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Implications of a new sampling theorem for sparse signal reconstruction on the sphere

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arXiv:1110.6298 arXiv:1205.1013

Astronomical Data Analysis (ADA) VII :: Cargèse, Corsica :: May 2012

Sampling theorems

Sparse signal reconstruction

Observations of the cosmic microwave background (CMB)

Full-sky observations of the CMB ongoing.



(a) COBE (launched 1989)



(b) WMAP (launched 2001)



(c) Planck (launched 2009)

Each new experiment provides dramatic improvement in precision and resolution of observations.

(cobe 2 wmap movie)

(planck movie)

(d) COBE to WMAP [Credit: WMAP Science Team]

(e) Planck observing strategy [Credit: Planck Collaboration]

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Sampling theorem

Sparse signal reconstruction

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Observations of the cosmic microwave background (CMB)



Credit: Max Tegmark

Sampling theorem

Sparse signal reconstruction

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Summary

Observations on the sphere



Credit: Alec MacAndrew

Sampling theorems

Sparse signal reconstruction

Summary

Observations on the three-ball (solid sphere)



 Boris Leistedt & JDM (2012), Exact wavelets on the ball, submitted to IEEE Trans. Sig. Proc., arXiv:1205.0792.

Sparse signal reconstruction

Outline

Harmonic analysis on the sphere

Spherical harmonic transform

2 Sampling theorems on the sphere

- Driscoll & Healy sampling theorem
- McEwen & Wiaux sampling theorem
- Comparison

Sparse signal reconstruction on the sphere

- Sparse signal reconstruction
- TV inpainting
- Low-resolution simulations
- High-resolution simulations

Summary

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Summary

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Harmonic analysis	Sampling theorems	Sparse signal reconstruction
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Spherical harmonic transform

• The spherical harmonics are the eigenfunctions of the Laplacian on the sphere: $\Delta_{s^2} Y_{\ell m} = -\ell(\ell+1)Y_{\ell m}$.



 Any square integrable scalar function on the sphere *f* ∈ L²(S²) may be represented by its spherical harmonic expansion:

$$f(\theta,\varphi) = \sum_{\ell=0}^{\infty} \sum_{m=-\ell}^{\ell} f_{\ell m} Y_{\ell m}(\theta,\varphi) .$$

• The spherical harmonic coefficients are given by the usual projection onto each basis function:

$$f_{\ell m} = \langle f, Y_{\ell m} \rangle = \int_{\mathrm{S}^2} \, \mathrm{d}\Omega(\theta,\varphi) f(\theta,\varphi) \; Y^*_{\ell m}(\theta,\varphi) \; .$$

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Harmonic analysis ●O	Sampling theorems	Sparse signal reconstruction
Spherical harmonic	transform	

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Figure: Spherical harmonic functions (real and imaginary parts).

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Spherical harn	nonic transform		
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Harmonic analysis	Sampling theorems	Sparse signal reconstruction	Summary

• We consider signals on the sphere band-limited at *L*, that is signals such that $f_{\ell m} = 0, \forall \ell \geq L$ \Rightarrow summations may be truncated at L - 1:

$$f(\theta,\varphi) = \sum_{\ell=0}^{L-1} \sum_{m=-\ell}^{\ell} f_{\ell m} Y_{\ell m}(\theta,\varphi) .$$

• For a band-limited signal, can we compute $f_{\ell m}$ exactly?

 \rightarrow Sampling theorems on the sphere.

• Aside: Generalise to spin functions on the sphere.

Square integrable spin functions on the sphere $sf \in L^2(S^2)$, with integer spin $s \in \mathbb{Z}$, are defined by their behaviour under local rotations. By definition, a spin function transforms as

$$_{s}f'(\theta,\varphi) = \mathrm{e}^{-\mathrm{i}s\chi} {}_{s}f(\theta,\varphi)$$

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under a local rotation by χ , where the prime denotes the rotated function.

Spherical harm	nonic transform		
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Harmonic analysis	Sampling theorems	Sparse signal reconstruction	Summary
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Sparse signal reconstruction

Outline



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Summary

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Sampling theorems

Sparse signal reconstruction

Summary O

Driscoll & Healy (DH) sampling theorem

- Canonical sampling theorem on the sphere derived by Driscoll & Healy (1994) for equiangular grids.
- Gives an explicit quadrature rule for the spherical harmonic transform:

$$f_{\ell m} = \sum_{t=0}^{2L-1} \sum_{p=0}^{2L-1} q_{\mathrm{DH}}(\theta_t) f(\theta_t, \varphi_p) Y_{\ell m}^*(\theta_t, \varphi_p) ,$$

where the sample positions are defined by $\theta_t = \pi t/2L$, for t = 0, ..., 2L - 1, and $\varphi_p = \pi p/L$, for p = 0, ..., 2L - 1

 \Rightarrow $N_{\rm DH} = (2L - 1)2L + 1 \sim 4L^2$ samples on the sphere.

• The quadrature weights are defined implicitly by the solution to

$$\sum_{t=0}^{2L-1} q_{\rm DH}(\theta_t) \, P_{\ell}(\cos \theta_t) = \frac{2\pi}{L} \, \delta_{\ell 0} \; , \quad \forall \ell < 2L$$

and are given explicitly by

$$q_{\rm DH}(\theta_t) = \frac{2\pi}{L^2} \sin \theta_t \sum_{k=0}^{L-1} \frac{\sin((2k+1)\theta_t)}{2k+1} \; .$$

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Sampling theorems

Sparse signal reconstruction

Summary O

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McEwen & Wiaux (MW) sampling theorem

- A new sampling theorem (with fast algorithms) has emerged very recently by performing a
 factoring of rotations and then by associating the sphere with the torus through a periodic
 extension.
- Similar to making a periodic extension in θ of a function f on the sphere.

Sampling theorems

Sparse signal reconstruction

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McEwen & Wiaux (MW) sampling theorem

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 factoring of rotations and then by associating the sphere with the torus through a periodic
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- Similar to making a periodic extension in θ of a function f on the sphere.



(a) Function on sphere



(b) Even function on torus



(c) Odd function on torus

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Figure: Associating functions on the sphere and torus

Sampling theorems

Sparse signal reconstruction

Summary O

McEwen & Wiaux (MW) sampling theorem

 By a factoring of rotations, a reordering of summations and a separation of variables, the inverse transform of *s* may be written:



where $\Delta_{mn}^{\ell} \equiv d_{mn}^{\ell}(\pi/2)$ are the reduced Wigner functions evaluated at $\pi/2$.

Sampling theorems

Sparse signal reconstruction

Summary O

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McEwen & Wiaux (MW) sampling theorem

 By a factoring of rotations, a reordering of summations and a separation of variables, the forward transform of sf may be written:

Forward spherical harmonic transform

$$f_{\ell m} = (-1)^{s} i^{m+s} \sqrt{\frac{2\ell+1}{4\pi}} \sum_{m'=-(L-1)}^{L-1} \Delta_{m'm}^{\ell} \Delta_{m',-s}^{\ell} G_{mm'}$$

$${}_{s}G_{mm'} = \int_{0}^{\pi} \mathrm{d}\theta \sin\theta {}_{s}G_{m}(\theta) \mathrm{e}^{-\mathrm{i}m'\theta}$$

$${}_{s}G_{m}(\theta) = \int_{0}^{2\pi} \mathrm{d}\varphi \, {}_{s}f(\theta,\varphi) \, \mathrm{e}^{-\mathrm{i}m\varphi}$$

- JDM (2011a), Fast, exact (but unstable) spin spherical harmonic transforms
- Huffenberger & Wandelt (2010), Fast and exact spin-s spherical harmonic transforms
- JDM & Wiaux (2011b), A novel sampling theorem on the sphere

 \Rightarrow $N_{\text{MW}} = (L-1)(2L-1) + 1 \sim 2L^2$ samples on the sphere.

Sampling theorems

Sparse signal reconstruction

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Sparse signal reconstruction

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Comparison



Figure: Number of samples (MW=red; DH=green; GL=blue)

Harmonic	analysis

Sparse signal reconstruction

Comparison



Figure: Numerical accuracy (MW=red; DH=green; GL=blue)

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Harmonic	analysis

Sparse signal reconstruction

Comparison



Figure: Computation time (MW=red; DH=green; GL=blue)

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Comparison

	DH Divide-and-conquer	DH Semi-naive	MW
Pixelisation scheme	equiangular	equiangular	equiangular
Asymptotic complexity	$\mathcal{O}(L^{5/2}\log_2^{1/2}L)$	$\mathcal{O}(L^3)$	$\mathcal{O}(L^3)$
Precomputation	Y	Ν	Ν
Stability	Ν	Y	Y
Flexibility of Wigner recursion	Ν	Ν	Y
Spin functions	Ν	Ν	Υ
Number of samples	$4L^2$	$4L^2$	2L ²

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Sparse signal reconstruction on the sphere

- A reduction in the number of samples required to represent a band-limited signal on the sphere has important implications for sparse signal reconstruction.
- Many natural signals are sparse in a spatially localised measure, such as in a wavelet basis, overcomplete dictionary, or in the magnitude of their gradient, for example.
- A more efficient sampling of a band-limited signal on the sphere improves both the dimensionality and sparsity of the signal in the spatial domain.
- For a given number of measurements, a more efficient sampling theorem improves the fidelity of sparse signal reconstruction.
- We develop a framework for total variation (TV) inpainting on the sphere to demonstrate this
 result.

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- We develop a framework for total variation (TV) inpainting on the sphere to demonstrate this
 result.

Harmonic analysis	Sampling theorems	Sparse signal reconstruction	Summar O
TV inpainting			

- Consider inpainting problem $y = \Phi x + n$ in the context of different sampling theorems, where:
 - the samples of *f* are denoted by the concatenated vector $x \in \mathbb{R}^N$;
 - N is the number of samples on the sphere of the chosen sampling theorem;
 - *M* noisy measurements $y \in \mathbb{R}^M$ are acquired;
 - the measurement operator $\Phi \in \mathbb{R}^{M \times N}$ represents a random masking of the signal;
 - the noise $n \in \mathbb{R}^M$ is assumed to be iid Gaussian with zero mean.

• Define TV norm on the sphere:



• TV inpainting problem solved directly on the sphere:

$$x^{\star} = \operatorname*{arg\,min}_{x} \|x\|_{\mathrm{TV}}$$
 such that $\|y - \Phi x\|_{2} \leq \epsilon$.

• TV inpainting problem solved in harmonic space:

$$\hat{x}^{\prime \star} = \operatorname*{arg\,min}_{\hat{x}^{\prime}} \| \Lambda' \hat{x}^{\prime} \|_{\mathrm{TV}} \text{ such that } \| y - \Phi \Lambda' \hat{x}^{\prime} \|_{2} \leq \epsilon \; ,$$

where Λ' represents the inverse spherical harmonic transform (while also including a conjugate symmetry extension to impose reality) and harmonic coefficients are represented by the concatenated vector $\hat{\mathbf{x}}' \in \mathbb{C}^{L(L+1)/2}$.

Harmonic analysis 00	Sampling theorems	Sparse signal reconstruction	Summary O
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$$\int_{\mathbb{S}^2} \mathrm{d}\Omega |\nabla f| \simeq \sum_{t=0}^{N_\theta - 1} \sum_{p=0}^{N_\varphi - 1} |\nabla f| q(\theta_t) \simeq \sum_{t=0}^{N_\theta - 1} \sum_{p=0}^{N_\varphi - 1} \sqrt{q^2(\theta_t) (\delta_\theta \mathbf{x})^2 + \frac{q^2(\theta_t)}{\sin^2 \theta_t} (\delta_\varphi \mathbf{x})^2} \equiv \|\mathbf{x}\|_{\mathrm{TV}}.$$

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Harmonic analysis 00	Sampling theorems	Sparse signal reconstruction	Summary O
TV inpainting			

- Consider inpainting problem $y = \Phi x + n$ in the context of different sampling theorems, where:
 - the samples of *f* are denoted by the concatenated vector $x \in \mathbb{R}^N$;
 - N is the number of samples on the sphere of the chosen sampling theorem;
 - *M* noisy measurements $y \in \mathbb{R}^M$ are acquired;
 - the measurement operator $\Phi \in \mathbb{R}^{M \times N}$ represents a random masking of the signal;
 - the noise $n \in \mathbb{R}^M$ is assumed to be iid Gaussian with zero mean.

Define TV norm on the sphere:

$$\int_{\mathbb{S}^2} \mathrm{d}\Omega |\nabla f| \simeq \sum_{t=0}^{N_\theta - 1} \sum_{p=0}^{N_\varphi - 1} |\nabla f| q(\theta_t) \simeq \sum_{t=0}^{N_\theta - 1} \sum_{p=0}^{N_\varphi - 1} \sqrt{q^2(\theta_t) (\delta_\theta \mathbf{x})^2 + \frac{q^2(\theta_t)}{\sin^2 \theta_t} (\delta_\varphi \mathbf{x})^2} \equiv \|\mathbf{x}\|_{\mathrm{TV}}.$$

• TV inpainting problem solved directly on the sphere:

$$oldsymbol{x}^\star = rgmin_{oldsymbol{x}} \|oldsymbol{x}\|_{ ext{TV}}$$
 such that $\|oldsymbol{y} - \Phi oldsymbol{x}\|_2 \leq \epsilon$.

TV inpainting problem solved in harmonic space:

$$\begin{split} \hat{\pmb{x}}'^{\star} &= \argmin_{\hat{\pmb{x}}'} \|\Lambda' \hat{\pmb{x}}' \|_{\mathrm{TV}} \; \text{ such that } \; \|\pmb{y} - \Phi \Lambda' \hat{\pmb{x}}' \|_2 \leq \epsilon \; , \end{split}$$

where Λ' represents the inverse spherical harmonic transform (while also including a conjugate symmetry extension to impose reality) and harmonic coefficients are represented by the concatenated vector $\hat{\mathbf{x}}' \in \mathbb{C}^{L(L+1)/2}$.

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TV inpainting: low-resolution simulations

• Solve TV inpainting problem on the sphere in the context of the Driscoll & Healy sampling theorem and our new sampling theorem (at *L* = 32).



Figure: Earth topographic data reconstructed in the harmonic domain for $M/L^2 = 1/2$

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TV inpainting: low-resolution simulations



Figure: Reconstruction performance for the DH and MW sampling theorems

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TV inpainting: high-resolution simulations

- Previously limited to low-resolution simulations.
- To solve high-resolution problem we require fast adjoint spherical harmonic transform operators in addition to fast forward spherical harmonic transforms to solve optimisation problems.
- Superiority of new sampling theorem clear, hence develop fast adjoints for this case only.

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Fast adjoint inverse spherical harmonic transform

$$\tilde{f}^{\dagger}(\theta_t, \varphi_p) = \begin{cases} sf(\theta_t, \varphi_p) , & t \in \{0, 1, \dots, L-1\} \\ 0, & t \in \{L, \dots, 2L-2\} \end{cases}$$

$${}_{s}F_{mm'}^{\dagger} = \sum_{l=0}^{2L-2} \sum_{p=0}^{2L-2} {}_{s}\tilde{f}^{\dagger}(\theta_{l},\varphi_{p}) e^{-i(m'\theta_{l}+m\varphi_{p})}$$

$$f_{\ell m}^{\dagger} = (-1)^{s} i^{m+s} \sqrt{\frac{2\ell+1}{4\pi}} \sum_{m'=-(L-1)}^{L-1} \Delta_{m'm}^{\ell} \Delta_{m',-s}^{\ell} {}^{s} F_{mm'}^{\dagger}$$

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TV inpainting: high-resolution simulations

Fast adjoint forward spherical harmonic transform

$${}_{s}G_{mm'}^{\dagger} = (-1)^{s} \operatorname{i}^{-(m+s)} \sum_{\ell=0}^{L-1} \sqrt{\frac{2\ell+1}{4\pi}} \Delta_{m'm}^{\ell} \Delta_{m',-s}^{\ell} \mathfrak{s}_{\ell m}^{s}$$

$${}_{s}F_{mm''}^{\dagger} = 2\pi \sum_{m'=-(L-1)}^{L-1} {}_{s}G_{mm'}^{\dagger} w(m'-m'')$$

$${}_{s}\tilde{F}_{m}^{\dagger}(\theta_{t}) = \frac{1}{2L-1} \sum_{m'=-(L-1)}^{L-1} {}_{s}F_{mm'}^{\dagger} e^{im'\theta_{t}}$$

$${}_{s}F_{m}^{\dagger}(\theta_{t}) = \begin{cases} s\tilde{F}_{m}^{\dagger}(\theta_{t}) + (-1)^{m+s} s\tilde{F}_{m}^{\dagger}(\theta_{2L-2-t}), & t \in \{0, 1, \dots, L-2\}\\ s\tilde{F}_{m}^{\dagger}(\theta_{t}), & t = L-1 \end{cases}$$

$$_{s}f^{\dagger}(\theta_{t},\varphi_{p})=rac{1}{2L-1}\sum_{m=-(L-1)}^{L-1}{}_{s}F_{m}^{\dagger}(\theta_{t})e^{im\varphi_{p}}$$

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TV inpainting: high-resolution simulations



Figure: Ground truth (L = 128)

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TV inpainting: high-resolution simulations



Figure: Measurements ($M/L^2 = 1/4$; L = 128)

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Figure: Reconstruction $(M/L^2 = 1/4; L = 128; SNR = 29 dB)$

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Outline

Harmonic analysis on the spher

• Spherical harmonic transform

Sampling theorems on the sphere

- Driscoll & Healy sampling theorem
- McEwen & Wiaux sampling theorem
- Comparison

Sparse signal reconstruction on the sphere

- Sparse signal reconstruction
- TV inpainting
- Low-resolution simulations
- High-resolution simulations

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Summary

- New MW sampling theorem on the sphere which captures all of the information content of a band-limited signal in only 2L² samples (compared to 4L² for the DH sampling theorem).
- A reduction in the number of samples required to represent a band-limited signal on the sphere has important implications for sparse signal reconstruction.
- For signals sparse in a spatially localised representation, a more efficient sampling of the sphere improves the fidelity of sparse signal reconstruction.
- We develop a framework for total variation (TV) inpainting on the sphere to demonstrate this result → superiority of the MW sampling theorem for sparse signal reconstruction clear.
- Develop fast adjoint spherical harmonic transforms for the MW sampling theorem to solve sparse signal reconstruction problems on the sphere at high-resolution.

Papers

- McEwen & Wiaux, A novel sampling theorem on the sphere, IEEE Trans. Sig. Proc., 59, 12, 5876–5887, arXiv:1110.6298, 2011.
- McEwen, Puy, Thiran, Vandergheynst, Van De Ville & Wiaux, Sparse signal reconstruction on the sphere: implications of a new sampling theorem, IEEE Trans. Sig. Proc., submitted, arXiv:1205.1013, 2012.

Code

SSHT Code to compute fast and exact, forward and adjoint (spin) spherical harmonic transforms based on the MW sampling theorem (Fortran, C, Matlab)

Available under the GPL from http://www.ssht.org.uk/

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