

Alma Mater Studiorum Università di Bologna Archivio istituzionale della ricerca

Discrete hilbert transform à la gundy-Varopoulos

This is the final peer-reviewed author's accepted manuscript (postprint) of the following publication:

Published Version:

Discrete hilbert transform à la gundy-Varopoulos / Arcozzi N.; Domelevo K.; Petermichl S.. - In: PROCEEDINGS OF THE AMERICAN MATHEMATICAL SOCIETY. - ISSN 0002-9939. - STAMPA. - 148:6(2020), pp. 2433-2446. [10.1090/proc/14492]

Availability:

This version is available at: https://hdl.handle.net/11585/802448 since: 2021-02-20

Published:

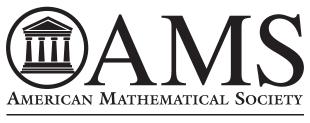
DOI: http://doi.org/10.1090/proc/14492

Terms of use:

Some rights reserved. The terms and conditions for the reuse of this version of the manuscript are specified in the publishing policy. For all terms of use and more information see the publisher's website.

This item was downloaded from IRIS Università di Bologna (https://cris.unibo.it/). When citing, please refer to the published version.

(Article begins on next page)



www.ams.org

N. Arcozzi, K. Domelevo, S. Petermichl Stochastic representation of a discrete Hilbert transform Proceedings of the American Mathematical Society DOI: 10.1090/proc/14492

Accepted Manuscript

This is a preliminary PDF of the author-produced manuscript that has been peer-reviewed and accepted for publication. It has not been copyedited, proofread, or finalized by AMS Production staff. Once the accepted manuscript has been copyedited, proofread, and finalized by AMS Production staff, the article will be published in electronic form as a "Recently Published Article" before being placed in an issue. That electronically published article will become the Version of Record.

This preliminary version is available to AMS members prior to publication of the Version of Record, and in limited cases it is also made accessible to everyone one year after the publication date of the Version of Record.

The Version of Record is accessible to everyone five years after publication in an issue.

STOCHASTIC REPRESENTATION OF A DISCRETE HILBERT TRANSFORM

N. ARCOZZI, K. DOMELEVO, AND S. PETERMICHL

ABSTRACT. We show that the centered discrete Hilbert transform on integers applied to a function can be written as the conditional expectation of a transform of stochastic integrals, where the stochastic processes considered have jump components. The stochastic representation of the function and that of its Hilbert transform are under differential subordination and orthogonality relation with respect to the sharp bracket of quadratic covariation. This illustrates the Cauchy Riemann relations of analytic functions in this setting. This result is inspired by the seminal work of Gundy and Varopoulos on stochastic representation of the Hilbert transform in the continuous setting.

1. Introduction

The subject of discrete analyticity and discrete Cauchy–Riemann relations has been investigated for a long time. It originated in the works of Ferrand [7] and Isaacs [12]. The relationship of Cauchy–Riemann relations to a certain discrete Hilbert transform was understood in Duffin [5] together with the corresponding notions of discrete harmonic conjugate functions. The discrete Hilbert transform also appeared in relation with the Riesz–Titchmarsh transform as described in the significant discovery by Matsaev [17]. See also Matsaev–Sodin [16].

Despite the fact that (different notions of) discrete Hilbert transforms exist for a long time, the precise L^p norm of these discrete operators are still a very famous open question. Optimal norm estimates are only known in the continuous case – see Pichorides [18], Verbitsky [22], Essen [6] – whose proofs have a probabilistic reinterpretation, in part through the formulae of Gundy-Varopoulos [10]. It is remarkable that such representations allow one to prove sharp L^p estimates for discrete second order Riesz transforms as proved in [4] by using the Bellman technique and in [2] by using stochastic tools. The versatility of these stochastic representations is also seen in [3] where various sharp estimates for discrete second order Riesz transforms are proved. Inspired by this fact, we aim at an understanding of a discrete Hilbert transform through a stochastic integral formula, resembling the continuous analog of Gundy-Varopoulos.

Regarding L^p -norm estimates for the discrete Hilbert transform of the type we study below, to the best of our knowledge the best constant is by Gohberg and Krupnik [8]. One important ingredient in the proof of sharp estimates is the use of harmonic polynomials and other special functions. There exist important discrepancies between harmonic functions and polynomials in the continuous and discrete settings, which is part of the reason why the sharp constants for the continuous case

Research supported by the ERC project CHRiSHarMa DLV-862402. S.P. is a member of IUF.

cannot be transferred in a straightforward manner to the discrete case. We quote in that direction the early work of Heilbronn [11] and the more recent works [13][14][9].

It is our main goal to prove a stocahstic representation formula for a discrete Hilbert transform. One should keep in mind that there are different naive ways of defining discrete Hilbert transforms from the space of sequences $\ell^2(\mathbb{Z})$ onto itself. A first manner found in the litterature is to mimic the continuous Hilbert transform $\mathcal{H}_{\mathbb{R}}$ defined on the real line as

$$\forall x \in \mathbb{R}, \quad \mathcal{H}_{\mathbb{R}}(f)(x) = -\frac{1}{\pi} \int \frac{f(x-y)}{y} dy,$$

where the integral is to be understood in the principal value sense. Indeed, a naive discrete counterpart of $\mathcal{H}_{\mathbb{R}}$ defined thanks to the discrete convolution

$$\mathcal{H}_{\mathbb{Z}}: \ell^2(\mathbb{Z}) \to \ell^2(\mathbb{Z}), \quad \forall x \in \mathbb{Z}, \quad \mathcal{H}_{\mathbb{Z}}(f)(x) = -\frac{1}{\pi} \sum_{z \in \mathbb{Z}_+} \frac{f(x-z)}{z}$$

preserves the idea of a principal value integral by skipping z=0 in the summation, but lacks many other important features of the continuous Hilbert transform. For instance, the operator $\mathcal{H}_{\mathbb{R}}$ is an isometry in $L^2(\mathbb{R})$, an anti-involution, i.e. $\mathcal{H}^2_{\mathbb{R}} := \mathcal{H}_{\mathbb{R}} \circ \mathcal{H}_{\mathbb{R}} = -\operatorname{Id}_{L^2(\mathbb{R})}$ and obeys $\mathcal{H}_{\mathbb{R}} \circ \sqrt{-\Delta_x} = \partial_x$. However, $\mathcal{H}_{\mathbb{Z}}$ above does not possess these basic properties. One observes that the iterate $\mathcal{H}^2_{\mathbb{Z}} = \mathcal{H}_{\mathbb{Z}} \circ \mathcal{H}_{\mathbb{Z}}$ is far from being neither an isometry of $\ell^2(\mathbb{Z})$ nor an anti-involution. One reason for that is the fact that the summation in the discrete convolution excludes the integer z=0.

In this paper, we take the following route, as done in Lust-Piquard [15]. Modelled after the defining equation for the continuous Hilbert transform, $\mathcal{H}_{\mathbb{R}} \circ \sqrt{-\Delta_x} = \partial_x$, let us recall the definition of the discrete Laplacian on \mathbb{Z} : we define the discrete derivatives as

$$(\partial_x^+ f)(x) := f(x+1) - f(x), \quad (\partial_x^- f)(x) := f(x) - f(x-1).$$

In $\ell^2(\mathbb{Z})$, it follows that $(\partial_x^{\pm})^* = -\partial_x^{\mp}$, $\Delta_x = \partial_x^{+}\partial_x^{-} = \partial_x^{-}\partial_x^{+} = \partial_x^{+} - \partial_x^{-}$, and $(-\Delta_x) = (\partial_x^{\pm})^*(\partial_x^{\pm})$. Equipped with these discrete operators, another classical definition for discrete Hilbert transforms is given by

$$\mathcal{H}^{\pm} \circ \sqrt{-\Delta_x} = \partial_x^{\pm}.$$

Those are the **left** (resp. **right**) discrete Hilbert transform \mathcal{H}^+ (resp. \mathcal{H}^-), most often used when defining Riesz transforms on discrete groups (see LUST-PIQUARD [15] for more details and applications to the discrete Riesz vector). Through explicit and simple calculations, we are going to see in the next section that

$$\mathcal{H}^+\mathcal{H}^- = \mathcal{H}^-\mathcal{H}^+ = -\operatorname{Id}, \qquad \|\mathcal{H}^{\pm}\|_{\ell^2(\mathbb{Z}) \to \ell^2(\mathbb{Z})} = 1$$

and that the kernels of these operators \mathcal{H}^{\pm} are

$$-\frac{1}{\pi} \frac{1}{n \pm 1/2}$$

respectively.

The Fourier multiplier of $\mathcal{H}_{\mathbb{R}}$ is constant on positive and negative frequencies respectively. This is a feature the operators \mathcal{H}^{\pm} lack. In fact, these operators have Fourier multipliers that are a modulation of the square wave function.

As mentioned earlier, another important and meaningful role is played by the interplay of $\mathcal{H}_{\mathbb{R}}$ and harmonic conjugate functions. If u^f is the Poisson extension of a function f to the upper half plane and v^f the Poisson extension of $\mathcal{H}_{\mathbb{R}}f$, then the pair u^f and v^f obeys Cauchy-Riemann relations. Using space—time Brownian motion, Gundy and Varopoulos have identified pairs of martingales M_t^f and $M_t^{\mathcal{H}_{\mathbb{R}}f}$ that are orthogonal and have differential subordination. In fact, in their formula, $M_t^{\mathcal{H}_{\mathbb{R}}f}$ is a martingale transform of M_t^f . The discrete counterpart of this feature of $\mathcal{H}_{\mathbb{R}}$ is the main focus of our note. Using stochastic integrals driven by semidiscrete random walks in the (semidiscrete) upper-half space, we will see that the **centered** discrete Hilbert transform defined as

(1)
$$\mathcal{H} = \frac{1}{2}(\mathcal{H}^+ + \mathcal{H}^-)$$

enjoys this stochastic representation. Indeed, we obtain for \mathcal{H} certain Cauchy-Riemann relations and an analog of the Gundy-Varopoulos formula. In fact, $M_t^{\mathcal{H}f}$ and M_t^f are orthogonal and differentially subordinate with respect to the sharp bracket $\langle \cdot, \cdot \rangle$.

The main goal of the paper is to prove the following representation formula \grave{a} la Gundy-Varopoulos [10]:

Theorem 1. (Stochastic representation) The centered discrete Hilbert transform $\mathcal{H}f$ of a function $f \in \ell^2(\mathbb{Z})$ as defined in (1) can be written as the conditional expectation

$$\forall x \in \mathbb{Z}, \quad \mathcal{H}f(x) = \mathbb{E}(N_0^f | \mathcal{Z}_0 = (x, 0))$$

where N_t^f , $-\infty < t \leq 0$, is a suitable martingale transform of a martingale M_t^f associated to f, and \mathcal{Z}_t is a suitable semidiscrete random walk on the semidiscrete upper-half space $\mathbb{Z} \times \mathbb{R}^+$.

Outline of the paper. The next section is devoted to a few basic properties of the discrete hilbert transforms mentionned above. Section 3 provides semidiscrete Poisson extensions, weak formulations and semidiscrete Cauchy–Riemann relations. We introduce the relevant stochastic integrals, martingale transforms and quadratic covariations in Section 4. Finally, we prove the representation \grave{a} la Gundy–Varopoulos of the centered discrete Hilbert transform in Section 5.

2. Basic properties

Let \mathcal{F} be the discrete fourier transform

$$\mathcal{F} : \ell^2(\mathbb{Z}) \longrightarrow L^2\left(\left] - \frac{1}{2}, \frac{1}{2}\right[\right)$$

$$\mathcal{F}(f)(\xi) := \hat{f}(\xi) = \sum_{x \in \mathbb{Z}} f(x) e^{-\mathrm{i} 2\pi x \xi}$$

with Fourier inverse

$$\mathcal{F}^{-1} : L^{2}\left(\left]-\frac{1}{2}, \frac{1}{2}\right[\right) \longrightarrow \ell^{2}(\mathbb{Z})$$
$$\mathcal{F}^{-1}(\hat{f})(x) = \int_{-1/2}^{1/2} \hat{f}(\xi)e^{+i2\pi x\xi}d\xi$$

Through explicit and simple calculations, we are going to see that

Proposition 1. (Equivalent definitions of \mathcal{H}^{\pm})

(2)
$$\widehat{\mathcal{H}^{\pm}}(\xi) = e^{\pm i\pi\xi} \frac{\sin(\pi\xi)}{|\sin(\pi\xi)|} = e^{i\pi\xi} \operatorname{SQ}(\pi\xi)$$

(3)
$$\mathcal{H}^{\pm}f(x) = -\frac{1}{\pi} \sum_{z \in \mathbb{Z}} \frac{f(x-z)}{z \pm \frac{1}{2}}$$

As a consequence, the reader may check that we also have the following properties,

Proposition 2. (Basic properties) The discrete Hilbert transforms \mathcal{H}^{\pm} obey the following analogs of the continuous Hilbert transform

$$(4) \qquad \forall f, g \in \ell^2(\mathbb{Z}), \quad (\mathcal{H}^{\pm}f, g)_{\ell^2(\mathbb{Z})} = -(f, \mathcal{H}^{\mp}g)_{\ell^2(\mathbb{Z})}, \quad i.e.(\mathcal{H}^{\pm})^* = -\mathcal{H}^{\mp}$$

(5)
$$\mathcal{H}^+\mathcal{H}^- = \mathcal{H}^-\mathcal{H}^+ = -\operatorname{Id}$$

(6)
$$\mathcal{H}^{\pm}\mathcal{H}^{\pm} = -S_{\pm 1} \operatorname{Id}$$

(7)
$$\|\mathcal{H}^{\pm}\|_{\ell^2(\mathbb{Z}) \to \ell^2(\mathbb{Z})} = 1$$

where S_{+1} (resp. S_{-1}) is the right (resp. left) shift operator $(S_{\pm 1}f)(x) = f(x \mp 1)$.

Proof. (of Proposition 1) One has succesively

$$\widehat{\partial_x^{\pm}}(\xi) = e^{\pm i2\pi\xi} - 1 = 2ie^{\pm i\pi\xi}\sin(\pi\xi)$$

$$\widehat{\Delta_x}(\xi) = -4\sin^2(\pi\xi)$$

$$\widehat{\sqrt{-\Delta_x}}(\xi) = 2|\sin(\pi\xi)|$$

$$\widehat{\mathcal{H}^{\pm}}(\xi) = ie^{\pm i\pi\xi}\frac{\sin(\pi\xi)}{|\sin(\pi\xi)|} =: ie^{i\pi\xi}\operatorname{SQ}(\pi\xi).$$

Now, computing the Fourier transform of the discrete kernel, we check

$$\begin{split} \sum_{y \in \mathbb{Z}} -\frac{1}{\pi} \frac{1}{y + \frac{1}{2}} e^{-2\pi \mathrm{i} y \xi} &= -\frac{2}{\pi} \sum_{y} \frac{e^{-\mathrm{i}(2y+1)\frac{\xi}{2}}}{2y+1} e^{2\pi \mathrm{i} \frac{\xi}{2}} \\ &= -\frac{2}{\pi} \mathrm{i} e^{\pi \mathrm{i} \xi} \sum_{m} \frac{\sin\left(-2\pi (2y+1)\frac{\xi}{2}\right)}{2y+1} \\ &= -\frac{4}{\pi} \mathrm{i} e^{2\pi \mathrm{i} \frac{\xi}{2}} \sum_{m \geqslant 0} \frac{\sin\left(2\pi (2y+1)\left(-\frac{\xi}{2}\right)\right)}{2y+1} \\ &= -\mathrm{i} e^{\pi \mathrm{i} \xi} \operatorname{SQ}\left(-\frac{\pi \xi}{2}\right) \\ &= \mathrm{i} e^{\pi \mathrm{i} \xi} \frac{\sin(\pi \xi)}{|\sin(\pi \xi)|}, \end{split}$$

where we used the Fourier transform of the square wave function

$$SQ(t) = \frac{4}{\pi} \sum_{n=0}^{\infty} \frac{\sin(\frac{2\pi}{T}(2n+1)t)}{2n+1}.$$

Similarly, the symbol of \mathcal{H}^- is $ie^{-\pi i\xi} \frac{\sin(\pi\xi)}{|\sin(\pi\xi)|} = -ie^{-\pi i\xi} \operatorname{SQ}\left(-\frac{\xi}{2}\right)$

Finally, for the centered discrete Hilbert transform, one has

(8)
$$\mathcal{H} \circ \mathcal{H} = [(\mathcal{H}^{+})^{2} + 2(\mathcal{H}^{+})(\mathcal{H}^{-}) + (\mathcal{H}^{-})^{2}]/4$$

$$= [(\mathcal{H}^{+})^{2} + 2(\mathcal{H}^{+})(\mathcal{H}^{-}) + (\mathcal{H}^{-})^{2}]/4$$

$$= [(\mathcal{H}^{+})S^{+}(\mathcal{H}^{-}) + 2(\mathcal{H}^{+})(\mathcal{H}^{-}) + S^{-}(\mathcal{H}^{+})(\mathcal{H}^{-})]/4$$

$$= -\left(\frac{1}{4}S_{-} + \frac{1}{2}\operatorname{Id} + \frac{1}{4}S_{+}\right),$$

where we recognize a smoothed version of minus the identity.

3. Semi-discrete Poisson extensions and weak formulations

Semidiscrete Poisson extensions. Defining the selfadjoint operator A as the square root $A = \sqrt{-\Delta_x}$, we set $P_y = e^{-yA}$, $y \in [0, \infty[$. The Poisson extension of a function $f \in \ell^2(\mathbb{Z})$ is the function $f(y, x) := (P_y f)(x)$ defined on $\mathbb{R}^+ \times \mathbb{Z}$. Note that we use the same name for f and it's Poisson extension. It follows that the function f(y, x) satisfies in $(0, \infty) \times \mathbb{Z}$

$$\partial_y f(x,y) = -Af(x,y), \quad \partial_y^2 f(x,y) = A^2 f(x,y) = -\Delta_x f(x,y)$$

that is f(x,y) is harmonic in $(0,\infty)\times\mathbb{Z}$:

$$(\Delta_y + \Delta_x)f = 0.$$

For any f and g in $\ell^2(\mathbb{Z})$, we note

$$(f,g)_{\ell^2(\mathbb{Z})} := \sum_{x \in \mathbb{Z}} f(x)g(x)$$

the scalar product of $\ell^2(\mathbb{Z})$. We will often omit the subscript in the scalar product and write (f,g) instead of $(f,g)_{\ell^2(\mathbb{Z})}$. Moreover we have

$$\forall f, g \in \ell^2(\mathbb{Z}), \quad (-\Delta_x f, g) := ((\partial_x^{\pm})^{\star}(\partial_x^{\pm})f, g) = (\partial_x^{\pm} f, \partial_x^{\pm} g)$$

and also

$$\forall f, g \in \ell^2(\mathbb{Z}), \quad (-\Delta_x f, g) = (A^2 f, g) = (Af, Ag) = (\partial_u f, \partial_u g)$$

We will collect below all the derivations in the 4-vector

$$\nabla_{y,x} = (\partial_y, \partial_x^+, \partial_y, \partial_x^-)^*.$$

Notice that we repeat twice the derivation ∂_y . The reasons for that will become clear later.

Theorem 2. (Weak formulation of the identity operator) Assume f and g in $\ell^2(\mathbb{Z})$. Let \mathcal{I} denote the identity operator. We have the Littlewood–Paley identity

$$(\mathcal{I}f,g) = \int_{y=0}^{\infty} \sum_{x \in \mathbb{Z}} (\nabla_{x,y} f(x,y), \nabla_{x,y} g(x,y)) y dy$$

Proof. Notice first that for any functions f and g in $\ell^2(\mathbb{Z})$, we have, using successively discrete integration by parts, and the definition of A,

$$(-\Delta_x f, g) := ((\partial_x^{\pm})^* (\partial_x^{\pm}) f, g) = (\partial_x^{\pm} f, \partial_x^{\pm} g)$$
$$= (A^2 f, g) = (Af, Ag)$$

In the particular case where both f = f(x, y) and g = g(x, y) are Poisson extensions, then

$$(Af, Ag) = (\partial_u f, \partial_u g)$$

To summarize, when the functions $f(y) = P_y f$ and $g(y) = P_y g$ are Poisson extensions, we have in the upper half space

$$(-\Delta_x f(y), g(y)) = (\partial_x^{\pm} f(y), \partial_x^{\pm} g(y)) = (\partial_y f(y), \partial_y g(y)) = (Af(y), Ag(y)).$$

But for any function F(y) smooth enough and decaying at infinity, we have

$$F(0) = \int_0^\infty F''(y)y dy$$

Applying this identity to $F(y) = (f(y), g(y))_{\ell^2(\mathbb{Z})}$ yields,

$$(f,g)_{\ell^{2}(\mathbb{Z})} = \int_{0}^{\infty} \{(\partial_{y}^{2}f,g) + 2(\partial_{y}f,\partial_{y}g) + (f,\partial_{y}^{2}g)\}ydy$$

$$= 4\int_{0}^{\infty} (\partial_{y}f,\partial_{y}g)ydy$$

$$= 4\int_{0}^{\infty} (\partial_{x}^{+}f,\partial_{x}^{+}g)ydy = 4\int_{0}^{\infty} (\partial_{x}^{-}f,\partial_{x}^{-}g)ydy$$

$$= \int_{0}^{\infty} \{(\partial_{y}f,\partial_{y}g) + (\partial_{x}^{+}f,\partial_{x}^{+}g) + (\partial_{y}f,\partial_{y}g) + (\partial_{x}^{-}f,\partial_{x}^{-}g)\}ydy$$

$$= \int_{0}^{\infty} (\nabla_{y,x}f,\nabla_{y,x}g)ydy,$$

as announced.

Cauchy-Riemman relations and weak formulation.

Theorem 3. (Cauchy-Riemann relations) Let f and g in $\ell^2(\mathbb{Z})$. Let $\mathcal{H}f(y,x)$ denote the Poisson extension of $\mathcal{H}f(x)$, and f(y,x) that of f(x). We have the semidiscrete Cauchy-Riemann relations, for all $(y,x) \in \mathbb{R}^+ \times \mathbb{Z}$,

$$\partial_y \mathcal{H}^{\pm} f = -\partial_x^{\mp} f, \quad \partial_x^{\pm} \mathcal{H}^{\pm} f = \partial_y f,$$

which implies for the centered discrete Hilbert transform, for all $(y, x) \in \mathbb{R}^+ \times \mathbb{Z}$,

$$\partial_y \mathcal{H}f = -\partial_x^0 f, \quad \partial_x^0 \mathcal{H}f = \left(\frac{1}{4}S_- + \frac{1}{2}\operatorname{Id} + \frac{1}{4}S_+\right)\partial_y f.$$

Theorem 4. (Weak formulation for the discrete Hilbert) Let $\mathcal{H}f$ denote the centered discrete Hilbert transform of f. Let f := f(y,x) denote the Poisson extension of f. Let g := g(y,x) denote the Poisson extension of a test function g. We have the weak formulation:

$$(\mathcal{H}f,g) = \int_0^\infty (A\nabla_{y,x}f, \nabla_{y,x}g)y dy$$

where $A \in \mathbb{R}^{4 \times 4}$ is the matrix

$$A = \left(\begin{array}{cccc} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{array}\right).$$

It is important to observe that the weak formulation involves the orthogonal matrix A such that $A^2 = -\operatorname{Id}$. This does not mean that the centered discrete Hilbert transform \mathcal{H} is an antiinvolution, as is clear from Theorem 3 and equation (8).

Proof. The semidiscrete Poisson extension convenient for us has up to normalization the kernel

$$P_y(x) = \int_0^1 e^{-2\pi i \xi x} e^{-2|\sin(\pi \xi)|y} d\xi$$

Let us write for convenience u the Poisson extension of f

$$u(x,y) = \int_0^1 \hat{f}(\xi) e^{-2\pi i \xi x} e^{-2|\sin(\pi \xi)|y} d\xi$$

and v^{\pm} the Poisson extension of $\mathcal{H}^{\pm}f$

$$v^{\pm}(x,y) = \int_{0}^{1} \widehat{\mathcal{H}_{\pm}f}(\xi) e^{-2\pi i \xi x} e^{-2|\sin(\pi \xi)|y} d\xi$$
$$= -i \int_{0}^{1} e^{\pm \pi i \xi} \frac{\sin(\pi \xi)}{|\sin(\pi \xi)|} \widehat{f}(\xi) e^{-2\pi i \xi x} e^{-2|\sin(\pi \xi)|y} d\xi$$

Observe that

$$\begin{array}{lcl} \partial_x^+ u(x,y) & = & \int_0^1 \hat{f}(\xi) (e^{-2\pi \mathrm{i}\xi(x+1)} - e^{-2\pi \mathrm{i}\xi x}) e^{-2|\sin(\pi\xi)|y} d\xi \\ & = & -2\mathrm{i} \int_0^1 e^{-\pi \mathrm{i}\xi} \sin(\pi\xi) \hat{f}(\xi) e^{-2\pi \mathrm{i}\xi x} e^{-2|\sin(\pi\xi)|y} d\xi \end{array}$$

While

$$\partial_x^- u(x,y) = \int_0^1 \hat{f}(\xi) (e^{-2\pi i \xi x} - e^{-2\pi i \xi (x-1)}) e^{-2|\sin(\pi \xi)|y} d\xi$$
$$= -2i \int_0^1 e^{\pi i \xi} \sin(\pi \xi) \hat{f}(\xi) e^{-2\pi i \xi x} e^{-2|\sin(\pi \xi)|y} d\xi$$

Similarly

$$\partial_x^+ v^+(x,y) = -2 \int_0^1 |\sin(\pi\xi)| \hat{f}(\xi) e^{-2\pi i \xi x} e^{-2|\sin(\pi\xi)|y} d\xi$$

and

$$\partial_x^- v^-(x,y) = -2 \int_0^1 |\sin(\pi\xi)| \hat{f}(\xi) e^{-2\pi i \xi x} e^{-2|\sin(\pi\xi)|y} d\xi$$

The continuous derivatives in the other variable are

$$\partial_y u(x,y) = -2 \int_0^1 |\sin(\pi\xi)| \hat{f}(\xi) e^{-2\pi i \xi x} e^{-2|\sin(\pi\xi)|y} d\xi$$

8

and

$$\partial_y v^{\pm}(x,y) = 2i \int_0^1 e^{\pm \pi i \xi} \sin(\pi \xi) \hat{f}(\xi) e^{-2\pi i \xi x} e^{-2|\sin(\pi \xi)|y} d\xi$$

This gives the following Cauchy Riemann equations

$$\partial_y v^+ = -\partial_x^- u,$$
$$\partial_u v^- = -\partial_x^+ u.$$

but

$$\partial_y u = \partial_x^+ v^+ = \partial_x^- v^-.$$

Let us note u^g the Poisson extension of a test function g and accordingly $u = u^f$ the Poisson extension of f. We have successively

$$\begin{split} (\mathcal{H}^+f,g) &=& 2\int_0^\infty (\partial_y v^+(y),\partial_y u^g(y))y\mathrm{d}y + 2\int_0^\infty (\partial_x^+ v^+(y),\partial_x^+ u^g(y))y\mathrm{d}y \\ &=& 2\int_0^\infty (-\partial_x^- u^f(y),\partial_y u^g(y))y\mathrm{d}y + 2\int_0^\infty (\partial_y u^f(y),\partial_x^+ u^g(y))y\mathrm{d}y \\ (\mathcal{H}^-f,g) &=& 2\int_0^\infty (-\partial_x^+ u^f(y),\partial_y u^g(y))y\mathrm{d}y + 2\int_0^\infty (\partial_y u^f(y),\partial_x^- u^g(y))y\mathrm{d}y \\ (\mathcal{H}f,g) &=& \int_0^\infty (-\partial_x^- u^f(y),\partial_y u^g(y))y\mathrm{d}y + \int_0^\infty (\partial_y u^f(y),\partial_x^+ u^g(y))y\mathrm{d}y \\ &&+ \int_0^\infty (-\partial_x^+ u^f(y),\partial_y u^g(y))y\mathrm{d}y + \int_0^\infty (\partial_y u^f(y),\partial_x^- u^g(y))y\mathrm{d}y \\ &=& \int_0^\infty \left(\begin{pmatrix} -\partial_x^- \\ \partial_y \\ -\partial_x^+ \\ \partial_y \end{pmatrix} u^f(y), \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \right)y\mathrm{d}y \\ &=& \int_0^\infty \left(\begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^f(y), \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \right)y\mathrm{d}y \\ &=& \int_0^\infty \left(\begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^f(y), \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \right)y\mathrm{d}y \\ &=& \int_0^\infty \left(\begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \right) \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^f(y), \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \right)y\mathrm{d}y \\ &=& \int_0^\infty \left(\begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \right) \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^f(y), \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \right)y\mathrm{d}y \\ &=& \int_0^\infty \left(\begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \right) \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y), \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \right)y\mathrm{d}y \\ &=& \int_0^\infty \left(\begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \right) \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y), \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \right)y\mathrm{d}y \\ &=& \int_0^\infty \left(\begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \right) \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x^+ \\ \partial_y \\ \partial_x^- \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\ \partial_x \\ \partial_y \\ \partial_x^+ \end{pmatrix} u^g(y) \\ \begin{pmatrix} \partial_y \\$$

This concludes the proof of Theorems 3 and 4.

4. Stochastic representations

Semidiscrete random walks. Let \mathcal{N}_t be a càdlàg Poisson process with parameter λ . Let $(T_k)_{k\in\mathbb{N}}$ be the instants of jumps. Let $(\varepsilon_k)_{k\in\mathbb{N}}$ be a sequence of independent Bernouilli variables,

$$\forall k$$
, $\mathbb{P}(\varepsilon_k = 1) = \mathbb{P}(\varepsilon_k = -1) = 1/2$.

This allows us to define the random walk $\mathcal{X}_t \in \mathbb{Z}$ as the compound Poisson process (see e.g. PROTTER [21], PRIVAULT [19][20])

$$\mathcal{X}_0 \in \mathbb{Z}, \quad \mathcal{X}_t = \sum_{k=1}^{\mathcal{N}_t} \varepsilon_k$$

Let \mathcal{Y}_t be a standard one-dimensional brownian process started at \mathcal{Y}_0 . We define the semidiscrete random walk \mathcal{Z}_t as $\mathcal{Z}_t := (\mathcal{Y}_t, \mathcal{X}_t) \in \mathbb{R}^+ \times \mathbb{Z}$, i.e.

$$\mathcal{Z}_{l} = (\mathcal{Y}_{l}, \mathcal{X}_{l}), \quad d\mathcal{Z}_{t} = (dB_{t}, \varepsilon_{\mathcal{N}_{t}} d\mathcal{N}_{t})$$

Stochastic integrals. Let f defined on $\mathbb{R}^+ \times \mathbb{Z}$ be a smooth function. We have the Itô formula

$$f(\mathcal{Z}_t) - f(\mathcal{Z}_0) = \int_0^t \partial_x^0 f(\mathcal{Z}_{s_-}) d\mathcal{X}_s + \int_0^t \frac{1}{2} \partial_x^2 f(\mathcal{Z}_{s_-}) d(\mathcal{N}_s - s) + \int_0^t \partial_y f(\mathcal{Z}_{s_-}) d\mathcal{Y}_s + \left\{ \frac{1}{2} \int_0^t (\partial_x^2 f + \partial_y^2 f)(\mathcal{Z}_{s_-}) ds \right\}$$

This formula can be derived thanks to telescopic sums involving jump times. We refer to Privault [19][20] for more details.

Quadratic covariation. Let f and g be two semidiscrete harmonic functions in $\mathbb{R}^+_{\star} \times \mathbb{Z}$, that is

$$\partial_x^2 f + \partial_y^2 f = \partial_x^2 g + \partial_y^2 g = 0$$
 in $\mathbb{R}_{\star}^+ \times \mathbb{Z}$

We define the corresponding martingales $M_t^f := f(\mathcal{Z}_t)$ and $M_t^g := g(\mathcal{Z}_t)$, so that

$$dM_t^f = \partial_x^0 f(\mathcal{Z}_{t_-}) d\mathcal{X}_t + \frac{1}{2} \partial_x^2 f(\mathcal{Z}_{t_-}) d(\mathcal{N}_t - t) + \partial_y f(\mathcal{Z}_{t_-}) d\mathcal{Y}_t$$

$$dM_t^g = \partial_x^0 g(\mathcal{Z}_{t_-}) d\mathcal{X}_t + \frac{1}{2} \partial_x^2 g(\mathcal{Z}_{t_-}) d(\mathcal{N}_t - t) + \partial_y g(\mathcal{Z}_{t_-}) d\mathcal{Y}_t$$

It follows that

$$\begin{split} \mathrm{d}[M^f,M^g]_t &= \partial_x^0 f(\mathcal{Z}_{t_-})\partial_x^0 g(\mathcal{Z}_{t_-})\mathrm{d}[\mathcal{X},\mathcal{X}]_t + \frac{1}{4}\partial_x^2 f(\mathcal{Z}_{t_-})\partial_x^2 g(\mathcal{Z}_{t_-})\mathrm{d}[\mathcal{N},\mathcal{N}]_t \\ &\quad + \left(\partial_x^0 f(\mathcal{Z}_{t_-})\frac{1}{2}\partial_x^2 g(\mathcal{Z}_{t_-}) + \partial_x^0 g(\mathcal{Z}_{t_-})\frac{1}{2}\partial_x^2 f(\mathcal{Z}_{t_-})\right)\mathrm{d}[\mathcal{X},\mathcal{N}]_t \\ &\quad + \partial_y f(\mathcal{Z}_{t_-})\partial_y g(\mathcal{Z}_{t_-})\mathrm{d}[\mathcal{Y},\mathcal{Y}]_t \\ &= \left[\partial_x^0 f(\mathcal{Z}_{t_-})\partial_x^0 g(\mathcal{Z}_{t_-}) + \frac{1}{2}\partial_x^2 f(\mathcal{Z}_{t_-})\frac{1}{2}\partial_x^2 g(\mathcal{Z}_{t_-})\right]\mathrm{d}\mathcal{N}_t \\ &\quad + \left[\partial_x^0 f(\mathcal{Z}_{t_-})\frac{1}{2}\partial_x^2 g(\mathcal{Z}_{t_-}) + \partial_x^0 g(\mathcal{Z}_{t_-})\frac{1}{2}\partial_x^2 f(\mathcal{Z}_{t_-})\right]\mathrm{d}\mathcal{X}_t \\ &\quad + \partial_y f(\mathcal{Z}_{t_-})\partial_y g(\mathcal{Z}_{t_-})\mathrm{d}t \\ &= \frac{1}{2}[\partial_x^+ f\partial_x^+ g + \partial_x^- f\partial_x^- g]\mathrm{d}\mathcal{N}_t \\ &\quad + \partial_y f(\mathcal{Z}_{t_-})\partial_y g(\mathcal{Z}_{t_-})\mathrm{d}t \\ &= [(\Delta\mathcal{X}_t)_+ \partial_x^+ f\partial_x^+ g + (\Delta\mathcal{X}_t)_- \partial_x^- f\partial_x^- g]\mathrm{d}\mathcal{N}_t \\ &\quad + \partial_y f(\mathcal{Z}_{t_-})\partial_y g(\mathcal{Z}_{t_-})\mathrm{d}t \end{split}$$

where $(\Delta \mathcal{X})_{\pm} := \max(0, \pm \Delta \mathcal{X}).$

Martingale transform. In order to define a martingale transform N_t^f of M_t^f such that N_t^f allows us to recover the discrete hilbert transform, we recall that the weak formulation (9) of Theorem 4 involves the "martingale transform"

$$\nabla_{y,x} = \begin{pmatrix} \partial_y^- \\ \partial_x^+ \\ \partial_y^+ \\ \partial_x^- \end{pmatrix} \longrightarrow \nabla_{y,x}^{\perp} = \begin{pmatrix} -\partial_x^- \\ \partial_y^+ \\ -\partial_x^+ \\ \partial_y^- \end{pmatrix}.$$

Let us first rewrite the martingale increments in terms of the ∂_x^{\pm} derivatives:

$$dM_{t}^{f} = \partial_{x}^{0} f(\mathcal{Z}_{t-}) d\mathcal{X}_{t} + \frac{1}{2} \partial_{x}^{2} f(\mathcal{Z}_{t-}) d(\mathcal{N}_{t} - t) + \partial_{y} f(\mathcal{Z}_{t-}) d\mathcal{Y}_{t}$$

$$= \frac{1}{2} \partial_{x}^{+} f(\mathcal{Z}_{t-}) (d\mathcal{X}_{t} + d(\mathcal{N}_{t} - t)) + \frac{1}{2} \partial_{x}^{-} f(\mathcal{Z}_{t-}) (d\mathcal{X}_{t} - d(\mathcal{N}_{t} - t))$$

$$+ \frac{1}{2} \partial_{y}^{-} f(\mathcal{Z}_{t-}) d\mathcal{Y}_{t} + \frac{1}{2} \partial_{y}^{+} f(\mathcal{Z}_{t-}) d\mathcal{Y}_{t}.$$

The Cauchy-Riemann relations therefore suggest to define

$$dN_t^f := \frac{1}{2} \partial_y^+ f(\mathcal{Z}_{t-}) (d\mathcal{X}_t + d(\mathcal{N}_t - t)) + \frac{1}{2} \partial_y^- f(\mathcal{Z}_{t-}) (d\mathcal{X}_t - d(\mathcal{N}_t - t))$$

$$+ \frac{1}{2} (-\partial_x^- f) (\mathcal{Z}_{t-}) d\mathcal{Y}_t + \frac{1}{2} (-\partial_x^+ f) (\mathcal{Z}_{t-}) d\mathcal{Y}_t$$

$$= \partial_y f(\mathcal{Z}_{t-}) d\mathcal{X}_t - (\partial_x^0 f) (\mathcal{Z}_{t-}) d\mathcal{Y}_t$$

To summarize, let f be harmonic. We have defined N_t^f as the stochastic integral

$$N_t^f := f(\mathcal{Z}_0) + \int_0^t \partial_y f(\mathcal{Z}_{s_-}) d\mathcal{X}_s - \partial_x^0 f(\mathcal{Z}_{s_-}) d\mathcal{Y}_s.$$

It is now easy to estimate the quadratic covariation

$$d[M^f, N^f]_t = \partial_x^0 f(\mathcal{Z}_{t_-}) \partial_y f(\mathcal{Z}_{t_-}) d[\mathcal{X}, \mathcal{X}]_t - \partial_y f(\mathcal{Z}_{t_-}) \partial_x^0 f(\mathcal{Z}_{t_-}) d[\mathcal{Y}, \mathcal{Y}]_t$$

$$= \partial_x^0 f(\mathcal{Z}_{t_-}) \partial_y f(\mathcal{Z}_{t_-}) d\mathcal{N}_t^0,$$

where we note $\mathcal{N}_t^0 := \mathcal{N}_t - t$ the compensated Poisson process, that is a martingale process.

Notice that M_t^f and N_t^f are not orthogonal martingales with respect to the bracket multiplication $[\cdot, \cdot]$. However, recall that the angular bracket $\langle \cdot, \cdot \rangle$, also known as the conditional quadratic covariation (see Protter [21]), is the compensator of $[\cdot, \cdot]$. But since \mathcal{N}_t^0 is a martingale, we have for the angular bracket

$$d\langle M^f, N^f \rangle_t = 0,$$

that is the martingales M_t^f and N_t^f are orthogonal with respect to the conditional quadratic covariation.

Similarly, the pairing of N_t^f with a test martingale M_t^g leads to the quadratic covariation

$$d[N^f, M^g]_t = \partial_y f(\mathcal{Z}_{t_-}) \left[\partial_x^0 g(\mathcal{Z}_{t_-}) + (\Delta_t \mathcal{X}) \frac{1}{2} \partial_x^2 g(\mathcal{Z}_{t_-}) \right] d\mathcal{N}_t$$

$$-\partial_x^0 f(\mathcal{Z}_{t_-}) \partial_y g(\mathcal{Z}_{t_-}) dt$$

$$= \partial_y f(\mathcal{Z}_{t_-}) [(\Delta_t \mathcal{X})_+ (\partial_x^- g)(\mathcal{Z}_{t_-}) + (\Delta_t \mathcal{X})_- (\partial_x^- g)(\mathcal{Z}_{t_-})] d\mathcal{N}_t$$

$$-\partial_x^0 f(\mathcal{Z}_{t_-}) \partial_y g(\mathcal{Z}_{t_-}) dt,$$

and the conditional quadratic covariation

$$d\langle N^f, M^g \rangle_t = \partial_y f(\mathcal{Z}_{t-}) \left[(\partial_x^0 g)(\mathcal{Z}_{t-}) + (\Delta_t \mathcal{X}) \frac{1}{2} (\partial_x^2 g)(\mathcal{Z}_{t-}) \right] d\mathcal{N}_t$$

$$-\partial_x^0 f(\mathcal{Z}_{t-}) \partial_y g(\mathcal{Z}_{t-}) dt$$

$$= \partial_y f(\mathcal{Z}_{t-}) [(\Delta_t \mathcal{X})_+ (\partial_x^- g)(\mathcal{Z}_{t-}) + (\Delta_t \mathcal{X})_- (\partial_x^- g)(\mathcal{Z}_{t-})] d\mathcal{N}_t$$

$$-\partial_x^0 f(\mathcal{Z}_{t-}) \partial_y g(\mathcal{Z}_{t-}) dt$$

$$= \partial_y f(\mathcal{Z}_{t-}) \left[\frac{1}{2} \partial_x^- g(\mathcal{Z}_{t-}) + \frac{1}{2} \partial_x^- g(\mathcal{Z}_{t-}) \right] dt$$

$$-\partial_x^0 f(\mathcal{Z}_{t-}) \partial_y g(\mathcal{Z}_{t-}) dt$$

$$= \{\partial_y f(\mathcal{Z}_{t-}) \partial_x^0 f(\mathcal{Z}_{t-}) - \partial_x^0 f(\mathcal{Z}_{t-}) \partial_y g(\mathcal{Z}_{t-}) \} dt.$$

5. Proof of Theorem 1

It remains to prove the representation formula stated in Theorem 1. Equipped with the martingale representations of the previous sections, it suffices to follow the lines of Gundy-Varopoulos [10] and Arcozzi [1]. For that, let $(\mathcal{Z}_t)_{-\infty < t \leqslant 0}$ be the so-called background noise. Those are semidiscrete random walks starting at infinity in the upper-half space $\mathbb{Z} \times \mathbb{R}^+$ and stopped at time t=0 when reaching the boundary \mathbb{Z} . Let f defined on \mathbb{Z} , $M_t^f:=(P_tf)(\mathcal{Z}_t)$ the associated martingale, and N_t^f the corresponding martingale transform as defined previously. Finally, introduce the projection operator

$$\mathcal{T}f(x) := \mathbb{E}(N_0^f | \mathcal{Z}_0 = x).$$

It follows that for any test function g defined on \mathbb{Z} , and the associated martingale M_t^g , we have

$$(\mathcal{T}f,g)_{\ell^{2}(\mathbb{Z})} = \sum_{x} \mathcal{T}f(x)g(x) = \sum_{x} \mathbb{E}(N_{0}^{f}|\mathcal{Z}_{0} = x)g(x) = \sum_{x} \mathbb{E}(N_{0}^{f}|\mathcal{Z}_{0} = x)M_{0}^{g}$$

$$= \sum_{x} \mathbb{E}(N_{0}^{f}M_{0}^{g}|\mathcal{Z}_{0} = x) = \sum_{x} \mathbb{E}\left(\int_{-\infty}^{0} d[N^{f}, M^{g}]_{t} \mid \mathcal{Z}_{0} = x\right)$$

$$= \sum_{x} \mathbb{E}\left(\int_{-\infty}^{0} \{\partial_{y}f(\mathcal{Z}_{t_{-}})\partial_{x}^{0}f(\mathcal{Z}_{t_{-}}) - \partial_{x}^{0}f(\mathcal{Z}_{t_{-}})\partial_{y}g(\mathcal{Z}_{t_{-}})\}dt \mid \mathcal{Z}_{0} = x\right)$$

$$= \sum_{x} \int_{-\infty}^{0} \{\partial_{y}f(y, x)\partial_{x}^{0}f(y, x) - \partial_{x}^{0}f(y, x)\partial_{y}g(y, x)\}2ydy$$

$$= (\mathcal{H}f, g)$$

where we used the fact that M_0^g depends only on \mathcal{Z}_0 but not on the trajectory, where we used the formula of the previous section for the quadratic covariations, and finally the fact that the density of the background noise \mathcal{Z}_t in the upper half space is the same as in the continuous setting, equal to 2ydy (see [10]).

References

- [1] Nicola Arcozzi. Riesz transforms on spheres and compact Lie groups. ProQuest LLC, Ann Arbor, MI, 1995. Thesis (Ph.D.)—Washington University in St. Louis.
- [2] N. Arcozzi, K. Domelevo, and S. Petermichl. Second Order Riesz Transforms on Multiply-Connected Lie Groups and Processes with Jumps. *Potential Anal.*, 45(4):777-794, 2016.
- [3] K. Domelevo, A. Osekowski, and S. Petermichl. Various sharp estimates for semi-discrete Riesz transforms of the second order. ArXiv e-prints, 2017. https://arxiv.org/abs/1701.04106.
- [4] K. Domelevo and S. Petermichl. Sharp L^p estimates for discrete second order Riesz transforms. Adv. Math., 262:932–952, 2014.
- [5] R. J. Duffin. Basic properties of discrete analytic functions. Duke Math. J., 23:335–363, 1956.
- [6] Matts Essén. A superharmonic proof of the M. Riesz conjugate function theorem. Ark. Mat., 22(2):241–249, 1984.
- [7] Jacqueline Ferrand. Fonctions préharmoniques et fonctions préholomorphes. Bull. Sci. Math. (2), 68:152–180, 1944.
- [8] I. C. Gohberg and M. G. Krein. Theory and applications of Volterra operators in Hilbert space. Translated from the Russian by A. Feinstein. Translations of Mathematical Monographs, Vol. 24. American Mathematical Society, Providence, R.I., 1970.
- [9] Maru Guadie and Eugenia Malinnikova. Stability and regularization for determining sets of discrete Laplacian. *Inverse Problems*, 29(7):075018, 17, 2013.
- [10] Richard F. Gundy and Nicolas Th. Varopoulos. Les transformations de Riesz et les intégrales stochastiques. C. R. Acad. Sci. Paris Sér. A-B, 289(1):A13-A16, 1979.
- [11] H.A. Heilbronn, On discrete harmonic functions. Proc. Camb. Philos. Soc., 45:194-206, 1949.
- [12] Rufus Philip Isaacs. A finite difference function theory. Rev., Ser. A, Univ. Nac. Tucuman, 2:177-201, 1941.
- [13] David Jerison, Lionel Levine, and Scott Sheffield. Internal DLA and the Gaussian free field. Duke Math. J., 163(2):267–308, 2014.
- [14] Gabor Lippner and Dan Mangoubi. Harmonic functions on the lattice: Absolute monotonicity and propagation of smallness. Duke Math. J., 164(13):2577–2595, 10 2015.
- [15] Françoise Lust-Piquard. Dimension free estimates for discrete Riesz transforms on products of abelian groups. Adv. Math., 185(2):289–327, 2004.
- [16] V. Matsaev and M. Sodin. Distribution of Hilbert transforms of measures. Geom. Funct. Anal., 10(1):160–184, 2000.
- [17] Vladimir Matsaev. Volterra operators obtained from self-adjoint operators by perturbation. Dokl. Akad. Nauk SSSR, 139:810–813, 1961.
- [18] S. K. Pichorides. On the best values of the constants in the theorems of M. Riesz, Zygmund and Kolmogorov. Studia Math., 44:165–179. (errata insert), 1972. Collection of articles honoring the completion by Antoni Zygmund of 50 years of scientific activity, II.
- [19] Nicolas Privault. Stochastic analysis in discrete and continuous settings with normal martingales, volume 1982 of Lecture Notes in Mathematics. Springer-Verlag, Berlin, 2009.
- [20] Nicolas Privault. Stochastic finance. Chapman & Hall/CRC Financial Mathematics Series. CRC Press, Boca Raton, FL, 2014. An introduction with market examples.
- [21] Philip E. Protter. Stochastic integration and differential equations, volume 21 of Stochastic Modelling and Applied Probability. Springer-Verlag, Berlin, 2005. Second edition. Version 2.1, Corrected third printing.
- [22] Igor E. Verbitsky. Estimate of the norm of a function in a Hardy space in terms of the norms of its real and imaginary parts. Mat. Issled., (54):16–20, 164–165, 1980. English transl.: Amer. Math. Soc. Transl. (2) 124 (1984), 11-15.

DIP. DI MATEMATICA, UNIVERSITA BOLOÑA, BOLOÑA, ITALY E-mail address: nicola.arcozzi@unibo.it

INST. MATH. TOULOUSE, UNIVERSITÉ PAUL SABATIER, TOULOUSE, FRANCE E-mail address: komla.domelevo@math.univ-toulouse.fr

Inst. Math. Toulouse, Université Paul Sabatier, Toulouse, France $E\text{-}mail\ address$: stefanie.petermichl@math.univ-toulouse.fr