## Li–Yau Inequalities for Dunkl Heat Equations

黎怀谦 (Li, Huaiqian)

Center for Applied Mathematics Tianjin University Tianjin 300072, P. R. China

*(joint work with Qian, Bin)*

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2 [Li–Yau inequalities for Dunkl heat equation](#page-21-0)

# **3** [Sharp Li–Yau inequalities for Dunkl heat kernel:](#page-25-0)  $\mathbb{Z}_2^d$  case

- 2 [Li–Yau inequalities for Dunkl heat equation](#page-21-0)
- <span id="page-2-0"></span>**3** [Sharp Li–Yau inequalities for Dunkl heat kernel:](#page-25-0)  $\mathbb{Z}_2^d$  case

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

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

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

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

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

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

## Examples of root systems



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

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

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

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

Note that  $D_{\xi} \circ D_{\eta} = D_{\eta} \circ D_{\xi}, \xi, \eta \in \mathbb{R}^d$ , 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) dt = \int_0^1 \partial_{\alpha} f(x - t \langle \alpha, x \rangle \alpha) dt.
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 $f(x) - f(r_{\alpha}x)$  $\frac{-f(r_\alpha x)}{\langle \alpha, x \rangle} = -\frac{1}{\langle \alpha, x \rangle}$  $\langle \alpha, x \rangle$  $\int_0^1$  $\frac{\partial}{\partial t} f(x - t \langle \alpha, x \rangle \alpha) dt = \int_0^1$  $\int_{0}^{1} \partial_{\alpha} f(x-t \langle \alpha, x \rangle \alpha) dt.$ 

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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*<sup>j</sup>* = D*e<sup>j</sup>* . 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.
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In particular, when  $\kappa = 0$ ,  $\nabla_0 = \nabla$  and  $\Delta_0 = \Delta$ .

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#### Example (rank-one case)

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

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

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.,

$$
dY_t = dB_t + \sum_{\alpha \in \mathfrak{R}_+} \kappa_\alpha \frac{dt}{\langle \alpha, Y_t \rangle}, \quad Y_0 = y \in \overline{W},
$$

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

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## A remark on Dunkl process

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

 $∀I ⊂ ℜ<sub>+</sub>$ , let

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

 $(X_t)_{t>0}$  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

$$
\mu_{\kappa}(\mathrm{d}x):=\omega_{\kappa}(x)\mathrm{d}x,
$$

where d*x* denotes the Lebesgue measure on 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.  $\mu_k$ , 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,

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

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

- 2 [Li–Yau inequalities for Dunkl heat equation](#page-21-0)
- **3** [Sharp Li–Yau inequalities for Dunkl heat kernel:](#page-25-0)  $\mathbb{Z}_2^d$  case

### 2 [Li–Yau inequalities for Dunkl heat equation](#page-21-0)

# <span id="page-21-0"></span>**3** [Sharp Li–Yau inequalities for Dunkl heat kernel:](#page-25-0)  $\mathbb{Z}_2^d$  case

## 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  $u:[0,T)\times \mathbb{R}\rightarrow (0,\infty)$  *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*

<span id="page-22-0"></span>
$$
- \Delta_{\kappa} (\log p_t(\cdot, y))(x) \leq \beta(t, x), \quad (t, x, y) \in (0, T) \times \mathbb{R}^d \times \mathbb{R}^d, (1)
$$

*is equivalent to*

<span id="page-22-1"></span>
$$
- \Delta_{\kappa} \big( \log u(t, \cdot) \big)(x) \leq \beta(t, x), \quad (t, x) \in (0, T) \times \mathbb{R}^d; \tag{2}
$$

*moreover, either* [\(1\)](#page-22-0) *or* [\(2\)](#page-22-1) *implies*

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

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  $s \in (0, 1)$ .

- 2 [Li–Yau inequalities for Dunkl heat equation](#page-21-0)
- **3** [Sharp Li–Yau inequalities for Dunkl heat kernel:](#page-25-0)  $\mathbb{Z}_2^d$  case

2 [Li–Yau inequalities for Dunkl heat equation](#page-21-0)

# <span id="page-25-0"></span>**3** [Sharp Li–Yau inequalities for Dunkl heat kernel:](#page-25-0)  $\mathbb{Z}_2^d$  case

#### Theorem (L.–Qian 2021)

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

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

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:  $\mathbb{Z}_2^d$  $_2^d$  case

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

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

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

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

## Sketched proofs (I)

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

$$
\Delta_{\kappa} \left( \log p_t(\cdot, y) \right)(x)
$$
\n
$$
= \sum_{j=1}^d \left( \frac{\kappa_j}{x_j^2} \left[ 2x_j \partial_{x_j} \log p_t^j(x_j, y_j) - \log p_t^j(x_j, y_j) + \log p_t^j(-x_j, y_j) \right] + \partial_{x_j x_j}^2 \log p_t^j(x_j, y_j) \right).
$$

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

*x* 2 *j*  $\left[2x_j\partial_{x_j}\log p_t^j(x_j,y_j)-\log p_t^j(x_j,y_j)+\log p_t^j(-x_j,y_j)\right]+\partial_{x_jx_j}^2\log p_t^j(x_j,y_j),$ 

and it reduces to the rank-one case.

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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

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

where

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

$$
E_{\kappa}\Big(\frac{x}{\sqrt{2t}}, \frac{y}{\sqrt{2t}}\Big) = C_{\kappa} \int_{-1}^{1} (1 - s)^{\kappa - 1} (1 + s)^{\kappa} e^{\frac{sy}{2t}} ds.
$$

Then we only need to estimate

$$
I_1 := \frac{\partial_{xx}^2 \log p_t(x, y)}{x^2}.
$$
  
\n
$$
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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Ignoring the subscript *j*, for every  $t > 0$  and  $x, y \in \mathbb{R}$ , we let

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

where

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

$$
E_{\kappa}\Big(\frac{x}{\sqrt{2t}}, \frac{y}{\sqrt{2t}}\Big) = C_{\kappa} \int_{-1}^{1} (1 - s)^{\kappa - 1} (1 + s)^{\kappa} e^{\frac{sy}{2t}} ds.
$$

Then we only need to estimate

$$
I_1 := \frac{\partial_{xx}^2 \log p_t(x, y)}{\partial x^2}.
$$
  
\n
$$
I_2 := \frac{\kappa}{x^2} \left[ 2x \partial_x \log p_t(x, y) - \log p_t(x, y) + \log p_t(-x, y) \right].
$$

## Sketched proofs (III)

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

By the Cauchy–Schwarz inequality,

$$
I_1 = -\frac{1}{2t} + \frac{y^2}{4t^2} \left( \frac{\int_{-1}^1 s^2 g(s) e^{as} ds}{\int_{-1}^1 g(s) e^{as} ds} - \frac{\left(\int_{-1}^1 s g(s) e^{as} ds\right)^2}{\left(\int_{-1}^1 g(s) e^{as} ds\right)^2} \right)
$$
  
\n
$$
\geq -\frac{1}{2t}.
$$

$$
I_2 = \frac{\kappa}{x^2} \left[ -\frac{x^2}{t} + 2a \frac{\int_{-1}^1 s g(s) e^{as} ds}{\int_{-1}^1 g(s) e^{as} ds} + \log \frac{\int_{-1}^1 g(s) e^{-as} ds}{\int_{-1}^1 g(s) e^{as} ds} \right]
$$
  
\n
$$
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$$

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.,* 

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

*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](#page-21-0)
- **3** [Sharp Li–Yau inequalities for Dunkl heat kernel:](#page-25-0)  $\mathbb{Z}_2^d$  case

2 [Li–Yau inequalities for Dunkl heat equation](#page-21-0)

# <span id="page-37-0"></span>**3** [Sharp Li–Yau inequalities for Dunkl heat kernel:](#page-25-0)  $\mathbb{Z}_2^d$  case

# Li–Yau inequalities of "gradient" type

Let 
$$
f, g \in C^2(\mathbb{R}^d)
$$
 and  $x \in \mathbb{R}^d$ . Then  
\n
$$
\Gamma(f, g)(x) := \frac{1}{2} \left[ \Delta_{\kappa}(fg) - f \Delta_k g - g \Delta_{\kappa} f \right](x)
$$
\n
$$
= \langle \nabla f(x), \nabla g(x) \rangle + \sum_{\alpha \in \mathfrak{R}_+} \kappa_{\alpha} \frac{\left( f(x) - f(r_{\alpha}x) \right) \left( g(x) - g(r_{\alpha}x) \right)}{\langle \alpha, x \rangle^2}.
$$

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

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

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

and

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

" Thank you everyone! "