

The multivariate anisotropic Wavelet Transform on the Torus

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Introduction

Wavelets: fast decomposition of a function or signal into "levels of detail". here: periodic function(s)

one-dimensional

- shift invariant spaces based on the shift $2\pi/N$, $N \in \mathbb{N}$
- periodic wavelets, e.g. in [PT95]
- characterizations and fast algorithm in freuency domain

d-dimensional

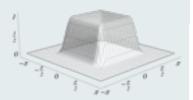
- "curse of dimension"
- reduced by focusing on certain directions [LP10, MS03]
- develop corresponding wavelets towards adaptivity/focus
- adapt characterizations and fast algorithms in frequency domain
- + characterize directions



Sampling

using patterns to specify certain directions of interest

Let's say our data is given by certain measurements of a 2π periodic d-dimensional function $f: \mathbb{T}^d \to \mathbb{R}$ (or \mathbb{C}),



Example: a simple "hill" (d = 2)

more preciesly by $2\pi \mathbf{y}, \mathbf{y} \in \mathcal{P}(\mathbf{M})$, where $\mathbf{M} \in \mathbb{Z}^{d \times d}$ is invertible and

$$\mathcal{P}(\mathbf{M}) \coloneqq \mathbf{M}^{-1} \mathbb{Z}^d \cap \left[-\frac{1}{2}, \frac{1}{2} \right]^d$$

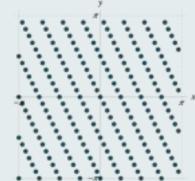
- specify "directions of interest" (columns of M⁻¹, the inverse of M)
- $m = |\mathcal{P}(\mathbf{M})| = |\det \mathbf{M}|$ number of points
- How to get information about "detail" in one of these directions?
 Idea: More points in directions of interest



The pattern $\mathcal{P}(\mathbf{M})$

Sampling with interest in a specific direction

$$\mathbf{M} = \begin{pmatrix} 32 & -8 \\ 8 & 4 \end{pmatrix}$$





The pattern corresp. to M

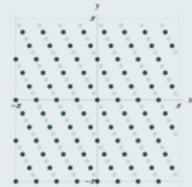
drawn on a torus



The pattern $\mathcal{P}(\mathbf{N})$

Sampling with less interest in a specific direction

$$\mathbf{M} = \mathbf{J}\mathbf{N} = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 16 & -4 \\ 8 & 4 \end{pmatrix}$$





The pattern corresp. to N

drawn on a torus

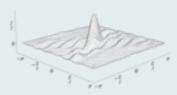
Shift Invariant spaces

Approximate f by shifts of one function.

Given a second function φ : Approximate f by a sum of translates of φ

$$f(\mathbf{x}) \approx f_{\mathbf{M}}(\mathbf{x}) = \sum_{\mathbf{y} \in \mathcal{P}(\mathbf{M})} a_{\mathbf{y}} \varphi(\mathbf{x} - 2\pi \mathbf{y})$$

Set of $a_y \in \mathbb{C}$: "a building plan" for f.



The Dirichlet kernel $\varphi = D_N$ [LP10]

necessary:
$$\{\varphi(\circ - 2\pi \mathbf{y})\}_{\mathbf{y} \in \mathcal{P}(\mathbf{M})}$$
 are linear independent

The space of all possible funktions of these translates is called the *shift* invariant space of φ with respect to \mathbf{M} , denoted by

$$V_{\mathbf{M}}^{\varphi} \coloneqq \operatorname{span}\{\varphi(\circ - 2\pi\mathbf{y}), \ \mathbf{y} \in \mathcal{P}(\mathbf{M})\}.$$



Decomposition

Looking at M = JN, where N is (still) integer valued and (for simplicity) $|\det J| = 2$ \Rightarrow decompose

$$V_{\mathbf{M}}^{\varphi} = V_{\mathbf{N}}^{\xi} \oplus W_{\mathbf{N}}^{\psi},$$

where

- construct $\xi \in V_{\mathbf{M}}^{\varphi}$ (analogously to φ before) \Rightarrow compute ψ (the wavelet)
- ullet both $V_{f N}^{\xi}$ and $W_{f N}^{\psi}$ are both shift invariant (w.r.t. f N) spaces

Hence, we have a **unique** decomposition for $f_{\mathbf{M}} \in V_{\mathbf{M}}$:

$$f_{\mathbf{M}} = f_{\mathbf{N}} + g_{\mathbf{N}}, \quad f_{\mathbf{N}} \in V_{\mathbf{N}}^{\xi}, \quad g_{\mathbf{N}} \in W_{\mathbf{N}}^{\psi},$$

where

- f_N is a (depending on J) certain rougher approximation
- q_N is the difference of both approximations \Rightarrow detail



The (fast) Fourier transform on $\mathcal{P}(\mathbf{M})$

The Fourier matrix with respect to M

$$\mathcal{F}(\mathbf{M}) \coloneqq \frac{1}{\sqrt{m}} \left(e^{-2\pi i \mathbf{h}^T \mathbf{y}} \right)_{\mathbf{h} \in \mathcal{G}(\mathbf{M}^T), \mathbf{y} \in \mathcal{P}(\mathbf{M})} \in \mathbb{C}^{m \times m}$$

where

- $\mathbf{h} \in \mathcal{G}(\mathbf{M}^T) := \mathbf{M}^T \mathcal{P}(\mathbf{M}^T)$ addresses the rows
- $\mathbf{y} \in \mathcal{P}(\mathbf{M})$ the columns
- $\mathbf{a} = (a_{\mathbf{y}})_{\mathbf{y} \in \mathcal{P}(\mathbf{M})} \in \mathbb{C}^m$ (same order als the colums): DFT on $\mathcal{P}(\mathbf{M})$

$$\hat{\mathbf{a}} = (\hat{a}_{\mathbf{h}})_{\mathbf{h} \in \mathcal{G}(\mathbf{M}^T)} = \sqrt{m} \mathcal{F}(\mathbf{M}) \mathbf{a} \in \mathbb{C}^m$$

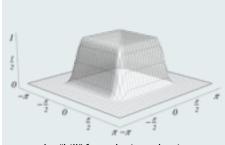
- \Rightarrow fast Fourier transform for special ordering [Be13], i.e. in $O(m \log m)$
 - Sample f on $\mathcal{P}(\mathbf{M})$ to compute $f_{\mathbf{M}} \in V_{\mathbf{M}}^{\varphi}$ (in Fourier domain)
- \Rightarrow decomposition (using Parseval equality on $L_2(\mathbb{T}^d)$) in O(m)



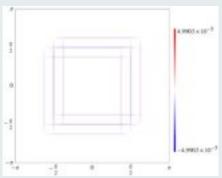
Example

search for changes of steepness in the "hill" - sampling

- $\mathbf{M} = \begin{pmatrix} 512 & 0 \\ 0 & 512 \end{pmatrix} \Rightarrow$ "classical" sampling
- $\varphi = D_{\mathbf{M}}, \xi = D_{\mathbf{N}}$, the Dirichlet kernels [LP10]



the "hill" from the introduction a Box spline



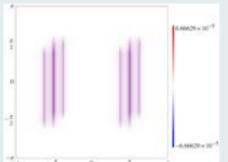
the error $f - f_{\mathbf{M}}$



Example

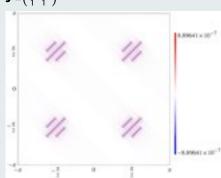
search for changes of steepness in the "hill" - decompose $f_{\mathbf{M}}$

$$\mathbf{J} = \left(\begin{smallmatrix} 2 & 0 \\ 0 & 1 \end{smallmatrix} \right)$$



the decomposition in order to detect in horizontal direction

$$\mathbf{J} = \left(\begin{array}{cc} 1 & -1 \\ 1 & 1 \end{array} \right)$$



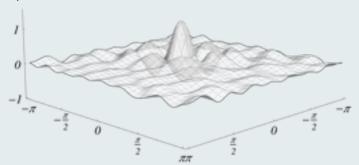
the decomposition in order to detect in both diagonal directions



Current work

The Wavelet should be

- localised
- still easy to compute (in Fourier coefficients)
- only dependent on J



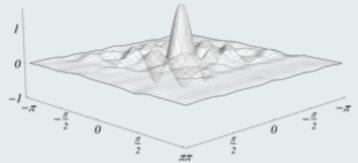
Dirichlet kernel Wavelet



Current work

The Wavelet should be

- localised
- still easy to compute (in Fourier coefficients)
- only dependent on J



A more localized wavelet "for the same directions"



Summary

In order to obtain directional information about (sampled data of) f

- Choose φ, ξ (depending on "direction" **J**), such that
 - they fulfill the requirements (w.r.t. M resp. N)
 - their translates approximate f well (at least w.r.t. certain criteria)
 - neccessary: $\xi \in V_{\mathbf{M}}^{\varphi}$
 - \Rightarrow translates of wavelet ψ model the detail
- further decomposions of N ⇒ Multiscale Analysis
- using the fact, that $f_{\mathbf{M}}, \xi, \psi \in V_{\mathbf{M}}^{\varphi}$
 - \Rightarrow all computations possible by using " $\{a_{\mathbf{y}}\}_{\mathbf{y}\in\mathcal{P}(\mathbf{M})}$ -type" data
 - ⇒ using their Fourier series and the discrete FourierTransforms
 - \Rightarrow fast algorithms in O(m)



Thanks for your attention.

...and of course the organizers for this nice workshop.

Literature

- [Be13] rb, The fast Fourier Transform and fast Wavelet Transform for Patterns on the Torus, to appear in ACHA (2013). doi: 10.1016/j.acha.2012.07.007 arXiv: http://arxiv.org/abs/1107.5415
- [LP10] D. Langemann, J. Prestin, *Multivariate periodic wavelet analysis*, ACHA 28 (2010) 46–66. doi: 10.1016/j.acha.2009.07.001
- [MS03] I. E. Maximenko, M. A. Skopina, Multivariate periodic wavelets, St. Petersbg. Math. J. 15 (2003) 165–190. doi: 10.1090/S1061-0022-04-00808-8
- [PT95] G. Plonka, M. Tasche, On the computation of periodic spline wavelets, ACHA 2 (1995) 1–14. doi: 10.1006/acha.1995.1001