(1-\alpha\hat{\rho}+z)} & \frac{\Gamma(z+1)\Gamma(\alpha-z)}{\Gamma(\alpha\hat{\rho})\Gamma(1-\alpha\hat{\rho})} \\ & \frac{\Gamma(z+1)\Gamma(\alpha-z)}{\Gamma(\alpha\rho)\Gamma(1-\alpha\rho)} & -\frac{\Gamma(\alpha-z)}{\Gamma(\alpha\rho-z)} \frac{\Gamma(1+z)}{\Gamma(1-\alpha\rho+z)} \end{bmatrix} \end{split}$$

as required.

# Exercise Set 2

## Question 1

Use the fact that the radial part of a *d*-dimensional  $(d \ge 2)$  isotropic stable process has MAP  $(\xi, \Theta)$ , for which the first component is a Lévy process with characteristic exponent given by

$$\Psi(z) = 2^{\alpha} \frac{\Gamma(\frac{1}{2}(-iz+\alpha))}{\Gamma(-\frac{1}{2}iz)} \frac{\Gamma(\frac{1}{2}(iz+d))}{\Gamma(\frac{1}{2}(iz+d-\alpha))}, \qquad z \in \mathbb{R}.$$

to deduce the following facts:

- Irrespective of its point of issue, we have  $\lim_{t\to\infty} |X_t| = \infty$  almost surely.
- By considering the roots of  $\Psi$  show that

$$\exp((\alpha - d)\xi_t), \qquad t \ge 0,$$

is a martingale.

• Deduce that

 $|X_t|^{\alpha-d}, \qquad t \ge 0,$ 

is a martingale.

## Solution Question 1

- One can use the fact that  $\mathbf{E}[\xi_1] = -i\Psi'(0)$  to deduce that  $\lim_{t\to\infty} \xi_t = \infty$  and hence, recalling that |X| is a pssMp, and that only three types of behaviour are possible for pssMp, one has that  $\lim_{t\to\infty} |X_t| = \infty$ .
- If we write  $\psi(\lambda) = -\Psi(-i\lambda) = \log \mathbb{E}[e^{\lambda X_1}]$  for the Laplace exponent of  $\xi$ , then it is well defined for  $\lambda \in (-d, \alpha)$  with roots at  $\lambda = 0$  and  $\lambda = \alpha d$ .
- Note that

$$\exp((\alpha - d)\xi_t), \qquad t \ge 0,$$

is a martingale

• Recalling that  $|X_t| = \exp(\xi_{\varphi_t})$  and that  $\varphi_t$  is an almost surely finite stopping time (because  $\lim_{t\to\infty} \xi_t = \infty$ ) we can deduce that

$$|X_t|^{\alpha-d}, \qquad t \ge 0,$$

is a martingale (effectively invoking an Esscher transform to  $\psi$ ).

## Question 2

Remaining in d-dimensions  $(d \ge 2)$ , recalling that

$$\frac{\mathrm{d}\mathbb{P}_x^{\circ}}{\mathrm{d}\mathbb{P}_x}\bigg|_{\mathcal{F}_t} = \frac{|X_t|^{\alpha-d}}{|x|^{\alpha-d}}, \qquad t \ge 0, x \ne 0,$$

show that under  $\mathbb{P}^{\circ}$ , X is absorbed continuously at the origin in an almost surely finite time.

# Question 2 solution

Note that if  $(\xi^{\circ}, \Theta)$  is the MAP underlying  $(X, \mathbb{P}^{\circ})$ , then it is still the case that  $\xi$  alone is a Lévy process because the change of measure  $\mathbb{P}^{\circ}$  is written in terms of the radial component only. Moreover, the characteristic exponent of  $\xi^{\circ}$  is given by

$$\Psi^{\circ}(z) = 2^{\alpha} \frac{\Gamma(\frac{1}{2}(-iz+d))}{\Gamma(-\frac{1}{2}(iz+\alpha-d))} \frac{\Gamma(\frac{1}{2}(iz+\alpha))}{\Gamma(\frac{1}{2}iz)}, \qquad z \in \mathbb{R}.$$

and one can check that  $\Psi^{\circ}(0) = 0$  and  $\mathbf{E}^{\circ}[\xi_t] = -i\Psi^{\circ\prime}(0) < 0$ . Remember that there are only three categories of pssMp and |X| must still be a pssMp because of isotropy of  $(X, \mathbb{P}^{\circ})$ . So |X| must fit the category of pssMp that is continuously absorbed at the origin.

# Question 3<sup>\*</sup>

Recall the following theorem.

Theorem. Define the function

$$g(x,y) = \pi^{-(d/2+1)} \Gamma(d/2) \sin(\pi\alpha/2) \frac{\left|1 - |x|^2\right|^{\alpha/2}}{\left|1 - |y|^2\right|^{\alpha/2}} |x - y|^{-d}$$

for  $x, y \in \mathbb{R}^d \setminus \mathbb{S}_{d-1}$ . Let

$$^{\oplus} := \inf\{t > 0 : |X_t| < 1\} \text{ and } \tau_a^{\ominus} := \inf\{t > 0 : |X_t| > 1\}.$$

(i) Suppose that |x| < 1, then

$$\mathbb{P}_x(X_{\tau^{\ominus}} \in \mathrm{d}y) = g(x, y)\mathrm{d}y, \qquad |y| \ge 1.$$

(ii) Suppose that |x| > 1, then

$$\mathbb{P}_x(X_{\tau^{\oplus}} \in \mathrm{d}y, \, \tau^{\oplus} < \infty) = g(x, y)\mathrm{d}y, \qquad |y| \le 1.$$

Prove (ii) (i.e. |x| > 1) from the identity in (i) (i.e. |x| < 1).

## Solution Question 3

• Start by noting from the Riesz–Bogdan–Żak transform that, for |x| > 1,

$$\mathbb{P}_x(X_{\tau^\oplus} \in D) = \mathbb{P}^{\circ}_{Kx}(KX_{\tau^\ominus} \in D),$$

where  $Kx = x/|x|^2$ , |Kx - Kz| = |x - z|/|x||z| and  $KD = \{Kx : x \in D\}$ .

• Noting that  $d(Kz) = |z|^{-2d} dz$ , we have

$$\begin{aligned} \mathbb{P}_{x}(X_{\tau^{\oplus}} \in D) \\ &= \int_{KD} \frac{|y|^{\alpha-d}}{|Kx|^{\alpha-d}} g(Kx, y) \mathrm{d}y \\ &= c_{\alpha,d} \int_{KD} |z|^{d-\alpha} |Kx|^{d-\alpha} \frac{|1-|Kx|^{2}|^{\alpha/2}}{|1-|y|^{2}|^{\alpha/2}} |Kx-y|^{-d} \mathrm{d}y \\ &= c_{\alpha,d} \int_{D} |z|^{2d} \frac{|1-|x|^{2}|^{\alpha/2}}{|1-|z|^{2}|^{\alpha/2}} |x-z|^{-d} \mathrm{d}(Kz) \\ &= c_{\alpha,d} \int_{D} \frac{|1-|x|^{2}|^{\alpha/2}}{|1-|z|^{2}|^{\alpha/2}} |x-z|^{-d} \mathrm{d}z \end{aligned}$$