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# A characterization of Markov processes enjoying the time-inversion property

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#### Abstract

We give a necessary and sufficient condition for a homogeneous Markov process taking values in  $\mathbb{R}^n$  to enjoy the time-inversion property of degree  $\alpha$ . The condition sets the shape for the semigroup densities of the process and allows to further extend the class of known processes satisfying the time-inversion property. As an application we recover the result of Watanabe in [24] for continuous and conservative Markov processes on  $\mathbb{R}_+$ . As new examples we generalize Dunkl processes and construct a matrix-valued process with jumps related to the Wishart process by a skew-product representation.

**Keywords**: homogeneous Markov processes; time-inversion property; Bessel processes; Dunkl processes; Wishart processes; semi-stable processes.

Mathematics Subject Classification (2000): 60J25; 60J60; 60J65; 60J99.

## 1 Introduction

Let  $\{(X_t, t \ge 0); (P_x)_{x \in \mathbb{R}^n}\}$  be a homogeneous Markov process with semigroup densities (assumed to exist):

$$P_t(x, dy) = p_t(x, y)dy. (1.1)$$

For all  $x \in \mathbb{R}^n$  and some  $\alpha > 0$ , the process  $\{(t^{\alpha}X_{\frac{1}{t}}, t > 0); P_x\}$  is Markov and in general inhomogeneous.

**Definition 1.1.** The process  $\{(X_t, t \ge 0); P_x\}$  is said to enjoy the *time-inversion property of degree*  $\alpha$  if the Markov process  $\{(t^{\alpha}X_{\frac{1}{2}}, t > 0); P_x\}$  is homogeneous.

Celebrated examples of Markov processes, known to enjoy this property for  $\alpha=1$ , are Brownian motions with drift in  $\mathbb{R}^n$  and Bessel processes with drift (see [21, 24]). Gallardo and Yor [12] recently worked out a sufficient condition on the semigroup densities for a Markov process to enjoy the time-inversion property. Their argument extended the class of processes to processes with jumps such as the Dunkl process [23] and matrix-valued processes such as the Wishart process [3]. The aim of the paper is to show that the former condition is actually necessary and sufficient and to provide some new examples.

Section 2 contains the main theorem of the paper, which is proved in section 3 using straightforward analytical arguments. Section 4 considers an application of the theorem to Markov processes on  $\mathbb{R}_+$ . The result is shown to be strong enough to entirely characterize the class of diffusion processes on  $\mathbb{R}_+$  that enjoy the time-inversion property and thus provide a different proof than that

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of Watanabe in [24]. Section 5 gives new examples of processes that enjoy the time-inversion property. We review how the generalized Dunkl process fits the requirements of the theorem and then introduce a matrix-valued process with jumps. The relation of the latter process to the Wishart process mimics the relation between the one-dimensional Dunkl and Bessel processes.

# 2 Markov processes which enjoy time-inversion

Fix  $x \in \mathbb{R}^n$ . Recall that under  $P_x$ , the process  $(t^{\alpha}X_{\frac{1}{t}}, t > 0)$  is Markov, inhomogeneous, with transitional probability densities  $q_{s,t}^{(x)}(z,y)$ ,  $(s < t; z, y \in \mathbb{R}^n)$ , which satisfy the following relation:

$$E_x \left[ f(t^{\alpha} X_{\frac{1}{t}}) \mid s^{\alpha} X_{\frac{1}{s}} = z \right] = \int dy \ f(y) \ q_{s,t}^{(x)}(z,y)$$
 (2.1)

where

$$q_{s,t}^{(x)}(a,b) = t^{-n\alpha} \frac{p_{\frac{1}{t}}\left(x, \frac{b}{t^{\alpha}}\right) p_{\frac{1}{s} - \frac{1}{t}}\left(\frac{b}{t^{\alpha}}, \frac{a}{s^{\alpha}}\right)}{p_{\frac{1}{s}}\left(x, \frac{a}{s^{\alpha}}\right)}.$$
 (2.2)

Throughout the paper, we assume the semigroup densities  $p_t(x, y)$  to be positive over the domain and regular enough to be at least twice differentiable in the space and time variables.

The process  $(t^{\alpha}X_{\frac{1}{t}}, t > 0)$  is not uniquely defined from the knowledge of the semigroup densities  $p_t(x, y)$ . Actually, there exists at least one transformation that leaves the semigroup densities  $q_{s,t}^{(x)}(a, b)$  unchanged: Doob's h-transform.

**Definition 2.1.** Doob's h-transform is the transformation

$$T_h: \mathcal{P}_x|_{\mathcal{F}_t} \mapsto \frac{h(X_t)}{h(x)} e^{-\nu t} \mathcal{P}_x|_{\mathcal{F}_t}.$$
 (2.3)

for some function h and some constant  $\nu > 0$ .

This remark leads to our first two assertions.

**Proposition 2.2.** Two processes related by h-transforms yield the same process by time-inversion.

**Proposition 2.3.**  $T_h$  is an equivalence relation over the class of homogeneous Markov processes.

It is hence legitimate to research for a criterium to classify all processes that enjoy the time-inversion property up to h-transforms. The following theorem gives a concise statement of our main result:

**Theorem 2.4.** The Markov process  $(t^{\alpha}X_{\frac{1}{t}}, t > 0)$  is homogeneous if and only if the semigroup densities of  $(X_t, t \geq 0)$  are of the form:

$$p_t(x,y) = t^{-\frac{n\alpha}{2}} \Phi\left(\frac{x}{t^{\frac{\alpha}{2}}}, \frac{y}{t^{\frac{\alpha}{2}}}\right) \theta\left(\frac{y}{t^{\frac{\alpha}{2}}}\right) \exp\left\{\rho\left(\frac{x}{t^{\frac{\alpha}{2}}}\right) + \rho\left(\frac{y}{t^{\frac{\alpha}{2}}}\right)\right\}$$
(2.4)

or if they are in h-transform relationship with it. The functions  $\Phi, \theta, \rho$  have the following properties for  $\lambda > 0$ :

- 1.  $\Phi(\lambda x, y) = \Phi(x, \lambda y)$ ;
- 2.  $\theta(\lambda y) = \lambda^{\beta} \theta(y)$  for some  $\beta \in \mathbb{R}$ ;
- 3.  $\rho(\lambda x) = \lambda^{\frac{2}{\alpha}} \rho(x)$ :

Moreover, if the symmetry condition  $\Phi(x,y) = \Phi(y,x)$  is satisfied, then the semigroup densities are related as follows:

$$q_t^{(x)}(a,b) = \frac{\Phi(x,b)}{\Phi(x,a)} \exp\left\{t\rho(x)\right\} p_t(a,b). \tag{2.5}$$

**Definition 2.5.** A Markov process  $(X_t, t \ge 0)$  is called semi-stable with index  $\gamma$  in the sense of Lamperti [18] if:

$$\left\{ (X_{ct}, t \ge 0); \mathcal{P}_x \right\} \stackrel{(d)}{=} \left\{ (c^{\gamma} X_t, t \ge 0); \mathcal{P}_{x/c^{\gamma}} \right\}. \tag{2.6}$$

As a consequence, the semigroup densities of a semi-stable process with index  $\gamma$  have the following property:

$$p_t(x,y) = t^{-n\gamma} p_1\left(\frac{x}{t^{\gamma}}, \frac{y}{t^{\gamma}}\right). \tag{2.7}$$

Remarking that the expression for the semigroup densities in Theorem 2.4 satisfies this property for  $\gamma = \alpha/2$  yields the following corollary:

Corollary 2.6. A Markov process that enjoys the time-inversion property of degree  $\alpha$  is a semi-stable process with index  $\alpha/2$ , or is in h-transform relationship with it. The converse is not true.

# 3 Proof of the Theorem

## 3.1 Sufficiency

If  $p_t(x,y)$  satisfies the condition (2.4), then from formula (2.2) the semigroup densities  $q_{s,t}^{(x)}(a,b)$  can be written as

$$q_{s,t}^{(x)}(a,b) = (t-s)^{-\frac{n\alpha}{2}} \prod_{i=1}^{3} R_{s,t}^{(i)}$$

where

$$R_{s,t}^{(1)} = \Phi\left(x \ t^{\frac{\alpha}{2}}, \frac{b}{t^{\alpha}} \ t^{\frac{\alpha}{2}}\right) \Phi\left(\frac{b}{t^{\alpha}} \left(\frac{st}{t-s}\right)^{\frac{\alpha}{2}}, \frac{a}{s^{\alpha}} \left(\frac{st}{t-s}\right)^{\frac{\alpha}{2}}\right) \Phi\left(x \ s^{\frac{\alpha}{2}}, \frac{a}{s^{\alpha}} \ s^{\frac{\alpha}{2}}\right)^{-1}$$

$$R_{s,t}^{(2)} = \theta\left(\frac{b}{t^{\alpha}} \ t^{\frac{\alpha}{2}}\right) \theta\left(\frac{a}{s^{\alpha}} \left(\frac{st}{t-s}\right)^{\frac{\alpha}{2}}\right) \theta\left(\frac{a}{s^{\alpha}} \ s^{\frac{\alpha}{2}}\right)^{-1}$$

$$R_{s,t}^{(3)} = \exp\left\{\rho\left(x \ t^{\frac{\alpha}{2}}\right) + \rho\left(\frac{b}{t^{\alpha}} \ t^{\frac{\alpha}{2}}\right) + \rho\left(\frac{b}{t^{\alpha}} \left(\frac{st}{t-s}\right)^{\frac{\alpha}{2}}\right) + \rho\left(\frac{a}{s^{\alpha}} \left(\frac{st}{t-s}\right)^{\frac{\alpha}{2}}\right) - \rho\left(\frac{a}{s^{\alpha}} \ s^{\frac{\alpha}{2}}\right)\right\}$$

$$-\rho\left(x \ s^{\frac{\alpha}{2}}\right) - \rho\left(\frac{a}{s^{\alpha}} \ s^{\frac{\alpha}{2}}\right)\right\}$$

Using the properties of  $\Phi, \theta, \rho$  described in Theorem 2.4, we obtain:

$$\begin{split} R_{s,t}^{(1)} &= \frac{\Phi\left(x,b\right)}{\Phi\left(x,a\right)} \, \Phi\left(\frac{b}{(t-s)^{\frac{\alpha}{2}}}, \frac{a}{(t-s)^{\frac{\alpha}{2}}}\right) \\ R_{s,t}^{(2)} &= \theta\left(\frac{b}{(t-s)^{\frac{\alpha}{2}}}\right) \\ R_{s,t}^{(3)} &= \exp\left\{(t-s)\rho(x) + \rho\left(\frac{a}{(t-s)^{\frac{\alpha}{2}}}\right) + \rho\left(\frac{b}{(t-s)^{\frac{\alpha}{2}}}\right)\right\}. \end{split}$$

Hence there is no separate dependence on s and t, but only on the difference t-s, which allows to conclude that

$$q_{s,t}^{(x)}(a,b) = q_{t-s}^{(x)}(a,b)$$

and proves homogeneity for the process  $(t^{\alpha}X_{\frac{1}{t}}, t > 0)$ . If in addition  $\Phi(x, y) = \Phi(y, x)$ , we obtain (2.5).

### 3.2 Necessity

For simplicity of notation, we prove the necessity of condition (2.4) in the case  $\alpha = 1$ . The extension to  $\alpha > 0$  is immediate by change of variables  $X_t \mapsto X_t^{\alpha}$ . Recall first the following definition of homogeneous functions:

**Definition 3.1.** A function  $f: \mathbb{R}^n \to \mathbb{R}$  that satisfies for  $x = (x_1, \dots, x_n), \forall \lambda \in \mathbb{R}_+$ 

$$f(\lambda x) = \lambda^{\beta} f(x)$$

is called homogeneous of degree  $\beta \in \mathbb{R}$ . If  $f \in C^1(\mathbb{R}^n)$ , Euler's Homogeneous Function Theorem gives a necessary and sufficient condition for the function f(x) to be homogeneous:

$$\sum_{i=1}^{n} x_i \frac{\partial}{\partial x_i} f(x) = \beta f(x).$$

Consider the function of 2n+1 variables:  $l(x,y,t) = \ln(p_t(x,y)), \ x,y \in \mathbb{R}^n$ . From (2.2), l must satisfy for s=t-h:

$$\frac{\partial}{\partial t} \left[ n \ln \frac{1}{t} + l \left( x, \frac{b}{t}, \frac{1}{t} \right) + l \left( \frac{b}{t}, \frac{a}{t-h}, \frac{1}{t-h} - \frac{1}{t} \right) - l \left( x, \frac{a}{t-h}, \frac{1}{t-h} \right) \right] = 0. \tag{3.1}$$

#### **3.2.1** The kernel $\Phi(x,y)$

Taking derivatives with respect to  $b_i$  and  $a_j$  for some  $0 \le i, j \le n$  yields:

$$\frac{\partial}{\partial b_i} \frac{\partial}{\partial a_j} \frac{\partial}{\partial t} \ l\left(\frac{b}{t}, \frac{a}{t-h}, \frac{1}{t-h} - \frac{1}{t}\right) = 0.$$

If we set  $\phi(x, y, t) = \frac{\partial}{\partial x_i} \frac{\partial}{\partial y_j} l(x, y, t)$ , the latter becomes

$$-\frac{1}{t^2(t-h)}\left(\phi+\frac{b}{t}\cdot\nabla_1\phi\right)-\frac{1}{t(t-h)^2}\left(\phi+\frac{a}{t-h}\cdot\nabla_2\phi\right)+\frac{1}{t(t-h)}\left(\frac{1}{t^2}-\frac{1}{(t-h)^2}\right)\partial\phi=0,$$

with the notation  $\nabla_1 = (\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n})^T$ ,  $\nabla_2 = (\frac{\partial}{\partial y_1}, \dots, \frac{\partial}{\partial y_n})^T$  and  $\phi = \phi\left(\frac{b}{t}, \frac{a}{t-h}, \frac{1}{t-h} - \frac{1}{t}\right)$ . For clarity, we change variables to

$$z_1 = \frac{b}{t}$$
 ,  $z_2 = \frac{a}{t-h}$  ,  $t_1 = -\frac{1}{t}$  ,  $t_2 = \frac{1}{t-h}$ .

Then  $\phi = \phi(z_1, z_2, t_1 + t_2)$  and

$$t_1\bigg(\phi + z_1 \cdot \nabla_1 \phi + t_1 \,\partial\phi\bigg) = t_2\bigg(\phi + z_2 \cdot \nabla_2 \phi + t_2 \,\partial\phi\bigg),$$

or equivalently

$$t_1 \left( \phi + z_1 \cdot \nabla_1 \phi + (t_1 + t_2) \, \partial \phi \right) = t_2 \left( \phi + z_2 \cdot \nabla_2 \phi + (t_1 + t_2) \, \partial \phi \right). \tag{3.2}$$

We change variables once more:  $u = \frac{t_1}{t_2}, v = t_1 + t_2$  to get:

$$\phi + z_1 \cdot \nabla_1 \phi + v \,\partial \phi = u \big( \phi + z_2 \cdot \nabla_2 \phi + v \,\partial \phi \big). \tag{3.3}$$

This gives the following proposition.

**Proposition 3.2.** Equation (3.3) is satisfied if and only if

$$\phi(\lambda x, \mu y, \lambda \mu t) = \frac{1}{\lambda \mu} \phi(x, y, t). \tag{3.4}$$

*Proof.* As the LHS of equation (3.3) is independent of u, one can readily take the equivalent condition:

$$\phi + z_1 \cdot \nabla_1 \phi + v \,\partial \phi = 0, \tag{3.5}$$

$$\phi + z_2 \cdot \nabla_2 \phi + v \,\partial \phi = 0. \tag{3.6}$$

Let  $g(\lambda) = \phi(\lambda z_1, z_2, \lambda v)$  and  $h(\mu) = \phi(z_1, \mu z_2, \mu v)$ . Equation (3.5) implies  $g(\lambda) + \lambda g'(\lambda) = 0$ , which solves to  $g(\lambda) = \lambda^{-1}g(1)$  and hence

$$\phi(\lambda z_1, z_2, \lambda v) = \frac{1}{\lambda} \phi(z_1, z_2, v).$$

Equation (3.6) implies  $h(\mu) + \mu h'(\mu) = 0$ , which solves to  $h(\mu) = \mu^{-1}h(1)$  and hence

$$\phi(z_1, \mu z_2, \mu v) = \frac{1}{\mu} \phi(z_1, z_2, v).$$

Combining the latter two equations yields (3.4).

Conversely, if g is homogeneous of degree -1 and h is homogeneous of degree -1, we get:

$$g(\lambda) + \lambda g'(\lambda) = h(\mu) + \mu h'(\mu) = 0,$$

which is equivalent to (3.5) and (3.6) and concludes the proof.

Note that the previous result remains valid for all second derivatives of l(x, y, t) with respect to  $x_i$  and  $y_j$ ,  $\forall i, j \in \{1, ..., n\}$ . By integration over  $x_i$  and  $y_j$ , one can thus already make an assumption on the general shape of the function l:

$$l(x, y, t) = k(x, y, t) + \varphi_1(x, t) + \varphi_2(y, t) + h_1(x) + h_2(y) + \tau(t), \tag{3.7}$$

where  $k: \mathbb{R}^{2n+1} \to \mathbb{R}$ ,  $\varphi_1, \varphi_2: \mathbb{R}^{n+1} \to \mathbb{R}$ ,  $\tau: \mathbb{R} \to \mathbb{R}$  and  $h_1, h_2: \mathbb{R}^n \to \mathbb{R}$ . Moreover, the kernel k(x, y, t) must satisfy the following property:

$$k(\lambda x, \mu y, \lambda \mu t) = k(x, y, t).$$

This is immediate from (3.4) and

$$\begin{split} \phi(\lambda x, \mu y, \lambda \mu t) &= \frac{\partial}{\partial \lambda x_i} \frac{\partial}{\partial \mu y_j} l(\lambda x, \mu y, \lambda \mu t) = \frac{\partial}{\partial \lambda x_i} \frac{\partial}{\partial \mu y_j} k(\lambda x, \mu y, \lambda \mu t) = \frac{1}{\lambda \mu} \frac{\partial}{\partial x_i} \frac{\partial}{\partial y_j} k(\lambda x, \mu y, \lambda \mu t), \\ \frac{1}{\lambda \mu} \phi(x, y, t) &= \frac{1}{\lambda \mu} \frac{\partial}{\partial x_i} \frac{\partial}{\partial y_j} l(x, y, t) = \frac{1}{\lambda \mu} \frac{\partial}{\partial x_i} \frac{\partial}{\partial y_j} k(x, y, t). \end{split}$$

This property is actually equivalent to

$$k\left(\lambda \frac{x}{\sqrt{t}}, \frac{y}{\sqrt{t}}, 1\right) = k\left(\frac{x}{\sqrt{t}}, \lambda \frac{y}{\sqrt{t}}, 1\right)$$

Setting  $\Phi\left(\frac{x}{\sqrt{t}}, \frac{y}{\sqrt{t}}\right) = \exp k(x, y, t)$  completes the proof of the first condition of the theorem, i.e.

$$\Phi(\lambda x, y) = \Phi(x, \lambda y).$$

#### **3.2.2** The function $\rho(x)$

We now focus on the functions  $\varphi_1(x,t), \varphi_2(y,t)$ . From (3.1), they must satisfy

$$\begin{split} \frac{\partial}{\partial t} & \left[ \varphi_1 \left( x, \frac{1}{t} \right) + \varphi_2 \left( \frac{b}{t}, \frac{1}{t} \right) + \varphi_1 \left( \frac{b}{t}, \frac{1}{t-h} - \frac{1}{t} \right) + \varphi_2 \left( \frac{a}{t-h}, \frac{1}{t-h} - \frac{1}{t} \right) \right. \\ & \left. - \varphi_1 \left( x, \frac{1}{t-h} \right) - \varphi_2 \left( \frac{a}{t-h}, \frac{1}{t-h} \right) \right] = 0. \end{split} \tag{3.8}$$

Recall that the variables a, b, x are independent from each other and that  $\varphi_1, \varphi_2$  are defined to be explicitly dependent on their variables. So taking the derivative of (3.8) with respect to  $x_i$  gives

$$\frac{1}{t^2} \partial_t \psi_1 \left( x, \frac{1}{t} \right) = \frac{1}{(t-h)^2} \partial_t \psi_1 \left( x, \frac{1}{t-h} \right)$$

where  $\psi_1(x,t) = \frac{\partial}{\partial x_i} \varphi_1(x,t)$ . Hence  $\psi_1(x,t)$  must be of the form  $\psi_1(x,t) = \frac{1}{t} \psi_{11}(x) + \psi_{12}(x)$ , which implies

$$\varphi_1(x,t) = \frac{1}{t}\varphi_{11}(x) + \varphi_{12}(x).$$

We omit the second term as it can always be recast as part of h(x) in (3.7). There are so far no further conditions to add on the function  $\varphi_{11}(x)$ .

Taking the derivative of (3.8) with respect to  $b_i$  leads to

$$\frac{\partial}{\partial b_i} \frac{\partial}{\partial t} \left[ \varphi_2 \left( \frac{b}{t}, \frac{1}{t} \right) + \frac{t(t-h)}{h} \varphi_{11} \left( \frac{b}{t} \right) \right] = 0,$$

which develops to

$$-\frac{1}{t^2}\partial_i\varphi_2 - \frac{b}{t^3}\cdot\nabla\partial_i\varphi_2 - \frac{1}{t^3}\partial_t\partial_i\varphi_2 + \frac{2t-h}{ht}\partial_i\varphi_{11} - \frac{t(t-h)}{h}\frac{1}{t^2}\partial_i\varphi_{11} - \frac{t(t-h)}{h}\frac{b}{t^3}\cdot\nabla\partial_i\varphi_{11} = 0.$$

Recall that  $\psi_{11}(x) = \frac{\partial}{\partial x_i} \varphi_{11}(x)$  and set  $\psi_2(x,t) = \frac{\partial}{\partial x_i} \varphi_2(x,t)$ . In this notation, we get

$$-\frac{1}{t}\psi_2 - \frac{b}{t^2} \cdot \nabla \psi_2 - \frac{1}{t^2}\partial_t \psi_2 + \frac{2t-h}{h}\psi_{11} - \frac{t(t-h)}{h}\frac{1}{t}\psi_{11} - \frac{t(t-h)}{h}\frac{b}{t^2} \cdot \nabla \psi_{11} = 0.$$

Let  $z = \frac{b}{t}$ ,  $t_1 = \frac{1}{t}$ ,  $t_2 = \frac{1}{t-h}$ , then we have

$$\psi_2(z,t_1) + z \cdot \nabla \psi_2(z,t_1) + t_1 \partial_t \psi_2(z,t_1) = \frac{t_2}{t_1(t_2 - t_1)} \psi_{11}(z) - \frac{1}{t_2 - t_1} z \cdot \nabla \psi_{11}(z).$$

As the LHS is independent of  $t_2$ , so must be the RHS, which implies

$$z \cdot \nabla \psi_{11}(z) = \psi_{11}(z).$$

Hence  $\psi_{11}$  is homogeneous of degree 1. In that case,

$$\psi_2(z, t_1) + z \cdot \nabla \psi_2(z, t_1) + t_1 \partial_t \psi_2(z, t_1) = \frac{1}{t_1} \psi_{11}(z).$$

We solve the equation for  $g(\lambda) = \psi_2(\lambda z, \lambda t_1)$  to get  $g(\lambda) = \frac{1}{t_1}\psi_{11}(z) + \frac{1}{\lambda}c(z, t_1)$ , where  $c(z, t_1)$  turns out to be a homogeneous function of degree -1. Integration over  $z_i$  yields some conditions for  $\varphi_2(z,t)$ . Once again, since the index  $i \in \{1,\ldots,n\}$  was chosen arbitrarily, we obtain the following characteristics:

$$\varphi_1(\lambda z, \lambda t) = \frac{\varphi_{11}(\lambda z)}{\lambda t}$$
 with  $\varphi_{11}(\lambda z) = \lambda^2 \varphi_{11}(z)$ 

and

$$\varphi_2(\lambda z, \lambda t) = \frac{\varphi_{21}(\lambda z)}{\lambda t} + \hat{c}(\lambda z, \lambda t),$$

where  $\varphi_{21}(\lambda z) = \lambda^2 \varphi_{21}(z)$  and even  $\varphi_{21}(z) = \varphi_{11}(z)$  without loss of generality. Let  $\rho(z) = \varphi_{11}(z) = \varphi_{12}(z)$ , then we recover the third condition of the theorem, that is

$$\rho(\lambda z) = \lambda^2 \rho(z).$$

#### **3.2.3** The function $\theta(y)$

In order to further investigate the properties of  $\hat{c}(z,t)$ , we derive (3.8) by  $a_i$ :

$$\frac{\partial}{\partial a_i} \frac{\partial}{\partial t} \left[ \frac{1}{h} \varphi(a) + \hat{c} \left( \frac{a}{t-h}, \frac{1}{t-h} - \frac{1}{t} \right) - \hat{c} \left( \frac{a}{t-h}, \frac{1}{t-h} \right) \right] = 0.$$

This leads to

$$c(z, t_2 - t_1) + z \cdot \nabla c(z, t_2 - t_1) + \frac{t_1^2 - t_2^2}{t_2} \partial_t c(z, t_2 - t_1) = c(z, t_2) + z \cdot \nabla c(z, t_2) + t_2 \partial_t c(z, t_2).$$

Since c(z,t) is homogeneous of degree -1, the RHS is zero. This only proves that  $\frac{\partial}{\partial t}c(z,t) = \frac{\partial}{\partial t}\frac{\partial}{\partial z_i}\hat{c}(z,t) = 0, \forall i$ . Hence,  $\hat{c}(z,t) = \hat{c}_1(z) + \hat{c}_2(t)$ . Going back to equation (3.8) derived by  $b_i$  gives this time for  $\hat{c}(z,t)$ ,

$$\frac{\partial}{\partial b_i} \frac{\partial}{\partial t} \left[ \hat{c} \left( \frac{b}{t}, \frac{1}{t} \right) \right] = \frac{\partial}{\partial b_i} \frac{\partial}{\partial t} \left[ \hat{c}_1 \left( \frac{b}{t} \right) \right] = 0,$$

which clearly implies  $\hat{c}_1(\frac{b}{t}) = \hat{c}_{11}(b) + \hat{c}_{12}(\frac{1}{t})$ . Using equation (3.8) once more gives

$$\frac{\partial}{\partial t} \left[ \hat{c}_{12} \left( \frac{1}{t} \right) + \hat{c}_2 \left( \frac{1}{t} \right) + \hat{c}_2 \left( \frac{1}{t-h} - \frac{1}{t} \right) - \hat{c}_2 \left( \frac{1}{t-h} \right) \right] = 0, \tag{3.9}$$

which is equivalent to

$$t_1^2 \ \hat{c}_{12}'(t_1) + t_1^2 \ \hat{c}_{2}'(t_1) + (t_1 + t_2)(t_2 - t_1) \ \hat{c}_{2}'(t_2 - t_1) - t_2^2 \ \hat{c}_{2}'(t_2) = 0.$$

We set  $g(t) = t\hat{c}'_2(t)$  so that

$$t_1^2 \hat{c}'_{12}(t_1) + t_1 g(t_1) = t_2 (g(t_2) - g(t_2 - t_1)) - t_1 g(t_2 - t_1).$$

Assuming  $\lim_{t_1\to 0}\frac{LHS}{t_1}=\frac{\beta}{2}\in\mathbb{R}$  with  $|\beta|<\infty$ , we divide the equation by  $t_1$  and take the limit as  $t_1\to 0$ . This yields  $\frac{1}{2}\beta=t_2g'(t_2)-g(t_2)$ , which solves for  $g(t)=-\frac{1}{2}\beta+\gamma t,\ \gamma\in\mathbb{R}$ . Hence,

$$\hat{c}_2(t) = -\frac{1}{2}\beta \ln t + \gamma t$$
 and  $\hat{c}_{12}(t) = \beta \ln t$ ,

where any additional constant term has been set to 0. It will be convenient to set  $\gamma = 0$ . A term  $\gamma \neq 0$  corresponds to the independent case of an affine function of t in the sum (3.7) and is included in the discussion for  $\tau(t)$ . To summarize, the function  $\hat{c}(z,t)$  must be expressed as

$$\hat{c}(z,t) = \hat{c}_1(z) - \frac{\beta}{2} \ln t,$$

where  $c_1(z)$  has the following property:  $\hat{c}_1(\lambda z) = \hat{c}_1(z) + \beta \ln \lambda$ . Let  $\theta(z) = \exp \hat{c}_1(z)$ . We then recover condition 2 of the theorem, i.e.

$$\theta(\lambda z) = \lambda^\beta \ \theta(z).$$

#### **3.2.4** The functions $h_1(x), h_2(y)$

The obvious solution to (3.1) in this case is  $h(z) = h_1(z) = -h_2(z)$ . exp h hence defines an h-transform, which is neglected since we classify all processes up to h-transforms.

#### **3.2.5** The function $\tau(t)$

It remains to explicitly formulate the form of the function  $\tau: \mathbb{R}_+ \to \mathbb{R}$  that is defined to satisfy

$$\frac{\partial}{\partial t} \left[ -n \ln t + \tau \left( \frac{1}{t} \right) + \tau \left( \frac{1}{t-h} - \frac{1}{t} \right) - \tau \left( \frac{1}{t-h} \right) \right] = 0.$$

But this equation is similar to (3.9), so we immediately have the solution  $(\gamma \in \mathbb{R})$ :

$$\tau(t) = -\frac{n}{2}\ln t + \gamma t.$$

Once again  $e^{\gamma t}$  can be included in an h-transform, so  $\gamma$  is set to 0 and  $e^{\tau(t)} = t^{-\frac{n}{2}}$  gives the first factor in the semigroup densities (2.4).

#### **3.2.6** The general case $\alpha > 0$

For  $\alpha > 0$ , we use the change of variables  $X_t \mapsto X_t^{\alpha}$ . The semigroup densities for  $X_t^{\alpha}$  become

$$p_t^{\alpha}(x,y) = p_t\left(x^{\frac{1}{\alpha}}, y^{\frac{1}{\alpha}}\right) J(y)$$

where J(y) is the Jacobian of the inverse transformation, that is  $J(y) = \alpha^{-n} y_1^{\frac{1}{\alpha}-1} \cdots y_n^{\frac{1}{\alpha}-1}$ , and we use the slight abuse of notation  $x^{\frac{1}{\alpha}} = \left(x_1^{\frac{1}{\alpha}}, \dots, x_n^{\frac{1}{\alpha}}\right)$ . The semigroup densities can be recast into

$$p_t^{\alpha}(x,y) = t^{-\frac{n\alpha}{2}} \Phi^{\alpha}\left(\frac{x}{t^{\frac{\alpha}{2}}}, \frac{y}{t^{\frac{\alpha}{2}}}\right) \theta^{\alpha}\left(\frac{y}{t^{\frac{\alpha}{2}}}\right) \exp\left\{\rho^{\alpha}\left(\frac{x}{t^{\frac{\alpha}{2}}}\right) + \rho^{\alpha}\left(\frac{y}{t^{\frac{\alpha}{2}}}\right)\right\}$$

where  $\Phi^{\alpha}(x,y) = \Phi\left(x^{\frac{1}{\alpha}},y^{\frac{1}{\alpha}}\right)$  satisfies condition 1,  $\theta^{\alpha}(y) = \theta\left(y^{\frac{1}{\alpha}}\right)J(y)$  satisfies condition 2 for  $\bar{\beta} = \frac{\beta+n}{\alpha} - n$  and  $\rho^{\alpha}(x) = \rho\left(y^{\frac{1}{\alpha}}\right)$  satisfies condition 3 of equation (2.4).

# 4 Application to diffusions on $\mathbb{R}_+$

The case of the diffusions on  $\mathbb{R}_+$  was entirely characterized by Watanabe in [24]. It was shown that only Bessel processes in the wide sense (which we recall the definition below) enjoy the time-inversion property of degree 1.

**Definition 4.1.** For some  $\nu > -1$  and  $c \ge 0$ , the diffusion process generated by

$$\mathcal{L} = \frac{1}{2} \frac{\partial^2}{\partial x^2} + \left( \frac{2\nu + 1}{2x} + \frac{h'_c(x)}{h_c(x)} \right) \frac{\partial}{\partial x}$$
 (4.1)

is called Bessel process in the wide sense. The function  $h_c(x)$  is given by

$$h_c(x) = 2^{\nu} \Gamma(\nu + 1) (\sqrt{2c} \ x)^{-\nu} I_{\nu}(\sqrt{2c} \ x),$$
 (4.2)

where  $I_{\nu}$  is the modified Bessel function.

Remark 4.2. The Bessel process in the wide sense is in h-transform relationship, for  $h \equiv h_c$ , with the Bessel process.

We show that the result in [24] is a consequence of Theorem 2.4, which has the following onedimensional formulation: **Theorem 4.3.** The Markov process  $(t^{\alpha}X_{\frac{1}{t}}, t > 0)$  on  $\mathbb{R}_{+}$  is homogeneous if and only if the semi-group densities of  $(X_{t}, t \geq 0)$  are of the form:

$$p_t(x,y) = t^{-\frac{\alpha}{2}(1+\beta)} \phi\left(\frac{xy}{t^{\alpha}}\right) y^{\beta} \exp\left\{-\frac{k^2}{2}\left(\frac{x^{\frac{2}{\alpha}}}{t} + \frac{y^{\frac{2}{\alpha}}}{t}\right)\right\}$$
(4.3)

for k > 0, or if it is in h-transform relationship with it. Moreover, the semigroup densities are related as follows:

$$q_t^{(x)}(a,b) = \frac{\phi(xb)}{\phi(xa)} \exp\left(-t \frac{k^2}{2} x^{\frac{2}{\alpha}}\right) p_t(a,b).$$
 (4.4)

*Proof.* Theorem 2.4 formulated on  $\mathbb{R}_+$  implies  $\Phi(x,y) = \phi(xy)$ ,  $\theta(y) = y^{\beta}$  and  $\rho(x) = -\frac{k^2}{2} x^{\frac{2}{\alpha}}$ , k > 0, for the condition 1-3 to be satisfied in one dimension.

We identify further the class of diffusion processes and provide a different proof for the result in [24].

**Proposition 4.4.** If  $(X_t, t \ge 0)$  is a diffusion process and  $(tX_{\frac{1}{t}}, t > 0)$  is homogeneous and conservative, then both are necessarily (possibly time-scaled) Bessel processes in the wide sense.

*Proof.* If  $(X_t, t \ge 0)$  is a diffusion process, then its infinitesimal generator has the following general structure:

$$\mathcal{L} = s(y)\frac{\partial^2}{\partial y^2} + \mu(y)\frac{\partial}{\partial y}$$

where it remains to identify the functions s(y) and  $\mu(y)$ . For a fixed x > 0, let  $\mathcal{L}^{(x)}$  be the infinitesimal generator of  $(tX_{\frac{1}{t}}, t > 0)$ . From equation (2.5),  $\mathcal{L}^{(x)}$  has the following relationship with  $\mathcal{L}$ :

$$\mathcal{L}^{(x)}: f(b) \mapsto \frac{1}{\phi(xb)} \mathcal{L}(\phi(xb)f(b)) - \frac{k^2}{2} x^2 f(b),$$

which develops to

$$\mathcal{L}^{(x)}f(b) = s(b)f''(b) + \left\{\mu(b) + 2x \ s(b) \ \frac{\phi'(xb)}{\phi(xb)}\right\}f'(b) + U(x,b)f(b).$$

For the process to be conservative, we require U(x,b) = 0, which implies no killing in the interior of the domain, that is

$$s(b) x^{2} \frac{\phi''(xb)}{\phi(xb)} + \mu(b) x \frac{\phi'(xb)}{\phi(xb)} - \frac{k^{2}}{2} x^{2} = 0.$$

We change variables to z = xb to obtain

$$s(z/x) \phi''(z) + \frac{\mu(z/x)}{x} \phi'(z) - \frac{k^2}{2} \phi(z) = 0.$$

Since the latter must be valid for all x > 0, we are led to set:  $s(b) = \frac{\sigma^2}{2}$  for  $\sigma > 0$  and  $\mu(b) = \frac{\sigma^2}{2}$  for  $\sigma > 0$  and  $\mu(b) = \frac{\sigma^2}{2}$  for  $\sigma > 0$ . This yields the following equation

$$\frac{1}{2}\phi''(z) + \frac{2\nu + 1}{2z} \phi'(z) - \frac{k^2}{2\sigma^2} \phi(z) = 0.$$

The general solution (non-singular at 0 and up to a constant factor) is expressed through the modified Bessel function of the first kind as follows:

$$\phi(z) = \left(\frac{kz}{\sigma}\right)^{-\nu} I_{\nu}\left(\frac{kz}{\sigma}\right).$$

Gathering the different factors in (4.3) leads to

$$p_t(x,y) = N t^{-\frac{1+\beta}{2}} \left(\frac{xy}{t}\right)^{-\nu} I_{\nu} \left(\frac{k}{\sigma} \frac{xy}{t}\right) y^{\beta} \exp\left\{-\frac{k^2}{2} \left(\frac{x^2}{t} + \frac{y^2}{t}\right)\right\},$$

where N is a normalization factor. The additional condition that  $\lim_{t\to 0} p_t(x,y) = \delta(x-y)$  implies  $\beta = 2\nu + 1$  and  $k = \frac{1}{\sigma}$ , which leads to the semigroup densities of a time-scaled Bessel process of dimension  $\nu$ :

$$p_t(x,y) = \frac{y}{\sigma^2 t} \left(\frac{y}{x}\right)^{\nu} I_{\nu} \left(\frac{xy}{\sigma^2 t}\right) \exp\left\{-\frac{x^2 + y^2}{2\sigma^2 t}\right\}.$$

The infinitesimal generator for  $(tX_{\frac{1}{2}}, t > 0)$  is given by

$$\mathcal{L}^{(x)} = \frac{\sigma^2}{2} \frac{\partial^2}{\partial b^2} + \sigma^2 \left\{ \frac{2\nu + 1}{2b} + x \frac{\phi'(xb)}{\phi(xb)} \right\} \frac{\partial}{\partial b},$$

where one recognizes expression (4.1) for  $h_c(b) = \phi(xb)$  and a time-scale  $t \mapsto \sigma^2 t$ .

Proposition 4.4 smoothly extends to any power  $\alpha > 0$  of the Bessel process through the mapping  $X_t \mapsto X_t^{\alpha}$ . In particular, it is worth remarking that the case  $\alpha = 2$  gives rise to squares of Bessel processes, which leads to the following result:

**Proposition 4.5.** If  $(X_t, t \ge 0)$  is a diffusion process and  $(t^2X_{\frac{1}{t}}, t > 0)$  is homogeneous and conservative, then both are necessarily (possibly time-scaled) squares of Bessel processes in the wide sense.

# 5 Examples

## 5.1 Generalized Dunkl processes and Jacobi-Dunkl processes

#### 5.1.1 Multidimensional Dunkl processes

We briefly review the construction of the Dunkl process in  $\mathbb{R}^n$  (see [22, 23]).

**Definition 5.1.** The Dunkl process in  $\mathbb{R}^n$  is the Markov càdlàg process with infinitesimal generator

$$\frac{1}{2}\mathcal{L}^{(k)} = \frac{1}{2}\sum_{i=1}^{n} T_i^2 \tag{5.1}$$

where  $T_i$ ,  $1 \le i \le n$ , is a one-dimensional differential-difference operator defined for  $u \in C^1(\mathbb{R}^n)$  by

$$T_i u(x) = \frac{\partial u(x)}{\partial x_i} + \sum_{\alpha \in R_+} k(\alpha) \alpha_i \frac{u(x) - u(\sigma_\alpha x)}{\langle \alpha, x \rangle}.$$
 (5.2)

 $\langle \cdot, \cdot \rangle$  is the usual scalar product. R is a root system in  $\mathbb{R}^n$  and  $R_+$  a positive subsystem.  $k(\alpha)$  is a non-negative multiplicity function defined on R and invariant by the finite reflection group W associated with R.  $\sigma_{\alpha}$  is the reflection operator with respect to the hyperplane  $H_{\alpha}$  orthogonal to  $\alpha$  such that  $\sigma_{\alpha}x = x - \langle \alpha, x \rangle \alpha$  and for convenience  $\langle \alpha, \alpha \rangle = 2$  (see [8, 9]).

A result obtained by M. Rösler [22] yields the semigroup densities as follows:

$$p_t^{(k)}(x,y) = \frac{1}{c_k t^{\gamma + n/2}} \exp\left(-\frac{|x|^2 + |y|^2}{2t}\right) D_k\left(\frac{x}{\sqrt{t}}, \frac{y}{\sqrt{t}}\right) \omega_k(y)$$
 (5.3)

where  $D_k(x,y) > 0$  is the Dunkl kernel,  $\omega_k(y) = \prod_{\alpha \in R_+} |\langle \alpha, y \rangle|^{2k(\alpha)}$  the weight function which is

homogeneous of degree 
$$2\gamma = 2\sum_{\alpha \in R_+} k(\alpha)$$
 and  $c_k = \int_{\mathbb{R}^n} e^{-\frac{|x|^2}{2}} \omega_k(x) dx$ .

Following a thorough study of the properties of the one-dimensional Dunkl process in [13], it was remarked in [12] that the Dunkl process in  $\mathbb{R}^n$  enjoys the time-inversion property of degree 1. Considering that the Dunkl kernel satisfies

$$D_k(x,y) = D_k(y,x)$$
 and  $D_k(\mu x, y) = D_k(x, \mu y),$  (5.4)

the proof is straightforward with

$$\Phi(x,y) \equiv D_k(x,y), \quad \theta(y) \equiv \omega_k(y), \quad \rho(x) \equiv -\frac{|x|^2}{2}.$$
 (5.5)

By equation (2.5), the semigroup densities of the time-inverted process is even in h-transform relationship with the semigroup densities of the original Dunkl process:

$$q_t^{(x)}(a,b) = \frac{D_k(x,b)}{D_k(x,a)} \exp\left(-\frac{|x|^2}{2}t\right) p_t^{(k)}(a,b).$$
 (5.6)

#### 5.1.2 Generalized Dunkl processes

In an attempt to generalize the Dunkl process, we extend the definition of the infinitesimal generator to

$$\mathcal{L}^{(k,\lambda)}f(x) = \frac{1}{2}\Delta f(x) + \sum_{\alpha \in R_+} k(\alpha) \frac{\langle \nabla f(x), \alpha \rangle}{\langle x, \alpha \rangle} + \sum_{\alpha \in R_+} \lambda(\alpha) \frac{f(\sigma_\alpha x) - f(x)}{\langle x, \alpha \rangle^2}$$
(5.7)

where  $\Delta$  is the usual Laplacian,  $f \in C^2(\mathbb{R}^n)$  and  $\lambda(\alpha)$  is a non-negative multiplicity function defined on R and invariant by the finite reflection group W, similarly as  $k(\alpha)$ . We retrieve the Dunkl process for  $\lambda(\alpha) = k(\alpha)$ . Note that these processes are no longer martingales for  $\lambda(\alpha) \neq k(\alpha)$ .

The one-dimensional case was introduced in [13], where the semigroup densities were explicitly derived,

$$p_t^{(k,\lambda)}(x,y) = \frac{1}{t^{k-\frac{1}{2}}} y^{k-\frac{1}{2}} \exp\left(-\frac{x^2+y^2}{2t}\right) D_{k,\lambda}\left(\frac{xy}{t}\right)$$
 (5.8)

with the generalized Dunkl kernel,  $(\nu=k-\frac{1}{2},\ \mu=\sqrt{\nu^2+4\lambda})$ 

$$D_{k,\lambda}(z) = 1_{\{y \in \mathbb{R}_{-}\}} \frac{1}{2z^{\nu}} (I_{\nu} - I_{\mu})(-z) + 1_{\{y \in \mathbb{R}_{+}\}} \frac{1}{2z^{\nu}} (I_{\nu} + I_{\mu})(z), \qquad (5.9)$$

for  $z = \frac{xy}{t}$ . From (5.4), the generalized Dunkl kernel satisfies

$$D_{k,\lambda}(x,y) = D_{k,\lambda}(y,x)$$
 and  $D_{k,\lambda}(\mu x,y) = D_{k,\lambda}(x,\mu y),$  (5.10)

which readily implies that the generalized Dunkl process also enjoys the time-inversion property of degree 1. The semigroup densities were derived as an application of the skew-product representation of the generalized Dunkl process  $(X_t, t \ge 0)$  in terms of its absolute value (a Bessel process) and an independent Poisson process  $N_t^{(\lambda)}$ :

$$X_{t} \stackrel{(d)}{=} |X_{t}| (-1)^{N_{A_{t}}^{(\lambda)}} \tag{5.11}$$

where  $A_t = \int_0^t \frac{ds}{X_s^2}$ .

In the n-dimensional case, the application of the skew-product representation derived by Chybiryakov [6] shows that the generalized Dunkl process enjoys time-inversion for some specific root systems. We first recall one of the main results of [6].

**Proposition 5.2.** Let  $(X_t, t \ge 0)$  be the generalized Dunkl process generated by (5.7), with  $(X_t^W, t \ge 0)$  its radial part, i.e. the process confined to a Weyl chamber. Let  $R_+ \equiv \{\alpha_1, \ldots, \alpha_l\}$  for some  $l \in \mathbb{N}$  be the corresponding positive root system and let  $(N_t^i, t \ge 0)$ ,  $i = 1, \ldots, l$  be independent Poisson processes of respective intensities  $\lambda(\alpha_i)$ . Then  $X_t$  may be represented as  $Y_t^l$ , which is defined by induction as follows:

$$Y_t^0 = X_t^W$$
 and  $Y_t^i = \sigma_{\alpha_i}^{N_{A_t^i}^i} Y_t^{i-1}, i = 1, \dots, l,$ 

where 
$$A_t^i = \int_0^t \frac{ds}{\langle Y_s^{i-1}, \alpha_i \rangle^2}$$
.

The proof follows the argument in [6], while replacing  $k(\alpha_i)$  by  $\lambda(\alpha_i)$  appropriately.

From now on, let  $R_+ \equiv \{\alpha_1, \ldots, \alpha_l\}$  for some  $l \leq n$  be an orthogonal positive root system, that is  $\langle \alpha_i, \alpha_j \rangle = 2\delta_{ij}$ . For this particular root system, one can show that the generalized Dunkl process enjoys time-inversion of degree 1. We first prove the following absolute continuity relation:

**Lemma 5.3.** Let  $(X_t^W, t \ge 0)$  be the radial Dunkl process with infinitesimal generator

$$\mathcal{L}_{k}^{W} f(x) = \frac{1}{2} \Delta f(x) + \sum_{i=1}^{l} k(\alpha_{i}) \frac{\langle \alpha_{i}, \nabla f(x) \rangle}{\langle \alpha_{i}, x \rangle}.$$
 (5.12)

Fixed  $\nu \in \{1, \ldots, l\}$ . Let  $k'(\alpha)$  be another coefficient function on the root system  $R_+$  such that  $k'(\alpha_{\nu}) > k(\alpha_{\nu})$  and  $k'(\alpha_i) = k(\alpha_i)$  for  $i \neq \nu$ . Then, denoting  $P_x^{(k)}$  the law of the radial Dunkl process  $X_t^W$  starting from x, we have

$$P_x^{(k')}\big|_{\mathcal{F}_t} = \left(\frac{\langle \alpha_{\nu}, X_t^W \rangle}{\langle \alpha_{\nu}, x \rangle}\right)^{k'(\alpha_{\nu}) - k(\alpha_{\nu})} \exp\left[-\frac{\left(k'(\alpha_{\nu}) - \frac{1}{2}\right)^2 - \left(k(\alpha_{\nu}) - \frac{1}{2}\right)^2}{2} \int_0^t \frac{ds}{\langle \alpha_{\nu}, X_s^W \rangle^2}\right] \cdot P_x^{(k)}\big|_{\mathcal{F}_t}.$$
(5.13)

*Proof.* Let  $k_0(\alpha)$  be a coefficient such that  $k_0(\alpha_{\nu}) = \frac{1}{2}$  for some fixed  $i \in \{1, ..., l\}$ .  $X_t^W$  has the following martingale decomposition (see [11]):

$$X_t^W = x + B_t^{(k_0)} + \sum_{i=1}^l k_0(\alpha_i) \int_0^t \frac{ds}{\langle \alpha_i, X_s^W \rangle} \alpha_i$$

where  $B_t^{(k_0)}$  is a  $(P_x^{(k_0)}, \mathcal{F}_t)$ -Brownian motion. Consider the local martingale

$$L_t^{(k')} = \exp\left(\left(k'(\alpha_\nu) - \frac{1}{2}\right) \int_0^t \frac{\langle \alpha_\nu, dB_s^{(k_0)} \rangle}{\langle \alpha_\nu, X_s^W \rangle} - \frac{\left(k'(\alpha_\nu) - \frac{1}{2}\right)^2}{2} \int_0^t \frac{ds}{\langle \alpha_\nu, X_s^W \rangle^2}\right),$$

for some coefficient function  $k'(\alpha)$  such that  $k'(\alpha_{\nu}) > \frac{1}{2}$  and  $k'(\alpha_{i}) = k_{0}(\alpha_{i})$  for  $i \neq \nu$ . The Itô formula for  $\ln\left(\langle \alpha_{\nu}, X_{t}^{W} \rangle\right)$  combined with the orthogonality of the roots yields

$$L_t^{(k')} = \left(\frac{\langle \alpha_{\nu}, X_t^W \rangle}{\langle \alpha_{\nu}, x \rangle}\right)^{k'(\alpha_{\nu}) - \frac{1}{2}} \exp\left(-\frac{\left(k'(\alpha_{\nu}) - \frac{1}{2}\right)^2}{2} \int_0^t \frac{ds}{\langle \alpha_{\nu}, X_s^W \rangle^2}\right).$$

Define the new law  $P_x^{(k')}|_{\mathcal{F}_t} = L_t^{(k')} \cdot P_x^{(k_0)}|_{\mathcal{F}_t}$ . By the Girsanov theorem,

$$B_t^{(k')} = B_t^{(k_0)} - \left(k'(\alpha_\nu) - \frac{1}{2}\right) \int_0^t \frac{ds}{\langle \alpha_\nu, X_s^W \rangle}$$

is a  $(P_x^{(k')}, \mathcal{F}_t)$ -Brownian motion and hence,

$$X_t^W = x + B_t^{(k')} + \sum_{i=1}^l k'(\alpha_i) \int_0^t \frac{ds}{\langle \alpha_i, X_s^W \rangle} \alpha_i$$

is a radial Dunkl process under  $(P_x^{(k')}, \mathcal{F}_t)$ .

Define k as another coefficient on the root system that satisfies the conditions enunciated in the lemma. The absolute continuity relation is then a consequence of the successive application of the latter result to the indices k' and k.

Now as an application of Proposition 5.2, we prove the following:

**Proposition 5.4.** Let  $R_+ \equiv \{\alpha_1, \ldots, \alpha_l\}$  be an orthogonal positive root system for  $l \leq n$ . Then the generalized Dunkl process  $(X_t, t \geq 0)$  generated by (5.7) enjoys the time-inversion property of degree 1.

*Proof.* Using orthogonality of the roots, remark that

$$\langle \alpha_i, \sigma_j x \rangle^2 = \langle \alpha_i, x - \langle \alpha_j, x \rangle \alpha_j \rangle^2 = \langle \alpha_i, x \rangle^2,$$

which implies in particular

$$\langle \alpha_i, Y_t^i \rangle^2 = \langle \alpha_i, X_t^W \rangle^2$$

so that the inductive representation of  $X_t$  in Proposition 5.2 becomes

$$X_t = \prod_{i=1}^l \sigma_{\alpha_i}^{N_{A_t^i}^i} X_t^W \quad \text{for} \quad A_t^i = \int_0^t \frac{ds}{\langle X_s^W, \alpha_i \rangle^2}.$$

The radial part of a Dunkl process enjoys the time-inversion property of degree 1. We need to show that the semigroup densities of the generalized Dunkl process are related to the semigroup densities of its radial parts. For  $f \in C^2(\mathbb{R}^n)$ ,

$$E_x[f(Y_t^{i})] = E_x[f(Y_t^{i-1})1_{\{N_{A_t^i}^{i} \text{ is even}\}}] + E_x[f(\sigma_{\alpha_i}Y_t^{i-1})1_{\{N_{A_t^i}^{i} \text{ is odd}\}}].$$

Since  $P(N_u^i \text{ is even}) = \frac{1}{2} (1 + \exp(-2\lambda(\alpha_i)u))$ , we obtain

$$\mathbf{E}_x \big[ f(Y_t^i) \big] = \mathbf{E}_x \left[ f(Y_t^{i-1}) \ \frac{1}{2} \left( 1 + \exp(-2\lambda(\alpha_i) A_t^i) \right) \right] + \mathbf{E}_x \left[ f(\sigma_{\alpha_i} Y_t^{i-1}) \ \frac{1}{2} \left( 1 - \exp(-2\lambda(\alpha_i) A_t^i) \right) \right].$$

The expectation  $E_x[f(X_t)]$  can thus be evaluated by induction on  $i \in \{1, ..., l\}$ . It follows that the semigroup densities of  $X_t$  can be expressed as the product of the semigroup densities of its radial parts times a function involving expectations of the form

$$\mathbf{E}_{x}^{(k)} \left[ \exp\left(-2\lambda(\alpha_{\nu})A_{t}^{\nu}\right) \left| X_{t}^{W} = y \right| \right],$$

for  $\nu \in \{1, \ldots, l\}$ . From Lemma 5.3,

$$\mathbf{E}_{x}^{(k)} \left[ \exp\left(-2\lambda(\alpha_{\nu})A_{t}^{\nu}\right) \middle| X_{t}^{W} = y \right] = \frac{p_{t}^{(k')}(x,y)}{p_{t}^{(k)}(x,y)} \left( \frac{\langle \alpha_{\nu}, y \rangle}{\langle \alpha_{\nu}, x \rangle} \right)^{k(\alpha_{\nu}) - k'(\alpha_{\nu})},$$

where  $k'(\alpha_i) = k(\alpha_i)$  for  $i \neq \nu$  and  $k'(\alpha_{\nu}) = \frac{1}{2} + \sqrt{\left(k(\alpha_{\nu}) - \frac{1}{2}\right)^2 + 4\lambda(\alpha_{\nu})}$ . The form of the semigroup densities in (5.3) implies that the expectation is a ratio of Dunkl kernels,

$$\mathbf{E}_{x}^{(k)}\left[\exp\left(-2\lambda(\alpha_{\nu})A_{t}^{\nu}\right)\left|X_{t}^{W}=y\right]=\frac{c_{k}}{c_{k'}}\frac{D_{k'}\left(\frac{x}{\sqrt{t}},\frac{y}{\sqrt{t}}\right)}{D_{k}\left(\frac{x}{\sqrt{t}},\frac{y}{\sqrt{t}}\right)}\frac{w_{k'}\left(\frac{y}{\sqrt{t}}\right)}{w_{k}\left(\frac{y}{\sqrt{t}}\right)}\left(\frac{\langle\alpha_{\nu},\frac{y}{\sqrt{t}}\rangle}{\langle\alpha_{\nu},\frac{x}{\sqrt{t}}\rangle}\right)^{k(\alpha_{\nu})-k'(\alpha_{\nu})}$$

which reduces to

$$\mathbf{E}_{x}^{(k)} \left[ \exp\left(-2\lambda(\alpha_{\nu})A_{t}^{\nu}\right) \left| X_{t}^{W} = y \right] = \frac{c_{k}}{c_{k'}} \frac{D_{k'}\left(\frac{x}{\sqrt{t}}, \frac{y}{\sqrt{t}}\right)}{D_{k}\left(\frac{x}{\sqrt{t}}, \frac{y}{\sqrt{t}}\right)} \left( \langle \alpha_{\nu}, \frac{y}{\sqrt{t}} \rangle \langle \alpha_{\nu}, \frac{x}{\sqrt{t}} \rangle \right)^{k'(\alpha_{\nu}) - k(\alpha_{\nu})},$$

by definition of  $k'(\alpha)$ . The conditional expectation thus satisfies condition 1 of Theorem 2.4. As a consequence, the conditions of Theorem 2.4 are satisfied for

$$\Phi(x,y) \equiv D_{k,\lambda}(x,y), \quad \theta(y) \equiv \omega_k(y), \quad \rho(x) \equiv -\frac{|x|^2}{2}$$

where  $D_{k,\lambda}(x,y)$  is a generalized Dunkl kernel given explicitly in terms of radial Dunkl kernels and satisfying equivalent conditions (see (5.4)).

#### 5.1.3 Jacobi-Dunkl processes

Gallardo et al. [5] derived the Jacobi-Dunkl process as the hyperbolic analog of the one-dimensional Dunkl process. It is defined as the process generated by

$$\mathcal{L}^{(\alpha,\beta)}f(x) = \frac{\partial^2 f(x)}{\partial x^2} + \frac{A'(x)}{A(x)} \frac{\partial f(x)}{\partial x} + \frac{\partial}{\partial x} \left(\frac{A'(x)}{A(x)}\right) \left(\frac{f(x) - f(-x)}{2}\right),\tag{5.14}$$

where  $A(x) = \left(\sinh^2(x)\right)^{\alpha + \frac{1}{2}} \left(\cosh^2(x)\right)^{\beta + \frac{1}{2}}$ . From the expression of the semigroup densities developed in [5], this process does not enjoy the time-inversion property. Its radial part however corresponds to the Jacobi process of index  $(\alpha, \beta)$  on  $\mathbb{R}_+$  (see [15, 16]). The infinitesimal generator of the Jacobi process, expressed by

$$\mathcal{L}^{(\alpha,\beta)}f(x) = \frac{1}{2}\frac{\partial^2 f(x)}{\partial x^2} + \frac{A'(x)}{A(x)}\frac{\partial f(x)}{\partial x},\tag{5.15}$$

is in h-transform relationship with the Laplacian operator for  $h(x) = \sqrt{A(x)}$ . Since the Brownian motion enjoys the time-inversion property of degree 1, so does the Jacobi process by Theorem 2.4.

#### 5.2 Matrix-valued processes

#### 5.2.1 Eigenvalue processes

Dyson in [10] described the eigenvalues of a Hermitian Brownian motion as the joint evolution of independent Brownian motions conditioned never to collide (see also [14] and [4]). It was further remarked that the process version of the Gaussian orthogonal ensemble does not admit such a representation for its eigenvalues. This work was extended by König and O'Connell in [17] to the process version of the Laguerre ensemble, denominated the Laguerre process and defined as follows:

**Definition 5.5.** Let  $B_t$  be an  $n \times m$  matrix with independent standard complex Brownian entries. The Laguerre process is the matrix-valued process defined by  $\{X_t = B'_t B_t, t \geq 0\}$ , where  $B'_t$  is the transpose of  $B_t$ .

From [17], the eigenvalues of the Laguerre process evolve like m independent squared Bessel processes conditioned never to collide. No such representation however exists for the case where the entries of  $B_t$  are real Brownian motions, i.e. the Wishart process considered by Bru in [3].

The main result of [17] is that both of the above mentioned eigenvalues processes (in the complex Brownian case) can be obtained as the h-transform of processes with m independent components. The joint eigenvalues process is thus in h-transform relationship with a process that enjoys the time inversion property of degree 1 in the case of the m-dimensional Brownian motion and degree 2 in the case of the m-dimensional squared Bessel process, as made explicit in the following proposition:

**Proposition 5.6.** Let  $p_t(x_i, y_i)$  (i = 1, ..., m) be the semigroup densities of squared Bessel processes (respectively Brownian motions), and let

$$h(x) = \prod_{i < j}^{m} (x_j - x_i) \tag{5.16}$$

for  $x = (x_1, \ldots, x_m)$ . Then, the semigroup densities of the joint eigenvalues process of the Laguerre process (respectively the Hermitian Brownian motion) are given by

$$\tilde{p}_t(x,y) = \frac{h(y)}{h(x)} \prod_{i=1}^m p_t(x_i, y_i)$$
(5.17)

with respect to the Lebesgue measure  $dy = \prod_{i=1}^{m} dy_i$ .

It follows immediately by Theorem 2.4 that the eigenvalues processes enjoy the time-inversion property. Moreover by Proposition 2.2, they yield the same process under time-inversion as the m-dimensional Brownian motion or the m-dimensional squared Bessel process respectively.

#### 5.2.2 Wishart processes

The Wishart process WIS $(\delta, tI_m, \frac{1}{t}x)$ , introduced by Bru in [2], is a continuous Markov process taking values in the space of real symmetric positive definite  $m \times m$  matrices  $S_m^+$ . It is solution to the following stochastic differential equation:

$$dX_t = \sqrt{X_t} dB_t + dB_t' \sqrt{X_t} + \delta I_m dt, \quad X_0 = x, \tag{5.18}$$

where  $B_t$  is an  $m \times m$  matrix with Brownian entries and  $I_m$  the identity matrix. Further results have been obtained in [1] and [3]. In [7], among other major findings about the Wishart process, the transition probability densities expressed with respect to the Lebesgue measure  $dy = \prod_{i \leq j} (dy_{ij})$ 

were derived in terms of generalized Bessel functions (we refer to the appendix for the definition):

$$p_t(x,y) = \frac{1}{(2t)^{\frac{m(m+1)}{2}}} \exp\left(-\frac{1}{2t}Tr(x+y)\right) \left(\frac{\det(y)}{\det(x)}\right)^{\frac{\delta-m-1}{4}} \tilde{\mathbf{I}}_{\frac{\delta-m-1}{2}}\left(\frac{xy}{4t^2}\right), \tag{5.19}$$

for  $x, y \in S_m^+$  and  $\delta > m-1$ . From the shape of its densities, the Wishart process was stated in [12] as an example of Markov processes enjoying the time-inversion property of degree 2. The hypothesis of Theorem 2.4 is indeed satisfied for  $n = \frac{1}{2}m(m+1)$  and

$$\Phi(x,y) \equiv \left(\det(x)\det(y)\right)^{-\frac{\delta-m-1}{4}} \tilde{\mathbf{I}}_{\frac{\delta-m-1}{2}} \left(\frac{xy}{4}\right) \quad , \quad \theta(y) \equiv \frac{1}{2^n} (\det(y))^{\frac{\delta-m-1}{2}} \quad , \quad \rho(x) \equiv -\frac{1}{2} Tr(x). \tag{5.20}$$

Next we use a skew-product representation, as for the Dunkl process, to elaborate on the Wishart process and derive a matrix-valued process with jumps. The skew-product representation allows the expression of the semigroup densities in terms of the Wishart transition probability densities.

**Definition 5.7.** Let  $(N_t^{(\lambda)}, t \geq 0)$  be a Poisson process with intensity  $\lambda$ . Let  $(X_t, t \geq 0)$  be a Wishart process WIS $(\delta, tI_m, \frac{1}{t}x)$  independent of the Poisson process. The skew-Wishart process  $(X_t^{(\lambda)}, t \geq 0)$  is defined through the skew-product

$$X_t^{(\lambda)} = X_t \ (-1)^{N_{A_t}^{(\lambda)}} \tag{5.21}$$

where 
$$A_t = \int_0^t Tr(X_s^{-1})ds$$
.

**Proposition 5.8.** The transition probability densities of the skew-Wishart process are related to the semigroup densities  $p_t(x, y)$  of the Wishart process  $X_t$  as follows

$$p_{t}^{(\lambda)}(x,y) = p_{t}(x,|y|) \left\{ 1_{\{y \in S_{m}^{+}\}} \frac{1}{2} \left( 1 + \left( \frac{\tilde{\mathbf{I}}_{\nu'}}{\tilde{\mathbf{I}}_{\nu}} \right) \left( \frac{xy}{4t^{2}} \right) \right) + 1_{\{y \in S_{m}^{-}\}} \frac{1}{2} \left( 1 - \left( \frac{\tilde{\mathbf{I}}_{\nu'}}{\tilde{\mathbf{I}}_{\nu}} \right) \left( \frac{-xy}{4t^{2}} \right) \right) \right\}.$$

$$(5.22)$$

$$for \ \nu = \frac{\delta - m - 1}{2}, \ \nu' = \sqrt{\nu^{2} + 4\lambda} \ and \ |y| = y(1_{\{y \in S_{m}^{+}\}} - 1_{\{y \in S_{m}^{-}\}}).$$

*Proof.* Let  $(\mathbf{P}_t)_{t>0}$  be the semigroup of the Dunkl-Wishart process. For x>0 and  $f\in C_c(M_m(\mathbb{R}))$ ,

$$\mathbf{P}_t f(x) = \mathbf{E}_x \left[ f(X_t^{(\lambda)}) \right] = \mathbf{E}_x \left[ f(X_t) \ \mathbf{1}_{\{N_{A_t}^{(\lambda)} \text{ is even}\}} \right] + \mathbf{E}_x \left[ f(-X_t) \ \mathbf{1}_{\{N_{A_t}^{(\lambda)} \text{ is odd}\}} \right].$$

With  $P(N_u^{(\lambda)} \text{ is even}) = \frac{1}{2}(1 + \exp(-2\lambda u))$ , we have

$$\mathbf{P}_{t}f(x) = \mathbf{E}_{x} \left[ f(X_{t}) \ \frac{1}{2} (1 + \exp(-2\lambda A_{t})) \right] + \mathbf{E}_{x} \left[ f(-X_{t}) \ \frac{1}{2} (1 - \exp(-2\lambda A_{t})) \right]. \tag{5.23}$$

Let  $\mathbf{Q}_{x}^{(\nu')}$  with  $\nu' = \frac{\delta' - m - 1}{2}$  denote the probability law of a Wishart process WIS $(\delta', tI, \frac{1}{t}x)$ , and  $\mathbf{Q}_{x}^{(\nu)}$  with  $\nu = \frac{\delta - m - 1}{2}$  the probability law of  $X_{t}$ . According to Theorem 1.2 (Remark 2.3) in [7], the probability laws are related as follows:

$$\mathbf{Q}_x^{(\nu')}\big|_{\mathcal{F}_t} = \left(\frac{\det X_t}{\det x}\right)^{\frac{\nu'-\nu}{2}} \exp\left(-\frac{\nu'^2 - \nu^2}{2} \int_0^t Tr(X_s^{-1}) ds\right) \cdot \mathbf{Q}_x^{(\nu)}\big|_{\mathcal{F}_t},$$

from which we deduce

$$\frac{\mathbf{p}_t^{(\nu')}(x,y)}{\mathbf{p}_t^{(\nu)}(x,y)} = \left(\frac{\det y}{\det x}\right)^{\frac{\nu'-\nu}{2}} \mathbf{Q}_x^{(\nu)} \left[\exp\left(-\frac{\nu'^2-\nu^2}{2}\int_0^t Tr(X_s^{-1})ds\right) \left|X_t=y\right|\right.$$

Thus, from the expression of the semigroup densities in (5.19), we have

$$\mathbf{E}_{x}^{(\nu)} \left[ \exp\left(-2\lambda \int_{0}^{t} Tr(X_{s}^{-1}) ds \right) \left| X_{t} = y \right] = \left(\frac{\tilde{\mathbf{I}}_{\sqrt{\nu^{2}+4\lambda}}}{\tilde{\mathbf{I}}_{\nu}}\right) \left(\frac{xy}{4t^{2}}\right).$$

Combining the latter with (5.23) yields the semigroup densities for the skew-Wishart process.

The skew-Wishart is an example of matrix-valued process with jumps that enjoys the time-inversion property of degree 2. Indeed, by setting

$$\Phi(x,y) \equiv \left(\det(x)\det(|y|)\right)^{-\frac{\nu}{2}} \left\{ 1_{\{y \in S_m^+\}} \frac{1}{2} \left( \tilde{\mathbf{I}}_{\nu} + \tilde{\mathbf{I}}_{\nu'} \right) \left( \frac{xy}{4} \right) + 1_{\{y \in S_m^-\}} \frac{1}{2} \left( \tilde{\mathbf{I}}_{\nu} - \tilde{\mathbf{I}}_{\nu'} \right) \left( \frac{-xy}{4} \right) \right\},$$

$$\theta(y) \equiv \frac{1}{2^n} \left(\det(|y|)\right)^{\nu} \quad , \quad \rho(x) \equiv -\frac{1}{2} Tr(|x|), \tag{5.24}$$

the conditions of Theorem 2.4 are satisfied for  $\alpha = 2$ .

# A Generalized hypergeometric functions

Using the notation in Muirhead [20], hypergeometric functions of matrix arguments are defined for a real symmetric  $m \times m$  matrix X,  $a_i \in \mathbb{C}$  and  $b_j \in \mathbb{C} \setminus \{0, \frac{1}{2}, 1, \dots, \frac{m-1}{2}\}$  by

$${}_{p}\mathbf{F}_{q}(a_{1},\ldots,a_{p};b_{1},\ldots,b_{q};X) = \sum_{k=0}^{\infty} \sum_{\kappa} \frac{(a_{1})_{\kappa}\cdots(a_{p})_{\kappa}}{(b_{1})_{\kappa}\cdots(b_{q})_{\kappa}} \frac{C_{\kappa}(X)}{k!}$$
(A.1)

where the second summation is over all partitions  $\kappa = (k_1, \dots, k_m), k_1 \ge \dots \ge k_m \ge 0$ , of  $k = \sum_{i=1}^m k_i, k! = k_1! \dots k_m!$  and the generalized Pochhammer symbols are given by

$$(a)_{\kappa} = \prod_{i=1}^{m} \left( a - \frac{i-1}{2} \right)_{k_i}, \quad (a)_k = a(a+1) \cdots (a+k-1), \quad (a)_0 = 1.$$

 $C_{\kappa}(X)$  is the zonal polynomial corresponding to  $\kappa$ , which is a symmetric, homogeneous polynomial of degree k in the eigenvalues of X that satisfies

$$C_{\kappa}(YX) = C_{\kappa}(\sqrt{Y}X\sqrt{Y}) \tag{A.2}$$

for some  $Y \in S_m^+$ . The function  ${}_p\mathbf{F}_q(a_1,\ldots,a_p;b_1,\ldots,b_q;YX)$  thus makes sense. Finally, we define the generalized modified Bessel function by

$$\tilde{\mathbf{I}}_{\nu}(X) = \frac{\left(\det(X)\right)^{\frac{\nu}{2}}}{\Gamma_{m}\left(\nu + \frac{m+1}{2}\right)} \, {}_{0}\mathbf{F}_{1}\left(\nu + \frac{m+1}{2}; X\right) \tag{A.3}$$

where the generalized gamma function is given as a product of the usual gamma functions,

$$\Gamma_{m}\left(\alpha\right) = \pi^{\frac{m(m-1)}{4}} \prod_{i=1}^{m} \Gamma\left(\alpha - \frac{i-1}{2}\right) \tag{A.4}$$

for  $\operatorname{Re}(\alpha) > \frac{m-1}{2}$ . Note that the generalized modified Bessel for m=1 relates to the usual modified Bessel function  $I_{\nu}(x)$  by  $\tilde{\mathbf{I}}_{\nu}(x) = I_{\nu}(2\sqrt{x})$  (see [19]).

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