## Li-Yau Inequalities for Dunkl Heat Equations

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(joint work with Qian, Bin)

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## **3** Sharp Li–Yau inequalities for Dunkl heat kernel: $\mathbb{Z}_2^d$ case

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# Li-Yau inequalities

 $(\mathbb{M}^n, \rho, |\cdot|, \Delta, \nabla)$ 

Li–Yau [Acta Math. 1986]: Assume Ric  $\geq 0$ . Then for every positive solution to the heat equation  $\partial_t u = \Delta u$  on  $(0, \infty) \times \mathbb{M}$ ,

$$-\Delta(\log u(t,\cdot))(x) \le \frac{n}{2t}, \quad t>0, x\in\mathbb{M},$$

and equivalently,

$$\frac{|\nabla u(t,\cdot)(x)|^2}{u(t,x)^2} - \frac{\partial_t u(t,x)}{u(t,x)} \le \frac{n}{2t}, \quad t > 0, x \in \mathbb{M},$$

which implies the Harnack inequality

$$u(s,x) \le u(t,y) \left(\frac{t}{s}\right)^{n/2} \exp\left(\frac{\rho(x,y)^2}{4(t-s)}\right), \quad 0 < s < t < \infty, x, y \in \mathbb{M}.$$

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## **Basic notions**

Consider the Euclidean space  $\mathbb{R}^d$  with the standard scalar product  $\langle \cdot, \cdot \rangle$  and the induced norm  $|\cdot| = \sqrt{\langle \cdot, \cdot \rangle}$ .

• Reflection  $r_{\alpha}$ : for  $\alpha \in \mathbb{R}^d \setminus \{0\}$ ,

$$r_{\alpha}x = x - 2\frac{\langle \alpha, x \rangle}{|\alpha|^2}\alpha, \quad x \in \mathbb{R}^d,$$

which is a reflection in the hyperplane  $\alpha^{\perp}$ .

• Root system  $\mathfrak{R}$ : finite set in  $\mathbb{R}^d \setminus \{0\}$  such that  $\forall \alpha \in \mathfrak{R}$ ,

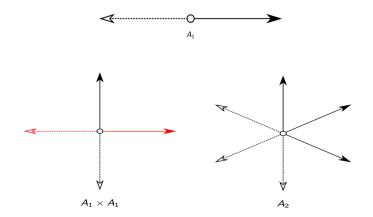
$$r_{\alpha}(\mathfrak{R}) = \mathfrak{R}$$
 and  $\mathfrak{R} \cap \alpha \mathbb{R} = \{\alpha, -\alpha\}.$ 

Normalize  $|\alpha| = \sqrt{2}, \alpha \in \mathfrak{R}.$ 

- Reflection group *G*: finite group generated by  $\{r_{\alpha} : \alpha \in \mathfrak{R}\}$ .
- Multiplicity function  $\kappa$ : *G*-invariant map  $\kappa$ . :  $\mathfrak{R} \to \mathbb{R}_+$ , i.e.,

$$\kappa_{g\alpha} = \kappa_{\alpha}, \quad g \in G, \ \alpha \in \mathfrak{R}.$$

# Examples of root systems



## Dunkl operator

Let  $\mathfrak{R}_+$  be the positive subsystem such that  $\mathfrak{R} = \mathfrak{R}_+ \uplus (-\mathfrak{R}_+)$ .

#### Definition (C.F. Dunkl: Trans. AMS 1989)

Given a root system  $\mathfrak{R}$  and a multiplicity function  $\kappa_{\cdot} : \mathfrak{R} \to \mathbb{R}_+$ , for every  $\xi \in \mathbb{R}^d$ , the Dunkl operator along  $\xi$  is defined by

$$\mathbf{D}_{\boldsymbol{\xi}}f(\boldsymbol{x}) = \partial_{\boldsymbol{\xi}}f(\boldsymbol{x}) + \sum_{\boldsymbol{\alpha}\in\mathfrak{R}_+}\kappa_{\boldsymbol{\alpha}}\langle\boldsymbol{\alpha},\boldsymbol{\xi}\rangle \frac{f(\boldsymbol{x}) - f(r_{\boldsymbol{\alpha}}\boldsymbol{x})}{\langle\boldsymbol{\alpha},\boldsymbol{x}\rangle}, \quad f\in C^1(\mathbb{R}^d), \, \boldsymbol{x}\in\mathbb{R}^d,$$

where  $\partial_{\xi}$  denotes the directional derivative along  $\xi$ .

Note that  $\mathrm{D}_{\xi}\circ\mathrm{D}_{\eta}=\mathrm{D}_{\eta}\circ\mathrm{D}_{\xi},$   $\xi,\eta\in\mathbb{R}^{d},$  and

$$\frac{f(x) - f(r_{\alpha}x)}{\langle \alpha, x \rangle} = -\frac{1}{\langle \alpha, x \rangle} \int_0^1 \frac{\partial}{\partial t} f(x - t \langle \alpha, x \rangle \alpha) \, \mathrm{d}t = \int_0^1 \partial_{\alpha} f(x - t \langle \alpha, x \rangle \alpha) \, \mathrm{d}t.$$

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# Dunkl gradient/Laplacian

Let  $\{e_j : j = 1, \dots, d\}$  be the standard orthonormal basis of  $\mathbb{R}^d$  and let  $D_j = D_{e_j}$ . The Dunkl gradient operator and the Dunkl Laplacian are respectively defined as

$$\nabla_{\kappa} = (\mathbf{D}_1, \cdots, \mathbf{D}_d), \quad \Delta_{\kappa} = \sum_{j=1}^d \mathbf{D}_j^2.$$

More precisely,  $\forall f \in C^2(\mathbb{R}^d)$ ,

$$\Delta_{\kappa}f(x) = \Delta f(x) + 2\sum_{\alpha \in \mathfrak{R}_{+}} \kappa_{\alpha} \Big(\frac{\langle \alpha, \nabla f(x) \rangle}{\langle \alpha, x \rangle} - \frac{f(x) - f(r_{\alpha}x)}{\langle \alpha, x \rangle^{2}}\Big), \quad x \in \mathbb{R}^{d}.$$

In particular, when  $\kappa = 0$ ,  $\nabla_0 = \nabla$  and  $\Delta_0 = \Delta$ .

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#### Example (rank-one case)

Let d = 1. Then  $\mathfrak{R} = \{-\sqrt{2}, \sqrt{2}\}, G = \{e, r\}$  with e(x) = x and r(x) = -x for every  $x \in \mathbb{R}$ . Given  $\lambda \in \mathbb{R}_+$ , the Dunkl operator is

$$\mathrm{D}f(x) = f'(x) + \lambda \frac{f(x) - f(-x)}{x}, \quad x \in \mathbb{R}, f \in C^1(\mathbb{R}),$$

and the Dunkl Laplacian is

$$D^{2}f(x) = f''(x) + \frac{\lambda}{x^{2}} [f(-x) - f(x) + 2xf'(x)], \quad x \in \mathbb{R}, f \in C^{2}(\mathbb{R}).$$

#### Example (radial Dunkl process)

Let  $W = \{x \in \mathbb{R}^d : \langle \alpha, x \rangle > 0, \ \alpha \in \mathfrak{R}_+\}$  and let  $\overline{W}$  be its closure. The *radial Dunkl process* is defined as the  $\overline{W}$ -valued Markov process with infinitesimal generator

$$\Delta_{\kappa}^{W} f(x) = \frac{1}{2} \Delta f(x) + \sum_{\alpha \in \mathfrak{R}_{+}} \kappa_{\alpha} \frac{\langle \alpha, \nabla f(x) \rangle}{\langle \alpha, x \rangle},$$

where  $f \in C^2(\overline{W})$  such that  $\langle \alpha, \nabla f(x) \rangle = 0$  whenever  $\langle \alpha, x \rangle = 0$ . It is known that the corresponding SDE, i.e.,

$$\mathrm{d}Y_t = \mathrm{d}B_t + \sum_{\alpha \in \mathfrak{R}_+} \kappa_\alpha \frac{\mathrm{d}t}{\langle \alpha, Y_t \rangle}, \quad Y_0 = y \in \overline{W},$$

has a unique strong solution for all  $t \ge 0$  (see e.g. O. Chybiryakov [SPA 2006], B. Schapira [PTRF 2007]).

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## A remark on Dunkl process

$$(\Delta_{\kappa}, \mathcal{D}(\Delta_{\kappa})) \iff \text{Dunkl process } (X_t)_{t \ge 0}$$

 $\forall I \subset \mathfrak{R}_+, \text{let}$ 

$$U_I = \{ \alpha \in \mathfrak{R}_+ : \langle \alpha, x \rangle = 0, \, x \in \cap_{\alpha \in I} \alpha^{\perp} \}.$$

 $(X_t)_{t\geq 0}$  is a càdlàg Markov process of jump type with jumping kernel

$$K(x,dy) = \begin{cases} \sum_{\alpha \in \mathfrak{R}_+} \frac{2\kappa_{\alpha}}{\langle \alpha, x \rangle^2} \delta_{r_{\alpha}x}(\mathrm{d}y), & x \in \mathbb{R}^d \setminus (\cup_{\alpha \in \mathfrak{R}_+} \alpha^{\perp}), \\ \sum_{\alpha \in \mathfrak{R}_+ \setminus U_I} \frac{2\kappa_{\alpha}}{\langle \alpha, x \rangle^2} \delta_{r_{\alpha}x}(\mathrm{d}y), & x \in \cap_{\alpha \in I} \alpha^{\perp}, \\ 0, & x = 0, \end{cases}$$

where *I* is any subset of  $\mathfrak{R}_+$ ,  $\delta$ . denotes the Dirac measure.

M. Rösler and M. Voit [Adv. App. Math. 1998], L. Gallardo and M. Yor [PTRF 2005] & [AOP 2006], B. Schapira [PTRF 2007], etc.

Let

$$\omega_{\kappa}(x) := \prod_{\alpha \in \mathfrak{R}_{+}} |\langle \alpha, x \rangle|^{\kappa_{\alpha}}, \quad x \in \mathbb{R}^{d},$$

which is a homogeneous function of degree

$$\lambda_{\kappa} := \sum_{\alpha \in \mathfrak{R}_+} \kappa_{\alpha}.$$

Let

$$\boldsymbol{\mu}_{\boldsymbol{\kappa}}(\mathrm{d} x) := \omega_{\boldsymbol{\kappa}}(x)\mathrm{d} x,$$

where dx denotes the Lebesgue measure on  $\mathbb{R}^d$ .

Let  $(P_t)_{t>0}$  be the Dunkl semigroup with  $P_t = e^{t\Delta_{\kappa}}$  for every t > 0. Then  $P_t$  admits the Dunkl heat kernel  $p_t(x, y)$  w.r.t.  $\mu_{\kappa}$ , which is a  $C^{\infty}$  function of all variables  $x, y \in \mathbb{R}^d$  and t > 0, and satisfies that

$$\partial_t p_t(x,y) = \Delta_{\kappa} p_t(\cdot,y)(x), \ p_t(x,y) = p_t(y,x) > 0, \ \int_{\mathbb{R}^d} p_t(x,y) \, \mathrm{d}\mu_{\kappa}(y) = 1;$$

moreover,

$$p_t(x,y) \leq rac{1}{c_\kappa (2t)^{d/2+\lambda_\kappa}} \exp\Big(-rac{\delta(x,y)^2}{4t}\Big), \quad x,y \in \mathbb{R}^d, \ t>0,$$

where  $c_{\kappa} := \int_{\mathbb{R}^d} e^{-|x|^2/2} \mu_{\kappa}(\mathrm{d}x)$  and  $\delta(x, y) := \min_{g \in G} |gx - y|$ .

2 Li–Yau inequalities for Dunkl heat equation

## **3** Sharp Li–Yau inequalities for Dunkl heat kernel: $\mathbb{Z}_2^d$ case

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# Li-Yau inequalities for Dunkl heat equation

#### Theorem (L.–Qian 2021)

Let  $T \in (0,\infty]$  and  $\beta : (0,T) \times \mathbb{R}^d \to \mathbb{R}$  be a function. Suppose that  $u : [0,T) \times \mathbb{R} \to (0,\infty)$  is any  $C^2$  solution to the Dunkl heat equation

 $\partial_t u(t,x) = \Delta_\kappa (u(t,\cdot))(x), \quad (t,x) \in (0,T) \times \mathbb{R}^d.$ 

Then

$$-\Delta_{\kappa} \big( \log p_t(\cdot, y) \big)(x) \le \beta(t, x), \quad (t, x, y) \in (0, T) \times \mathbb{R}^d \times \mathbb{R}^d, (1)$$

is equivalent to

$$-\Delta_{\kappa} \big( \log u(t, \cdot) \big)(x) \le \beta(t, x), \quad (t, x) \in (0, T) \times \mathbb{R}^d;$$
(2)

moreover, either (1) or (2) implies

$$\frac{|\nabla u(t,\cdot)(x)|^2}{u(t,x)^2} - \frac{\partial_t u(t,x)}{u(t,x)} \le \beta(t,x), \quad (t,x) \in (0,T) \times \mathbb{R}^d.$$

The idea of proof is motivated by C. Yu and F. Zhao [JGA, 2020], where sharp Li–Yau inequalities for the Laplace–Beltrami operator on hyperbolic spaces were obtained.

Recently, similar idea was employed by F. Weber and R. Zacher in [arXiv:2012.12974] to prove Li–Yau inequalities for the fractional Laplacian  $(-\Delta)^s$  with  $s \in (0, 1)$ .

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#### Theorem (L.–Qian 2021)

Suppose that G is isomorphic to  $\mathbb{Z}_2^d = \{0, 1\}^d$ . Then

$$-\Delta_{\kappa} (\log p_t(\cdot, y))(x) \le \frac{d+2\lambda_{\kappa}}{2t}, \quad x, y \in \mathbb{R}^d, t > 0.$$

If  $\kappa = 0$ , then  $\Delta_{\kappa} = \Delta$ ,  $\lambda_{\kappa} = 0$ , and  $(p_t)_{t>0}$  is the heat kernel associated to  $\Delta$  on  $\mathbb{R}^d$ , i.e.,

$$p_t(x,y) = \frac{1}{(4\pi t)^{d/2}} \exp\left(-\frac{|x-y|^2}{4t}\right), \quad x,y \in \mathbb{R}^d, \ t > 0;$$

hence

$$-\Delta\big(\log p_t(\cdot, y)\big)(x) = \frac{d}{2t}, \quad x, y \in \mathbb{R}^d, \ t > 0$$

# Dunkl heat kernel: $\mathbb{Z}_2^d$ case

Let  $\Gamma(\cdot)$  be the Gamma function. For every t > 0,  $u, v \in \mathbb{R}$  and each  $i = 1, \dots, d$ , let

$$p_t^i(u,v) = \frac{1}{c_{\kappa_i}(2t)^{\kappa_i+1/2}} \exp\Big(-\frac{u^2+v^2}{4t}\Big) E_{\kappa_i}\Big(\frac{u}{\sqrt{2t}},\frac{v}{\sqrt{2t}}\Big),$$

where  $c_{\kappa_i} := \Gamma(\kappa_i + 1/2)$  and

$$E_{\kappa_i}(u,v) := \frac{\Gamma(\kappa_i+1/2)}{\Gamma(1/2)\Gamma(\kappa_i)} \int_{-1}^1 (1-s)^{\kappa_i-1} (1+s)^{\kappa_i} e^{suv} \,\mathrm{d}s.$$

Then, for every t > 0 and every  $x, y \in \mathbb{R}^d$  with  $x = (x_1, \dots, x_d)$  and  $y = (y_1, \dots, y_d)$ ,

$$p_t(x,y) = \prod_{i=1}^d p_t^i(x_i, y_i).$$

# Sketched proofs (I)

Let t > 0 and  $x, y \in \mathbb{R}^d$  with  $x = (x_1, \cdots, x_d), y = (y_1, \cdots, y_d)$ . Then  $\begin{aligned} & \Delta_\kappa \left(\log p_t(\cdot, y)\right)(x) \\ &= \sum_{j=1}^d \left(\frac{\kappa_j}{x_j^2} \left[2x_j \partial_{x_j} \log p_t^j(x_j, y_j) - \log p_t^j(x_j, y_j) + \log p_t^j(-x_j, y_j)\right] \\ &+ \partial_{x_j x_j}^2 \log p_t^j(x_j, y_j) \end{aligned}$ 

Hence, it suffices to estimate the terms in the parentheses, i.e.,

 $\frac{\kappa_j}{x_j^2} \Big[ 2x_j \partial_{x_j} \log p_t^j(x_j, y_j) - \log p_t^j(x_j, y_j) + \log p_t^j(-x_j, y_j) \Big] + \partial_{x_j x_j}^2 \log p_t^j(x_j, y_j),$ 

and it reduces to the rank-one case.

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$$\begin{split} & \Delta_{\kappa} \big(\log p_t(\cdot, y)\big)(x) \\ &= \sum_{j=1}^d \Big(\frac{\kappa_j}{x_j^2} \big[ 2x_j \partial_{x_j} \log p_t^j(x_j, y_j) - \log p_t^j(x_j, y_j) + \log p_t^j(-x_j, y_j) \big] \\ &+ \partial_{x_j x_j}^2 \log p_t^j(x_j, y_j) \Big). \end{split}$$

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and it reduces to the rank-one case.

# Sketched proofs (II)

Ignoring the subscript *j*, for every t > 0 and  $x, y \in \mathbb{R}$ , we let

$$\mathbf{I} := \partial_{xx}^2 \log p_t(x, y) + \frac{\kappa}{x^2} \big[ 2x \partial_x \log p_t(x, y) - \log p_t(x, y) + \log p_t(-x, y) \big],$$

where

$$p_t(x,y) = \frac{1}{c_{\kappa}(2t)^{\kappa+1/2}} \exp\left(-\frac{x^2+y^2}{4t}\right) E_{\kappa}\left(\frac{x}{\sqrt{2t}}, \frac{y}{\sqrt{2t}}\right),$$
$$E_{\kappa}\left(\frac{x}{\sqrt{2t}}, \frac{y}{\sqrt{2t}}\right) = C_{\kappa} \int_{-1}^{1} (1-s)^{\kappa-1} (1+s)^{\kappa} e^{\frac{sxy}{2t}} \, \mathrm{d}s.$$

Then we only need to estimate

$$I_1 := \partial_{xx}^2 \log p_t(x, y),$$
  

$$I_2 := \frac{\kappa}{x^2} \left[ 2x \partial_x \log p_t(x, y) - \log p_t(x, y) + \log p_t(-x, y) \right].$$

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where

$$p_t(x,y) = \frac{1}{c_{\kappa}(2t)^{\kappa+1/2}} \exp\left(-\frac{x^2+y^2}{4t}\right) E_{\kappa}\left(\frac{x}{\sqrt{2t}}, \frac{y}{\sqrt{2t}}\right) E_{\kappa}\left(\frac{x}{\sqrt{2t$$

Then we only need to estimate

$$I_1 := \partial_{xx}^2 \log p_t(x, y),$$
  

$$I_2 := \frac{\kappa}{x^2} \left[ 2x \partial_x \log p_t(x, y) - \log p_t(x, y) + \log p_t(-x, y) \right].$$

# Sketched proofs (III)

Let 
$$g(s) = (1 - s)^{\kappa - 1} (1 + s)^{\kappa}$$
 and  $a = xy/(2t)$ .

By the Cauchy-Schwarz inequality,

$$I_{1} = -\frac{1}{2t} + \frac{y^{2}}{4t^{2}} \left( \frac{\int_{-1}^{1} s^{2}g(s)e^{as} ds}{\int_{-1}^{1} g(s)e^{as} ds} - \frac{\left(\int_{-1}^{1} sg(s)e^{as} ds\right)^{2}}{\left(\int_{-1}^{1} g(s)e^{as} ds\right)^{2}} \right)$$
  

$$\geq -\frac{1}{2t}.$$

$$I_{2} = \frac{\kappa}{x^{2}} \left[ -\frac{x^{2}}{t} + 2a \frac{\int_{-1}^{1} sg(s)e^{as} \, ds}{\int_{-1}^{1} g(s)e^{as} \, ds} + \log \frac{\int_{-1}^{1} g(s)e^{-as} \, ds}{\int_{-1}^{1} g(s)e^{as} \, ds} \right]$$
  
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Thus,  $I = I_1 + I_2 \ge -(1 + 2\kappa)/(2t)$ .

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An immediate consequence of the Li–Yau inequality is the Harnack inequality.

#### Corollary (L.-Qian 2021)

Let  $T \in (0, \infty]$ . Suppose that G is isomorphic to  $\mathbb{Z}_2^d$  and  $u : [0, T) \times \mathbb{R}$  $\rightarrow (0, \infty)$  is a  $C^2$  solution to the Dunkl heat equation, i.e.,

$$\partial_t u(t,x) = \Delta_\kappa (u(t,\cdot))(x), \quad (t,x) \in (0,T) \times \mathbb{R}^d.$$

Then, for every 0 < s < t < T and every  $x, y \in \mathbb{R}^d$ ,

$$u(s,x) \leq u(t,y) \left(\frac{t}{s}\right)^{\lambda_{\kappa}+d/2} \exp\left(\frac{|x-y|^2}{4(t-s)}\right).$$

2 Li–Yau inequalities for Dunkl heat equation

## **3** Sharp Li–Yau inequalities for Dunkl heat kernel: $\mathbb{Z}_2^d$ case

2 Li–Yau inequalities for Dunkl heat equation

## 3 Sharp Li–Yau inequalities for Dunkl heat kernel: $\mathbb{Z}_2^d$ case

# Li-Yau inequalities of "gradient" type

Let 
$$f, g \in C^2(\mathbb{R}^d)$$
 and  $x \in \mathbb{R}^d$ . Then  

$$\Gamma(f, g)(x) := \frac{1}{2} \Big[ \Delta_{\kappa}(fg) - f \Delta_k g - g \Delta_{\kappa} f \Big](x)$$

$$= \langle \nabla f(x), \nabla g(x) \rangle + \sum_{\alpha \in \mathfrak{R}_+} \kappa_{\alpha} \frac{\big(f(x) - f(r_{\alpha} x)\big) \big(g(x) - g(r_{\alpha} x)\big)}{\langle \alpha, x \rangle^2}.$$

Set  $\Gamma(f) = \Gamma(f, f)$  for convenience.

Q: Does any of the following hold, i.e.,

$$\frac{\Gamma(p_t(\cdot, y))(x)}{p_t(x, y)^2} - \frac{\partial_t p_t(x, y)}{p_t(x, y)} \le \frac{d + 2\lambda_{\kappa}}{2t}$$

and

$$\frac{|\nabla_{\kappa} p_t(\cdot, y)(x)|^2}{p_t(x, y)^2} - \frac{\partial_t p_t(x, y)}{p_t(x, y)} \le \frac{d + 2\lambda_{\kappa}}{2t}?$$

" Thank you everyone! "