

Stability for nonlinear inverse problems with a finite number of measurements

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Outline

Motivations

Calderon's problem with a finite number of measurements:
global uniqueness and Lipschitz stability

Lipschitz stability: linear subspaces

Lipschitz stability: manifolds

X, Y Banach spaces, $F : X \rightarrow Y$ possibly nonlinear

Given $y = F(x) \in Y$, determine $x \in X$.

X, Y Banach spaces, $F : X \rightarrow Y$ possibly nonlinear

Given $y = F(x) \in Y$, determine $x \in X$.

Stability estimate

$$\|x_1 - x_2\|_X \leq g(\|F(x_1) - F(x_2)\|_Y), \quad \text{where } g(t) \rightarrow 0, \text{ as } t \mapsto 0^+.$$

X, Y Banach spaces, $F : X \rightarrow Y$ possibly nonlinear

Given $y = F(x) \in Y$, determine $x \in X$.

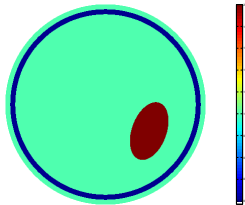
Logarithmic stability estimate

$$\|x_1 - x_2\|_X \leq C \left| \log(\|F(x_1) - F(x_2)\|_Y^{-1}) \right|^{-1}, \quad C > 0$$

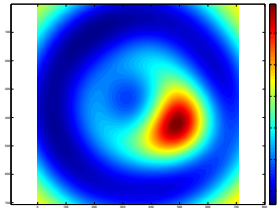
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► target



► reconstruction



based on: A. Greenleaf, M. Lassas, M. Santacesaria, S. Siltanen, and G. Uhlmann. Analysis & PDE 11, no. 8 (2018).

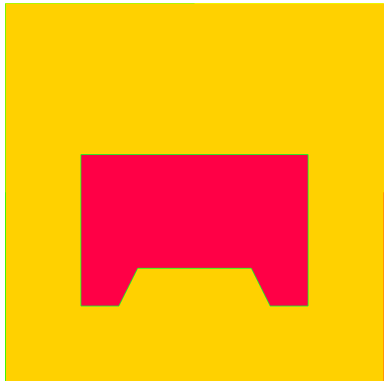
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$$\|x_1 - x_2\|_X \leq C \|F(x_1) - F(x_2)\|_Y, \quad C > 0$$

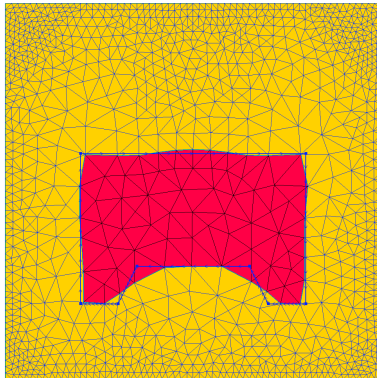
Lipschitz stability estimate

$$\|\mathbf{x}_1 - \mathbf{x}_2\|_X \leq C \|F(\mathbf{x}_1) - F(\mathbf{x}_2)\|_Y, \quad C > 0$$

► target



► reconstruction



E. Beretta, S. Micheletti, S. Perotto, and M. Santacesaria. Journal of Computational Physics 353 (2018).

Other motivations

- ▶ Lipschitz stability can be used to prove a nonlinear RIP (restricted isometry property) in compressed sensing [Candès-Tao (2006), Blumensath(2013)].
- ▶ Lipschitz continuous mapping can be well approximated by neural networks [Yarotsky (2017)].

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Motivations

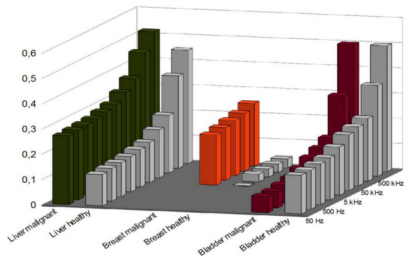
Calderon's problem with a finite number of measurements:
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Lipschitz stability: linear subspaces

Lipschitz stability: manifolds

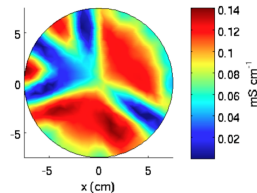
Conductivity imaging: motivations

► medical imaging



Low frequency contrast of σ

► nondestructive testing



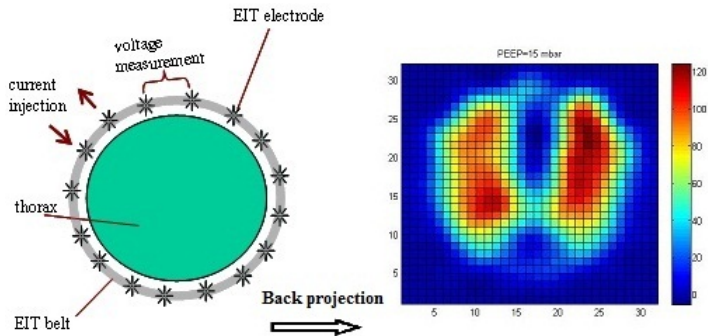
[Karhunen, Seppänen, Lehtikainen, Monteiro & Kaipio 2010]

[Karhunen, Seppänen, Lehtikainen, Monteiro, Kaipio, Blunt, Hyvönen]

credits: Widlak, T., Scherzer, O. (2012). Inverse Problems, 28(8), 084008.

Pros: high contrast, cheap, safe. **Cons:** low resolution.

Electrical Impedance Tomography (EIT)



Example: monitoring lung ventilation distribution

credits: Zhao et al. Crit Care. 2010

Calderón's problem for EIT

- ▶ $D \subset \mathbb{R}^d$, $d \geq 2$: bounded Lipschitz domain
- ▶ $\sigma \in L^\infty(D)$, $\sigma(x) \geq \sigma_0 > 0$: unknown conductivity
- ▶ Conductivity equation:

$$\begin{cases} -\operatorname{div}(\sigma \nabla u) = 0 & \text{in } D, \\ u = f & \text{on } \partial D. \end{cases} \quad (1)$$

- ▶ Dirichlet-to-Neumann (DN) map $\Lambda_\sigma : H^{1/2}(\partial D) \rightarrow H^{-1/2}(\partial D)$:

$$f \longmapsto \sigma \frac{\partial u}{\partial \nu} \Big|_{\partial D}$$

Calderón's problem

Given Λ_σ , determine σ in D .

Some known results

Basic questions:

- ▶ Uniqueness: injectivity of $\sigma \mapsto \Lambda_\sigma$
- ▶ stability estimates: continuity of $\Lambda_\sigma \mapsto \sigma$
- ▶ reconstruction algorithm

Theoretical contributions by: Calderón, Sylvester–Uhlmann, Nachman, Novikov, Alessandrini, Astala–Päivärinta, Haberman, Caro–Rogers and many others.

Usual reduction to the Gel'fand-Calderón inverse problem for the Schrödinger equation

$$(-\Delta + q)u = 0 \quad \text{in } D, \quad \Lambda_q(u|_{\partial D}) = \frac{\partial u}{\partial \nu} \Big|_{\partial D},$$

which will be considered for the next few slides.

A finite number of measurements

$$\begin{cases} (-\Delta + q)u = 0 & \text{in } D, \\ u = f & \text{on } \partial D, \end{cases} \quad \Lambda_q(f) = \frac{\partial u}{\partial \nu} \Big|_{\partial D}.$$

- Most results need an infinite number of measurement.

“Realistic” Calderón’s problem

$$\{(f_l, \Lambda_q(f_l))\}_{l=1,\dots,N} \quad \leadsto \quad q$$

A priori assumptions: $q \in \mathcal{W}_R$ if

- $q \in \mathcal{W}$: known finite dimensional subspace of $L^\infty(D)$;
- 0 is not a Dirichlet eigenvalue for $-\Delta + q$ in D ;
- $\|q\|_{L^\infty(D)} \leq R$ for some $R > 0$.

Nonlinear problem - global uniqueness

Theorem (G.S. Alpert, M.S. (2018)¹)

Take $d \geq 3$ and let $D \subseteq \mathbb{R}^d$ be a bounded Lipschitz domain and $\mathcal{W} \subseteq L^\infty(D)$ be a finite dimensional subspace. There exists $N \in \mathbb{N}$ such that for any $R > 0$ and $q_1 \in \mathcal{W}_R$, the following is true.

There exist $\{f_l\}_{l=1}^N \subseteq H^{1/2}(\partial D)$ such that for any $q_2 \in \mathcal{W}_R$, if

$$\Lambda_{q_1} f_l = \Lambda_{q_2} f_l, \quad l = 1, \dots, N,$$

then

$$q_1 = q_2.$$

Similar result for Calderón's problem as well.

Ideas of the proof

- Alessandrini's identity to go from the boundary to the interior.

$$\langle g, (\Lambda_q - \Lambda_0)f \rangle_{H^{\frac{1}{2}}(\partial D) \times H^{-\frac{1}{2}}(\partial D)} = \int_D q u_g^0 u_f^q dx$$

- CGO solutions (Faddeev, Sylvester-Uhlmann): the complex parameters belong to a countable subset of \mathbb{C}^d . For $k \in \mathbb{Z}^d$, take $u^0(x) = e^{\zeta_2^k \cdot x}$ and CGO solution $u^q(x) = e^{\zeta_1^k \cdot x} (1 + r^k(x))$, with $\zeta_1^k, \zeta_2^k \in \mathbb{C}^d$ such that

$$\zeta_j^k \cdot \zeta_j^k = 0, \quad \zeta_1^k + \zeta_2^k = -2\pi i k, \quad \|r^k\|_{L^2(\mathbb{T}^d)} \leq c/t_k$$

- Order the frequencies: $\rho: l \in \mathbb{N} \mapsto k_l \in \mathbb{Z}^d$ (bijection)
- Define the nonlinear measurement operator $U: L^\infty([0, 1]^d) \rightarrow \ell^\infty$ by

$$(U(q))_l = \int_D q(x) e^{-2\pi i k_l \cdot x} (1 + r^{k_l}(x)) dx$$

- $U = F + B$, where, F Fourier transform, B is a contraction (t_k large)

Sketch of the proof

- Define the nonlinear operator $U: L^\infty(D) \rightarrow \ell^\infty$ by

$$(U(q))_l = \int_D q(x) e^{-2\pi i k_l \cdot x} (1 + r^{k_l}(x)) dx, \quad U = F + B$$

- Assume that $\Lambda_{q_1} f_l = \Lambda_{q_2} f_l$ for $l = 1, \dots, N$
- Then $(P_N U)(q_1) = (P_N U)(q_2)$
- Using that B is a contraction we obtain $q_1 = q_2$, since

$$\begin{aligned} \|q_1 - q_2\|_{L^2} &= \|F(q_1 - q_2)\|_{\ell^2} \\ &\leq \|P_N^\perp F(q_1 - q_2)\|_{\ell^2} + \|P_N(B(q_2) - B(q_1))\|_{\ell^2} \\ &\leq \|P_N^\perp F_{\mathcal{W}}(q_1 - q_2)\|_{\ell^2} + \frac{1}{2} \|q_1 - q_2\|_{L^2}, \end{aligned}$$

provided that N is chosen so that

$$\|P_N^\perp F_{\mathcal{W}}\|_{L^2([0,1]^d) \rightarrow \ell^2} \leq \frac{1}{4}.$$

On the number of measurements N

- ▶ The number of measurements N depends only on \mathcal{W} through

$$\|(I - P_N)FP_{\mathcal{W}}\|_{\mathcal{H} \rightarrow \ell^2} \leq 1/4.$$

- ▶ Relation with sampling theory: how many Fourier measurements does one need to reconstruct a function in \mathcal{W} ?
- ▶ It allows for an explicit calculation of N :
 - bandlimited potentials

$$N = \dim \mathcal{W}$$

- piecewise constant potentials

$$N = O((\dim \mathcal{W})^4)$$

(up to log factors, and possibly not optimal)

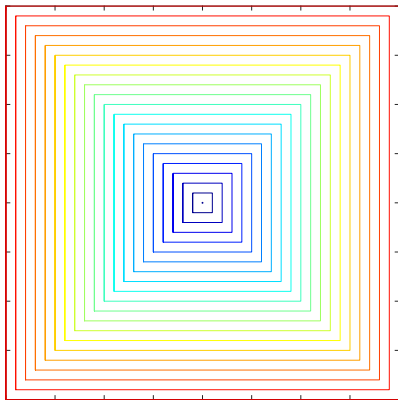
- low-scale wavelets

$$N = O(\dim \mathcal{W})$$

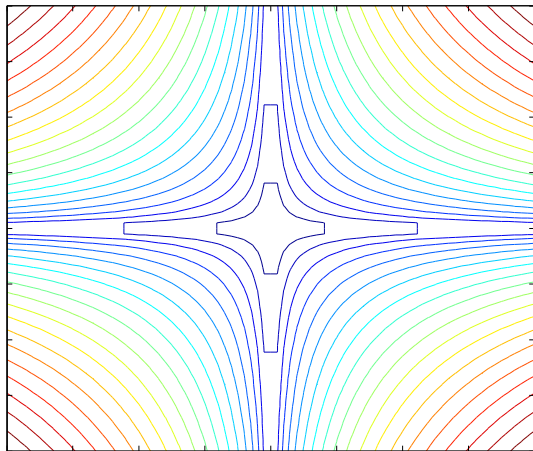
(up to log factors, proven only in 1D, but easy generalization)

- ▶ The ordering of \mathbb{Z}^d is crucial

Possible orderings of \mathbb{Z}^d



(a) Linear ordering



(b) Hyperbolic ordering (Jones, Adcock, Hansen, 2017)

Lipschitz stability

Theorem (G.S. Alberti, M.S. (2018)²)

Under the same assumptions, there exist $\{f_l\}_{l=1}^N \subseteq H^{1/2}(\partial D)$ such that for every $q_2 \in \mathcal{W}_R$, we have

$$\|q_2 - q_1\|_{L^2(D)} \leq e^{CN^{\frac{1}{2}+\alpha}} \left\| (\Lambda_{q_2} f_l - \Lambda_{q_1} f_l)_{l=1}^N \right\|_{H^{-1/2}(\partial D)^N}$$

for some $C > 0$ depending only on D , R and α .

- ▶ Many Lipschitz stability estimates with the full DN map (Alessandrini, Beretta, Francini, Gaburro, de Hoop, Scherzer, Sincich, Vessella...).
- ▶ The exponential $e^{CN^{\frac{1}{2}+\alpha}}$ is consistent with the severe ill-posedness of this IP.
- ▶ Nonlinear reconstruction algorithm based on Banach fixed point theorem.

Local stability with a priori known boundary measurements

Theorem (G.S. Alpert, M.S. (2019)³)

Let $d \in \{3, 4\}$, let N be as in the previous Theorem and take $q_0 \in \mathcal{W}_R$.

There exist $\delta, C > 0$ and $L \in \mathbb{N}$ depending only on $\Omega, C_\rho, R, \mathcal{W}$ and

$\|(-\Delta + q_0)^{-1}\|_{H^{-1}(\Omega) \rightarrow H_0^1(\Omega)}$ such that for every $q_1, q_2 \in \mathcal{W}_R$, if

$$\|q_0 - q_j\|_{L^2(\Omega)} \leq \delta \quad j = 1, 2, \quad (2)$$

then

$$\|q_2 - q_1\|_{L^2(\Omega)} \leq C \left\| (f_{n,1}^L - f_{n,2}^L)_{n=1}^N \right\|_{H^{1/2}(\partial\Omega)^N},$$

where

$$f_{n,j}^L = \sum_{l=1}^L ((S_{\zeta_n}^{q_0}(\Lambda_{q_j} - \Lambda_{q_0}))^l(f_{q_0}^n)), \quad j = 1, 2, \quad (3)$$

and $S_{\zeta_n}^{q_0}$ is the generalized single layer operator corresponding to the Faddeev-Green function and to the potential q_0 and $f_{q_0}^n$ are Faddeev-CGO functions associated to q_0 .

Other results

- ▶ (Rüland-Sincich 2019) fractional Calderón problem,
- ▶ (Harrach 2019) complete electrode model for EIT.

(Partially) Open questions

- ▶ Two-dimensional case.
- ▶ Extensions to other infinite dimensional IP, e.g. inverse scattering, elasticity.
- ▶ General Lipschitz stability result for a class of ill-posed inverse problems.

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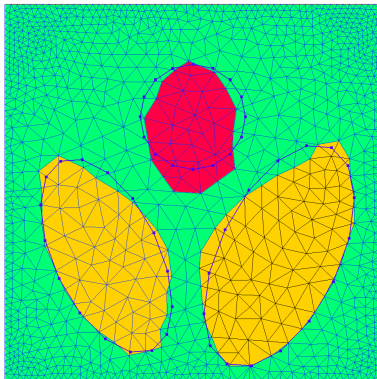
Lipschitz stability: linear subspaces

Lipschitz stability: manifolds

Lipschitz stability: basic result

- ▶ X, Y Banach spaces, $A \subseteq X$ open subset
- ▶ **(priors)** $W \subseteq X$ finite-dimensional, $K \subseteq W \cap A$ compact and convex.

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Theorem

Let $F \in C^1(A, Y)$ be such that $F|_{W \cap A}$ and $F'(x)|_W, x \in W \cap A$, are injective.

Then there exists a constant $C > 0$ such that

$$\|x_1 - x_2\|_X \leq C \|F(x_1) - F(x_2)\|_Y, \quad x_1, x_2 \in K.$$

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Many Lipschitz stability estimates for EIT with **full** DN map: Alessandrini, Beretta, Francini, Gaburro, de Hoop, Meftahi, Scherzer, Sincich, Vessella and many others.

Lipschitz stability with finite measurements: setting

Discretization: $Q_N: Y \rightarrow Y$ bounded linear map such that there exists a subspace $\tilde{Y} \subseteq Y$ satisfying:

1. $\|Q_N|_{\tilde{Y}}\|_{\mathcal{L}(\tilde{Y}, Y)} \leq D$ for every $N \in \mathbb{N}$ and some $D > 0$;
2. $Q_N|_{\tilde{Y}} \rightarrow I_{\tilde{Y}}$ as $N \rightarrow \infty$ with respect to the strong operator topology, i.e.

$$\lim_{N \rightarrow \infty} \|y - Q_N y\|_Y = 0$$

for every $y \in \tilde{Y}$.

Examples

- Y Hilbert space, $\{G_j\}_{j \in \mathbb{N}}$ exhaustive sequence of finite dimensional and nested subspaces.

$$Q_N = P_{G_N} \text{ orthogonal projection onto } G_N.$$

$$\tilde{Y} = Y.$$

- $Y = \mathcal{L}_c(Y^1, Y^2)$ with Y^1, Y^2 Banach spaces. $P_N^2 \rightarrow I_{Y^2}$ and $(P_N^1)^* \rightarrow I_{Y^1}$ strongly.

$$Q_N(y) = P_N^2 y P_N^1.$$

$$\tilde{Y} = \{T \in Y : T \text{ is compact}\}.$$

Lipschitz stability with finite measurements: main result

Theorem (G.S. Alberti, M.S. (2019) ⁴)

Let $K \subseteq A$ be convex. Suppose there exists $C > 0$ such that

$$\|x_1 - x_2\|_X \leq C \|F(x_1) - F(x_2)\|_Y, \quad \text{for } x_1, x_2 \in K.$$

(i) If $K \subseteq W \cap A$ is compact, where W is a finite dimensional subset of X and for every $\xi \in K$, $\text{ran}(F'(\xi)|_W) \subseteq \tilde{Y}$, then

$$\lim_{N \rightarrow +\infty} s_N = 0, \quad s_N = \sup_{\xi \in K} \|(I - Q_N)F'(\xi)\|_{W \rightarrow Y}.$$

(ii) If $s_N \leq \frac{1}{2C}$, then

$$\|x_1 - x_2\|_X \leq 2C \|Q_N(F(x_1)) - Q_N(F(x_2))\|_Y, \quad x_1, x_2 \in K.$$

The smoothing condition: $\text{ran}(F'(\xi)|_W) \subseteq \tilde{Y}$

- Y Hilbert space, $\{G_j\}_{j \in \mathbb{N}}$ exhaustive sequence of finite dimensional and nested subspaces. $Q_N = P_{G_N}$ orthogonal projection onto G_N .

Since $Q_N \rightarrow I_Y$ strongly and $\tilde{Y} = Y$, the condition is satisfied.

- $Y = \mathcal{L}_c(Y^1, Y^2)$ with Y^1, Y^2 Banach spaces. $P_N^2 \rightarrow I_{Y^2}$ and $(P_N^1)^* \rightarrow I_{Y^1}$ strongly.

$$Q_N(y) = P_N^2 y P_N^1.$$

Assuming that $F'(\xi)\tau : Y^1 \rightarrow Y^2$ is **compact**, i.e. $F'(\xi)\tau \in \tilde{Y}$, for every $\xi \in K, \tau \in W$ then the condition is satisfied.

On the number of measurements N

N depends on the Lipschitz constant C for the full data and on the subspace W :

$$\sup_{\xi \in K} \|(I - Q_N)F'(\xi)\|_{W \rightarrow Y} \leq \frac{1}{2C}$$

which can be explicitly computed in several cases.

Example I: electrical impedance tomography

Let \mathcal{N}_σ be the Neumann-to-Dirichlet map and assume

$$\|\sigma_1 - \sigma_2\|_{L^\infty(\Omega)} \leq C \|\mathcal{N}_{\sigma_1} - \mathcal{N}_{\sigma_2}\|_{L^2_\diamond(\partial\Omega) \rightarrow L^2_\diamond(\partial\Omega)}, \quad \sigma_1, \sigma_2 \in K,$$

where K is a compact subset of a finite dimensional subspace of L^∞ conductivities ($L^2_\diamond(\partial\Omega) = \{f \in L^2(\partial\Omega) : \int_{\partial\Omega} f \, ds = 0\}$). Then there exists $N \in \mathbb{N}$ such that

$$\|\sigma_1 - \sigma_2\|_\infty \leq 2C \|\mathcal{P}_N \mathcal{N}_{\sigma_1} \mathcal{P}_N - \mathcal{P}_N \mathcal{N}_{\sigma_2} \mathcal{P}_N\|_{L^2_\diamond(\partial\Omega) \rightarrow L^2_\diamond(\partial\Omega)}, \quad \sigma_1, \sigma_2 \in K.$$

$\Omega \subseteq \mathbb{R}^2$ unit disk. Let \mathcal{P}_N be the projection on the trigonometric current patterns $\sin(n\theta), \cos(n\theta)$, for $n \leq N, \theta \in \partial\Omega$.

Then we have $N = O(C^2)$ (recall that for EIT $C = O(\exp(\dim W))$).

Example II: inverse scattering

$$\begin{cases} \Delta u + k^2 n(x)u = 0 & \text{in } \mathbb{R}^3, \\ u = e^{ikx \cdot d} + u^s & \text{in } \mathbb{R}^3, \\ \text{radiation condition for } u^s \end{cases}$$

- ▶ $k > 0$ is the (fixed) wavenumber, $d \in S^2$,
- ▶ $n \in L^\infty(\mathbb{R}^3; \mathbb{C})$ is the refractive index with $\text{Im}(n) \geq 0$ in \mathbb{R}^3 and $\text{supp}(1 - n) \subseteq B$ for some open ball B .

Problem. Given the far field $u_n^\infty(\hat{x}, d) \in L^2(S^2 \times S^2)$ at fixed $k > 0$, find n in B .

Example II: inverse scattering

- ▶ $X = L^\infty(B; \mathbb{C})$, $Y = L^2(S^2 \times S^2)$;
- ▶ $A = L_+^\infty(B) = \{f \in L^\infty(B; \mathbb{C}) : \operatorname{Im}(n) \geq \lambda \text{ in } B \text{ for some } \lambda > 0\}$;
- ▶ W finite-dimensional subspace of $L^\infty(B; \mathbb{C})$
- ▶ K convex and compact subset of $W \cap A$;

[Bourgeois 2013] proved Lipschitz stability in this case:

$$\|n_1 - n_2\|_{L^\infty(B)} \leq C \|u_{n_1}^\infty - u_{n_2}^\infty\|_{L^2(S^2 \times S^2)}, \quad n_1, n_2 \in K.$$

- ▶ $Q_N: L^2(S^2 \times S^2) \rightarrow L^2(S^2 \times S^2)$ bounded linear maps, $Q_N \rightarrow I_{L^2(S^2 \times S^2)}$ strongly.

$$\|n_1 - n_2\|_{L^\infty(B)} \leq 2C \|Q_N(u_{n_1}^\infty) - Q_N(u_{n_2}^\infty)\|_{L^2(S^2 \times S^2)}, \quad n_1, n_2 \in K.$$

Example: Q_N projections onto the span of the first N spherical harmonics.

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Low-dimensional manifolds

$M \subseteq X$ (Banach space) a n -dimensional C^1 manifold with atlas $\{(U_i, \varphi_i)\}_{i \in I}$:

- ▶ topologically embedded in X (i.e. not a C^1 submanifold),
- ▶ α -Hölder for $\alpha \in (0, 1]$, that is $\varphi_i^{-1} : \varphi_i(U_i) \rightarrow M$ is α -Hölder, for $i \in I$.

Examples:

- ▶ Indicator functions on balls with variable centres, radii and intensities,
 - Hölder in $L^p(\mathbb{R}^n)$, $p > 1$, Lipschitz in $L^1(\mathbb{R}^n)$, not C^1 embedded.
- ▶ Indicator functions on simplexes.
- ▶ A manifold generated by a neural network (e.g. autoencoder, GAN, etc.)

Hölder-Lipschitz stability from an infinite amount of measurements

Theorem (G.S. Alberti, A. Arroyo, M.S. (2020) ⁵)

Let X and Y be Banach spaces, $\alpha \in (0, 1]$, $M \subseteq X$ be an n -dimensional differentiable manifold α -Hölder in X and $K \subseteq M$ be a compact set. Consider $F \in C^1(M, Y)$ with:

1. F is injective,
2. the differential $dF_x: T_x M \rightarrow Y$ is injective for every $x \in M$.

Then there exists a constant $C > 0$ such that

$$\|x - y\|_X \leq C \|F(x) - F(y)\|_Y^\alpha, \quad x, y \in K.$$

Key ideas:

- ▶ short distance and long distance cases;
- ▶ workaround for the lack of convexity of K .

Finite number of measurements

Let $Q_N: Y \rightarrow Y$ and $\tilde{Y} \subseteq Y$ be as in the linear case.

Theorem (G.S. Alberti, A. Arroyo, M.S. (2020)⁶)

Let X and Y be Banach spaces, $M \subseteq X$ an n -dimensional C^1 manifold Lipschitz in X , $K \subseteq M$ a compact set. Consider $F \in C^1(M, Y)$ satisfying:

1. F is injective;
2. the differential $dF_x: T_x M \rightarrow Y$ is injective for every $x \in M$;
3. $\text{ran}(F|_K) \subseteq \tilde{Y}$;
4. $\text{ran}(dF_x) \subseteq \tilde{Y}$ for every $x \in M$.

Then

$$\|x - y\|_X \leq C \|Q_N F(x) - Q_N F(y)\|_Y, \quad x, y \in K,$$

for some $C > 0$ and every sufficiently large $N \in \mathbb{N}$.

Ongoing work and conclusions

- ▶ Global reconstruction algorithms for both linear and manifold cases.
- ▶ Applications of the results on manifolds on classical inverse problems.
- ▶ Add sparsity and random subsampling: non-linear compressed sensing.
- ▶ Approximate the inverse mapping with a deep neural network.

Thank you!