# Scattered Data Approximation on Submanifolds

(Combined) Ambient Solutions

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Bernried, 27.02.-03.03.2017

## Part I: Sparse Scattered Data

## Part II: Regular well-sampled Scattered Data

## Part III: Irregular Scattered Data

#### PART I: SPARSE SCATTERED DATA

#### **PROBLEM STATEMENT**

Setting:

▶  $\mathbb{M} \subseteq \mathbb{R}^d$  an embedded submanifold with  $q = \dim \mathbb{M} < 4$ 

◊ closed 
◊ compact 
◊ without boundary

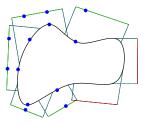
- $\blacktriangleright \Xi \subseteq \mathbb{M}$  a set of **sparse** sites scattered over  $\mathbb{M}$
- ▶  $\Upsilon$  function values to  $\Xi$

Task: Determine a »reasonable« function  $f : \mathbb{M} \to \mathbb{R}$  such that

 $f(\xi) = y_{\xi}$  for all  $\xi \in \Xi$  and corresponding  $y_{\xi} \in \Upsilon$ 

#### COMMON APPROACHES: CHARTS AND BLENDING

- ► Use charts and define function spaces there to solve problem locally.
- ► Blend local solutions to obtain global solution.



Problems:

- There might be charts without any sites
- Blending tends to produce undesireable »gluing breaks«

## COMMON APPROACHES: INTRINSIC FUNCTIONS

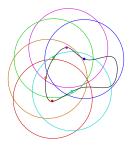
- ► Determine purely intrinsic function spaces  $\mathcal{F}_{\mathbb{M}}$  like spherical harmonics on  $\mathbb{S}^q$ .
- ► Use these for a solution

#### Problems:

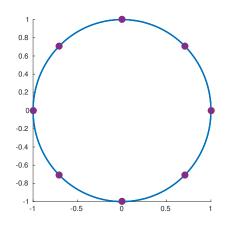
- These spaces are  $\mathbb{M}$ -specific and for arbitrary  $\mathbb{M}$  hard to determine.
- Also, they are often costly to evaluate.

#### COMMON APPROACHES: EXTRINSIC DIRECT INTERPOLATION

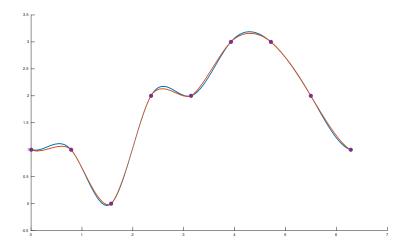
- ► Use some function space in a suitable ambient neighbourhood of M: RBF, Splines...
- ► Solve standard interpolation problem in neighbourhood, ignore the geometry of M.
- $\blacktriangleright$  Restrict the solution to  $\mathbb{M}.$



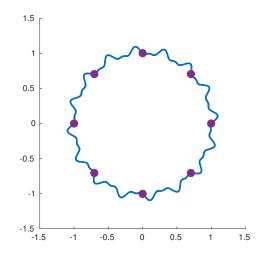
Directly applicable and works always — but how good?



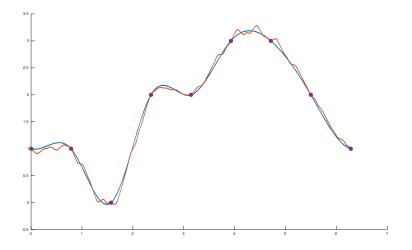
Data sites with function values: 1, 1, 0, 2, 2, 3, 3, 2



Periodic cubic spline Restricted thin-plate spline Interpolation sites



Data sites with function values: 1, 1, 0, 2, 2, 3, 3, 2



Periodic cubic spline interpolant

#### Restricted thin-plate spline interpolant

## COMMON APPROACHES: EXTRINSIC INTERPOLATION

- ► Use some function space in a suitable ambient neighbourhood of M: RBF, Splines...
- ► Solve standard interpolation problem in neighbourhood, ignore the geometry of M.
- ► Restrict the solution to M.

#### Problems:

- Difficulties occur for sparse data.
- Suffers extremely from intricate geometries.

## NEW APPROACH (M., REIF)

- $\blacktriangleright$  Use some function space in a suitable ambient neighbourhood of  $\mathbb{M}:$  RBF, Splines...
- ► Transfer intrinsic properties into extrinsic (ambient) properties approximately.
- ► Solve approximately intrinsic problem with extrinsic methods.

#### Pro's:

- Extrinsic function spaces are well understood.
- Extrinsic function space are applicable to any submanifold.
- Easily understood and implemented even for non-mathematicians.

DEFINITION (Tangent Derivative)

Let  $f : \mathbb{M} \to \mathbb{R}$  be a sufficiently (weakly) differentiable function and  $\tilde{f}$  an extension into  $U(\mathbb{M})$ . The **Tangent Derivative Operator**  $\mathbf{d}_{\mathbb{M}}$  is defined as

$$\mathbf{d}_{\mathbb{M}}f := \mathbf{d}\tilde{f} - \pi_{N}(\mathbf{d}\tilde{f})$$

and independent of the choice of  $\tilde{f}$ .

THEOREM (e.g. [Dziuk/Elliot, 2013])

Let  $f \in C^2(\mathbb{M})$ , and  $\overline{f}$  an extension that is **constant in normal directions** of  $\mathbb{M}$ . Then the 1<sup>st</sup> and 2<sup>nd</sup> **tangent** derivatives of f <u>coincide</u> with the **euclidean** 1<sup>st</sup> and 2<sup>nd</sup> derivatives of  $\overline{f}$  on the tangent space.

**ТНЕОВЕМ** (М. 2015)

Let  $f \in C^2(\mathbb{M})$ , and  $\tilde{f}$  an arbitrary extension. Then the deviations of the 1<sup>st</sup> and 2<sup>nd</sup> tangent derivatives of f from the **euclidean** 1<sup>st</sup> and 2<sup>nd</sup> derivatives of  $\tilde{f}$  on the tangent space are Lipschitz functions of the first normal derivatives.

What does that mean?

- ► Euclidean derivatives of *t̃* give **approximate** access to tangent derivatives of *f*
- Standard methods can be used to handle intrinsic problems approximately
- ► Intrinsic functionals are easily **approximated** by standard functionals

## NEW APPROACH (M., REIF): SOLUTION IDEA

Let  $U(\mathbb{M})$  be a suitable neighbourhood of  $\mathbb{M}$ , such that any point there has a unique closest point on  $\mathbb{M}$ .

Consider a suitable function space  $\mathcal{F}(U(\mathbb{M}))$  on  $U(\mathbb{M})$ .

Minimize squared  $2^{\textit{nd}}$  derivative in tangent directions over  $\mathbb M$ 

such that:

Interpolation in  $\Xi$  holds

Normal derivatives  $\rightarrow 0$ 

#### **OPTIMIZATION FUNCTIONAL: TENSOR-PRODUCT B-SPLINES**

Minimize for grid width h > 0 and  $\tau_1, ..., \tau_q$  an ONB of  $T_p(\mathbb{M}), \nu_1, ..., \nu_{d-q}$  ONB of  $N_p(\mathbb{M})$ 

$$\int_{\mathbb{M}} \sum_{i,j=1}^{q} \left| \frac{\partial^{2}}{\partial \tau_{i} \partial \tau_{j}} s_{h} \right|^{2} + Ch^{-\sigma} \int_{\mathbb{M}} \sum_{i=1}^{d-q} \left| \frac{\partial}{\partial \nu_{i}} s_{h} \right|^{2} + \int_{U_{h}(\mathbb{M})} w(h^{-\sigma}, N_{x}) \sum_{i=1}^{d-q} \left| \frac{\partial}{\partial \nu_{i}(clp(\cdot))} s_{h} \right|^{2}$$

for suitable penalty parameter  $\sigma$  and radial weight function w, under side conditions

$$s_h(\xi) = y_{\xi} \quad \forall \xi \in \Xi$$

#### THEORY: SOLVABILITY AND CONVERGENCE

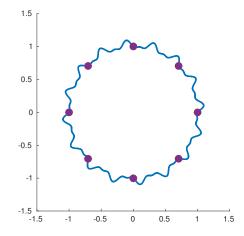
ТНЕОВЕМ (М. 2015):

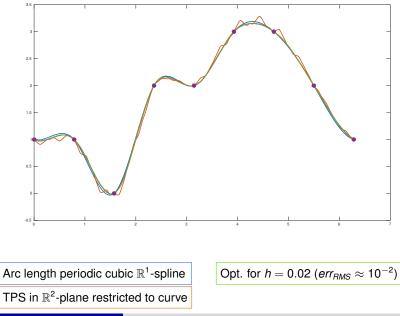
1. For TP-B-Splines and sufficiently small *h*, the above problem is uniquely solvable.

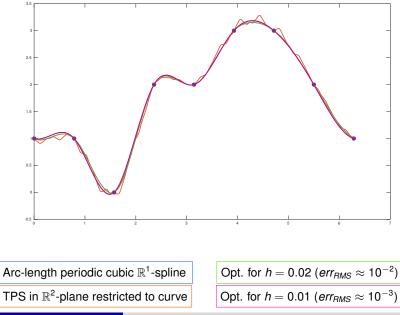
2. For  $h \to 0$  the squared second derivative energy converges to the unique optimal energy in  $\mathcal{H}^2(\mathbb{M})$ .

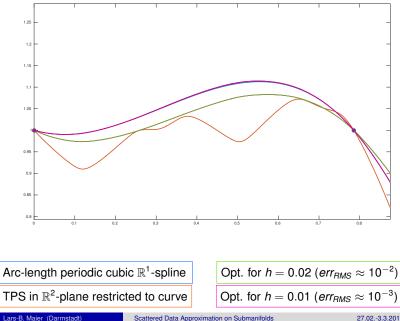
3. Restrictions of optimal splines  $s_h|_{\mathbb{M}}$  approach unique optimum  $f^* \in \mathcal{H}^2(\mathbb{M})$ :

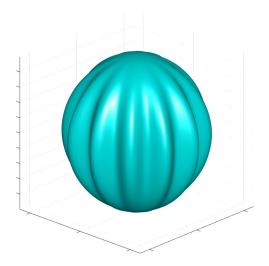
$$\|s_h - f^*\|_{\mathcal{H}^2(\mathbb{M})} o 0 \quad ext{ as } \quad h o 0.$$





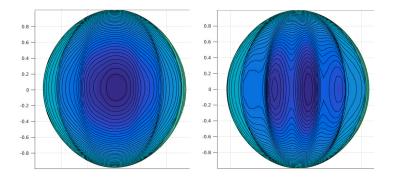






Sites with corresponding interpolation values:

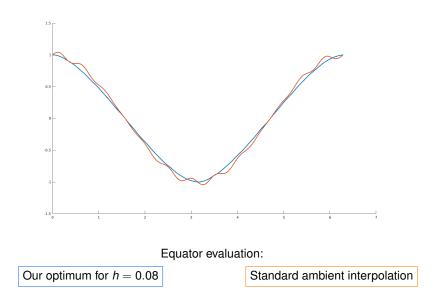
$$s(\pm e_1) = \pm 1, \; s(\pm e_2) = 0, \; s(\pm e_3) = 0$$



#### Reprojections onto $\mathbb{S}^2$ :

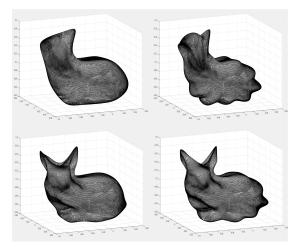
Our optimum for h = 0.08

Standard ambient interpolation



#### Standard ambient interpolation

#### Our optimum for h = 0.0625



Stanford bunny as an image of the pumpkin, using  $\approx$  80 (upper row) and  $\approx$  160 sites (lower row)

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Our optimum for h = 0.0625

#### Standard ambient interpolation

Stanford bunny as an image of the pumpkin, using  $\approx 350$  (upper row) and  $\approx 700$  sites (lower row)

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#### FURTHER REMARKS

▶ With fill distance  $h_{\Xi,M} = \max_{p \in M} \min_{\xi \in \Xi} \|p - \xi\|_2$  decreasing, the convergence is about that of thin-plate splines

But: One will have to increase the number of DOF correspondingly to meet

- More interpolation conditions
- Sufficient decrease of derivatives in normal directions

► One can use RBF or other approaches instead of TPBS — but their locality makes them a very good choice.

► Application for M with boundary is possible, but requires reasonable boundary conditions (ongoing research).

 $\blacktriangleright$  Similar approaches work for **Smoothing** and **Elliptical PDEs** on M.

## PART II: WELL SAMPLED REGULAR SCATTERED DATA

#### **PROBLEM STATEMENT**

#### Setting:

 $\blacktriangleright$   $\mathbb{M} \subseteq \mathbb{R}^d$  an embedded submanifold

◊ closed 
◊ compact 
◊ without boundary

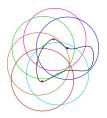
- $\blacktriangleright \ \Xi \subseteq \mathbb{M}$  a set of quasi-uniform sites scattered over  $\mathbb{M}, \ |\Xi|$  rather large
- ▶  $\Upsilon$  function values to  $\Xi$

Task: Determine a function  $f : \mathbb{M} \to \mathbb{R}$  such that

- ▶  $f(\xi) \approx y_{\xi}$  for all  $\xi \in \Xi$  and corresponding  $y_{\xi} \in \Upsilon$
- ►  $y_{\xi}$  sampled from function g:  $f \approx g$

#### COMMON APPROACHES: EXTRINSIC DIRECT INTERPOLATION

Pleasant conv. rates for RBF: Only loss of  $O(h_{\Xi}^{\text{codim}(\mathbb{M})/p})$  [Fuselier/Wright '12]



Problems:

- Polynomials/Splines/MLS presumably unstable
- RBF still face a loss of order although it is moderate

## RECENT APPROACH (REIF ET AL.): CONSTANT EXTENSION INTO AMBIENT SPACE

- ► Use some function space in ambient neighbourhood of M: RBF, MLS, Splines...
- ► Extend function constantly into ambient space and solve problem there.



Extension along normals:

 $\bullet$  Scattered sites are extended along the normals of  $\mathbb M$ 

## RECENT APPROACH (REIF ET AL.): CONSTANT EXTENSION INTO AMBIENT SPACE

- ► Use some function space in ambient neighbourhood of M: RBF, MLS, Splines...
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Extension along some suitable flow:

• Scattered sites are extended along e.g. gradient flow of implicit function

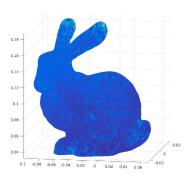
## RECENT APPROACH: CONSTANT EXTENSION INTO AMBIENT SPACE

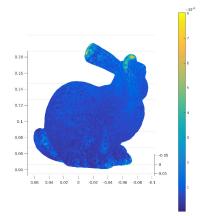
- ► Use some function space in ambient neighbourhood of M: RBF, MLS, Splines...
- ► Extend function constantly into ambient space and solve problem there.

THEOREM (M., 2016)

- Let  $f \in W_p^m(\mathbb{M})$  extended constantly along normals.
- Use quasi-interpolation with **TPB-Splines of order** *m*.

 $\implies$  Optimal convergence order m - k in  $W_{\rho}^{k}(\mathbb{M})$  for k < m - 1 regardless codim( $\mathbb{M}$ ).





Stanford Bunny Approximation

Locally weighted Least Squares and TPBS cells of edge length 0.025,

applied to about 30.000 data sites

Lars-B. Maier (Darmstadt)

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## FURTHER REMARKS

► Comparable convergence results hold for other approximation methods and/or extension flows at least if  $k < m - \text{codim}(\mathbb{M})$ .

- ► Convergence requires the data to increase as the degrees of freedom
- ► Application for M with boundary is possible, but requires reasonable boundary handling, e.g. coupling of splines.

# PART III: IRREGULAR SCATTERED DATA

### **PROBLEM STATEMENT**

without boundary

Setting:

▶  $\mathbb{M} \subseteq \mathbb{R}^d$  an embedded submanifold with  $q = \dim \mathbb{M} < 4$ 

◊ closed 
◊ compact

- $\Xi \subset \mathbb{M}$  a set of sites scattered over  $\mathbb{M}$
- ► Ξ contains clusters or considerable holes / gaps
- ▶  $\Upsilon$  function values to  $\Xi$

Task: Determine a function  $f : \mathbb{M} \to \mathbb{R}$  such that

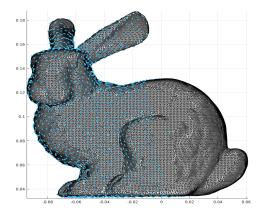
- ►  $f(\xi) \approx y_{\xi}$  for all  $\xi \in \Xi$  and corresponding  $y_{\xi} \in \Upsilon$
- ►  $y_{\xi}$  sampled from function g:  $f \approx g$

APPROACH (M., REIF): MULTI-LEVEL/HIERACHY APPROXIMATION

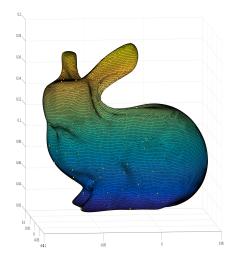
- ► Thin data sites until suitably sparse
- Apply »Sparse Data« method on thinned sites
- Approximate error in well-sampled regions using all sites
- ► Hierarchical approach might be used in second level

Advantage:

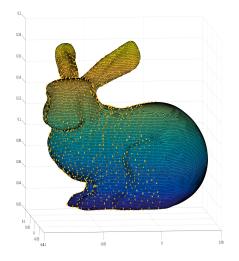
- Use of the same function space in first and following levels:
  - $\Rightarrow$  single explicit expression for solution



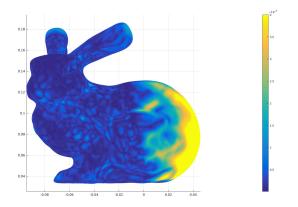
Hole in the data sites ( $\approx$  2.500): Rear side is lost



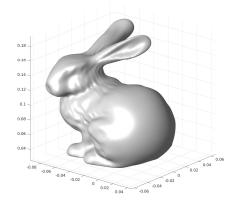
Result of first level with data sites ( $\approx$  160) over »pumpkin«



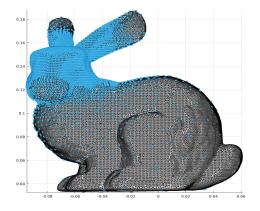
#### Result of first and second level with data sites



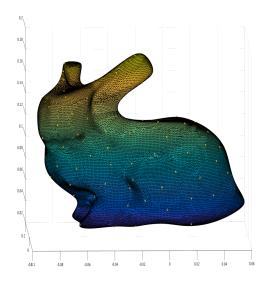
Error plot on resulting model, error cut at  $4 \cdot 10^{-3}$  for comparability reasons



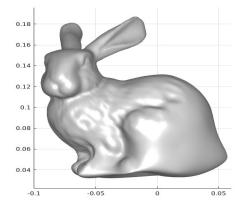
### Resulting model after both levels



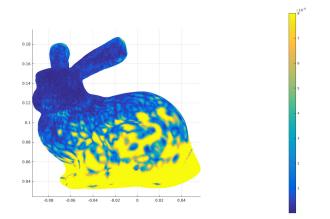
Data sites ( $\approx$  7.000) clustering in one area around the head



Result and sites ( $\approx$  130) of first step (h = 0.1)



Resulting model after 3 hierarchy levels (0.1, 0.05, 0.025) in second step



Error plot on resulting model, error cut at  $8 \cdot 10^{-4}$  for comparability reasons

### SUMMARY

New approaches to all kinds of scattered data approximation and/or interpolation on certain submanifolds:

- ► Overcome common difficulties
- Apply well-known concepts in novel setting
- Easy to implement
- Produce pleasant results

# THANK YOU!