

Instability mechanisms for inverse problems

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Outline

1. Background
2. Entropy and capacity estimates
3. Examples

Image deblurring

Blurred versions of an image $f \in L^2(\mathbb{R}^2)$:

Original



f

Little Blur



$\chi_1 * f$

Much Blur



$\chi_2 * f$

Instability

Try to recover an image f from its blurred version $f_b = \chi * f$.
Fourier transform: $\hat{f}_b(\xi) = \hat{\chi}(\xi)\hat{f}(\xi)$. Different scenarios:

- ▶ $\hat{\chi} \in C_c^\infty(\mathbb{R}^2) \rightsquigarrow$ high frequencies are lost completely
- ▶ $\hat{\chi}(\xi) \sim e^{-c|\xi|} \rightsquigarrow$ high frequencies exponentially damped
- ▶ $\hat{\chi}(\xi) \sim |\xi|^{-s} \rightsquigarrow$ high frequencies polynomially damped

If one measures $m = \chi * f + \varepsilon$ where ε is **noise**, naive reconstruction (multiply by $\frac{1}{\hat{\chi}(\xi)}$ on Fourier side) gives

$$f_{\text{naive}} = f + \mathcal{F}^{-1}\left\{\frac{1}{\hat{\chi}(\xi)}\hat{\varepsilon}(\xi)\right\}.$$

High frequency noise can lead to huge errors in reconstruction!

Instability

Heuristics:

- ▶ **smoothing** (blurring) **implies instability**
- ▶ **strong smoothing** (fast decay of singular values) **implies strong instability**

In deblurring, the forward operator was the simple linear operator

$$F : L^2(\mathbb{R}^2) \rightarrow L^2(\mathbb{R}^2), \quad F(f) = \mathcal{F}^{-1}\{\hat{\chi}(\xi)\hat{f}(\xi)\}$$

and the analysis could be done directly on the Fourier side.

What happens for more complicated inverse problems?

Instability

We give a framework for studying rigorously the **inherent instability** in various linear and nonlinear inverse problems, e.g.

- ▶ (geodesic) X-ray/Radon transforms also with limited data
- ▶ analytic/unique continuation
- ▶ control and inverse problems for heat/wave equations
- ▶ Calderón problem

We identify three mechanisms for instability:

1. Strong global smoothing
2. Microlocal smoothing
3. Weak global smoothing (“iterated small regularity gain”)

Abstract problem

Abstract inverse problem. Consider a map $F : X \rightarrow Y$ between metric spaces. Given $y \in Y$, find $x \in X$ with

$$F(x) = y.$$

Conditions for well-posedness [Hadamard 1902]:

1. (Existence) Given $y \in Y$, there is $x \in X$ with $F(x) = y$.
2. (Uniqueness) If $F(x_1) = F(x_2)$, then $x_1 = x_2$.
3. (Stability) The solution x depends continuously on y .

In IP stability typically fails, but may have **conditional stability**. Important for convergence guarantees for statistical algorithms

[Abraham, Giordano, Monard, Nickl, Paternain, 2019–].

Conditional stability

Fact. If $F : X \rightarrow Y$ is an injective continuous map and $K \subset X$ is compact, then $F|_K$ is a homeomorphism.

Restricting to a compact set $K \subset X$ (a priori bounds) gives conditional stability: there is a modulus of continuity $\omega = \omega_{F,X,Y,K}$ so that

$$d_X(x_1, x_2) \leq \omega(d_Y(F(x_1), F(x_2))), \quad x_j \in K.$$

Examples:

1. If $\omega(t) = t$, one has Lipschitz stability.
2. If $\omega(t) = t^\alpha$, one has Hölder stability.
3. If $\omega(t) = |\log t|^{-\sigma}$, one has logarithmic stability.

Calderón problem/EIT

Example. Consider $\operatorname{div}(\gamma \nabla u) = 0$ in $\Omega \subset \mathbb{R}^n$. In this case

$$X = (L_+^\infty(\Omega), \|\cdot\|_{L^\infty}),$$

$$Y = (B(H^{1/2}, H^{-1/2}), \|\cdot\|_* = \|\cdot\|_{H^{1/2} \rightarrow H^{-1/2}}),$$

$$F : \gamma \mapsto \Lambda_\gamma \quad (\text{Dirichlet-to-Neumann map}).$$

- ▶ Uniqueness is highly nontrivial [Sylvester-Uhlmann 1987, Astala-Päivärinta 2006, ...].
- ▶ Stability fails: F^{-1} is not continuous $F(X) \rightarrow X$ [Alessandrini 1988].

Conditional stability

Logarithmic stability [Alessandrini 1988]: if $n \geq 3$ and $K = \{\gamma \in L^\infty(\Omega) : \gamma \geq E^{-1}, \|\gamma\|_{H^{n/2+2+\varepsilon}(\Omega)} \leq E\}$, then

$$\|\gamma_1 - \gamma_2\|_{L^\infty} \leq \omega(\|F(\gamma_1) - F(\gamma_2)\|_*), \quad \gamma_j \in K,$$

where ω is a **logarithmic** modulus of continuity.

Many improvements [Barceló-Faraco-Ruiz 2007, ...]. Also, if K is contained in an N -dim. space, get Lipschitz stability with constant blowing up as $N \rightarrow \infty$ [Alessandrini-Vessella 2005].

Exponential instability [Mandache 2001]: logarithmic stability is **optimal**, i.e. if the above estimate holds for some ω , then $\omega(t) \geq c|\log t|^{-\sigma}$ for some $\sigma = \sigma(n)$.

Instability

The argument in [Mandache 2001] was based on

- ▶ **capacity estimates** for $K \subset X$
- ▶ **entropy estimates** for $F(K) \subset Y$

These estimates were proved by ad hoc constructions, e.g. by spherical harmonics and separation of variables in the ball.

We will replace these ad hoc constructions by **structural properties** of the forward operator and related spaces.

This works for many inverse problems (not just EIT), and for general geometries and coefficients.

Message: while a stability result is **“hard”** (requires a quantitative uniqueness result), an instability result is **“soft”** (follows from structural “compression” properties).

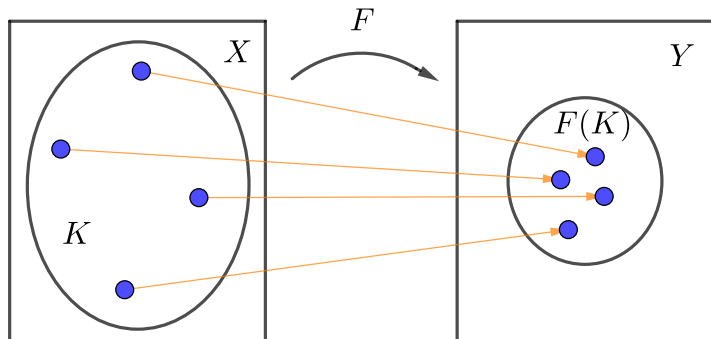
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Instability

Consider a map $F : X \rightarrow Y$ between metric spaces, and the inverse problem of solving $F(x) = y$ when $x \in K$.

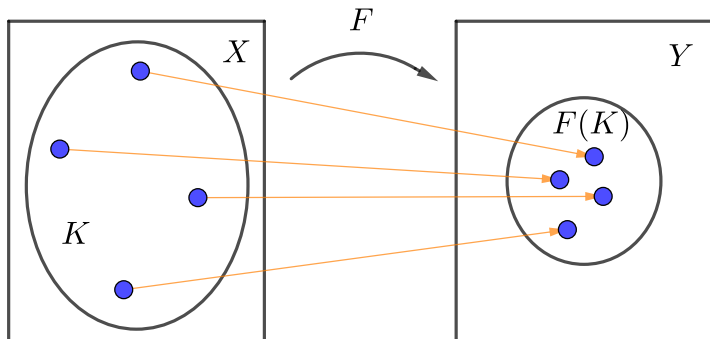
One expects instability if F “strongly compresses distances”.



Instability [Mandache 2001]

Expect instability if K is “extended” (\exists large ε -discrete sets),
whereas $F(K)$ is “compressed” (\exists relatively small δ -nets).

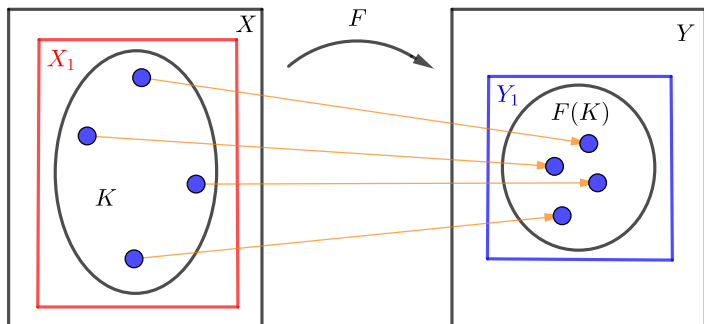
Then pigeonhole principle \implies instability.



Instability

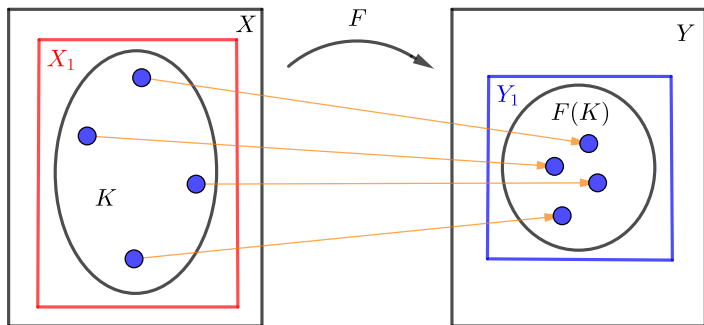
X and Y are often Banach spaces and K is a bounded set in some subspace $X_1 \subset X$. Suppose that:

$F(K) \subset Y_1$ where $Y_1 \subset Y$ is a “compressed” subspace.



Instability

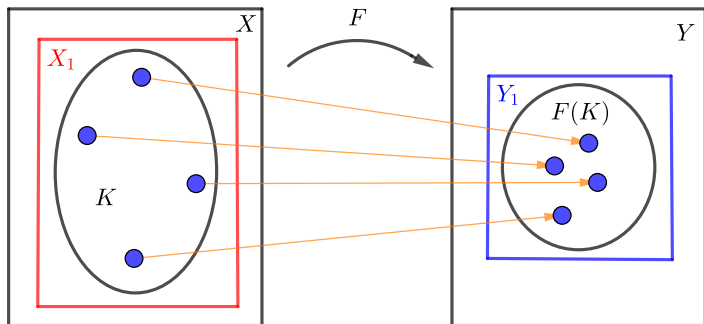
Then ε -discrete sets in K can be studied via the embedding $i: X_1 \rightarrow X$, and δ -nets of $F(K)$ are related to the embedding $j: Y_1 \rightarrow Y$.



Instability

If $F(K) \subset Y_1$, enough to study embeddings between function spaces (the forward operator “disappears”!). The ideal tool:

Capacity and entropy numbers (see [Edmunds-Triebel 2008]).



Entropy and capacity numbers

Let $A : X \rightarrow Y$ be a bounded operator between Banach spaces. Let $U_X = \{x \in X : \|x\|_X \leq 1\}$. For $k \geq 1$, define

$e_k(A) = \inf \{ \delta : \text{there is a } \delta\text{-net of } A(U_X) \text{ with } 2^{k-1} \text{ elements} \},$

$c_k(A) = \sup \{ \varepsilon : A(U_X) \text{ has an } \varepsilon\text{-discrete set of } > 2^{k-1} \text{ elements} \}.$

Enough to consider **entropy numbers**, since $c_k(A) \sim e_k(A)$.
In our case, study $e_k(i : X_1 \rightarrow X)$ and $e_k(j : Y_1 \rightarrow Y)$.

The numbers $e_k(A)$ measure compactness of A :

- ▶ A is compact iff $e_k(A) \rightarrow 0$ as $k \rightarrow \infty$;
- ▶ A has finite rank iff $e_k(A)$ decay exponentially.
- ▶ In Hilbert spaces, $e_k(A)$ related to **singular values** $\sigma_k(A)$.

Entropy number bounds

Theorem. (Smooth spaces are compressed)

Let $\Omega \subset \mathbb{R}^n$ be a bounded domain with smooth boundary.
Then

$$e_k(i : H^{s+\delta}(\Omega) \rightarrow H^s(\Omega)) \sim k^{-\delta/n}.$$

Similar bounds are valid for $H^s(\partial\Omega)$ and on manifolds.
(Argument based on [Weyl law](#) for eigenvalues.)

If $\partial\Omega$ is real-analytic and $A^R(\partial\Omega)$ is the space of real-analytic functions with uniform Cauchy bounds¹, then

$$e_k(i : A^R(\partial\Omega) \rightarrow H^s(\partial\Omega)) \lesssim e^{-ck^{\frac{1}{n}}}.$$

¹ $|\partial^\alpha f(x)| \leq CR^{|\alpha|} \alpha!$ for some $C > 0$

Entropy number bounds

For EIT, need to study spaces of operators.

Theorem. (Spaces of smoothing operators are compressed)

Define the following spaces of operators on $\partial\Omega$:

$$Z^m = \{T \in B(H^{1/2}, H^{-1/2}) : T = T^*, T(H^{1/2}) \subset H^{-1/2+m}\},$$
$$W^R = \{T \in B(H^{1/2}, H^{-1/2}) : T = T^*, T(H^{1/2}) \subset A^R\}.$$

Then

$$e_k(i : Z^m \rightarrow B(H^{1/2}, H^{-1/2})) \lesssim k^{-\frac{m}{2n} + \delta},$$
$$e_k(i : W^R \rightarrow B(H^{1/2}, H^{-1/2})) \lesssim e^{-ck^{\frac{1}{2n-1}}}.$$

Question. Optimality of exponents?

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Instability mechanisms

We will give examples of instability in inverse problems, based on the abstract approach and **three instability mechanisms**:

1. Strong global smoothing
2. Microlocal smoothing
3. Weak global smoothing

Calderón problem

Let Λ_q be the DN map for $(-\Delta_g + q)u = 0$ in M , where (M, g) is compact with boundary.

Theorem. If $M, g, \partial M$ are **real-analytic** and

$$\|q_1 - q_2\|_{H^s} \leq \omega(\|\Lambda_{q_1} - \Lambda_{q_2}\|_*), \quad \|q_j\|_{H^{s+\delta}} \leq 1,$$

with $q_j = 0$ near ∂M , then $\omega(t) \geq c|\log t|^{-\sigma}$ for some $\sigma > 0$.

Proof. Take $X_1 = H^{s+\delta}$, $X = H^s$, $Y = B(H^{1/2}, H^{-1/2})$. Need to show that

$$F : q \mapsto \Lambda_q - \Lambda_0$$

is compressing. But if $q = 0$ near ∂M , then $\Lambda_q - \Lambda_0$ is in the strongly compressed space of **analytic smoothing operators**. (Earlier result: Euclidean ball [Mandache 2001].)

Radon transform with limited data

Let Rf be Radon transform of $f \in H_K^s(\mathbb{R}^2)$, $K \subset \mathbb{R}^2$ compact.

Theorem. If $\mathcal{L} \subsetneq \{\text{lines in } \mathbb{R}^2\}$ is closed, and if

$$\|f\|_{H^s(\mathbb{R}^2)} \leq \omega(\|Rf\|_{H^t(\mathcal{L})}), \quad \|f\|_{H_K^{s+\delta}} \leq 1,$$

then $\omega(t) \geq c|\log t|^{-\sigma}$ for some $\sigma > 0$.

Proof. $\chi_{\mathcal{L}}R$ is microlocally smoothing (it smooths out any singularity near some (x_0, ξ_0)). Testing with wave packets at (x_0, ξ_0) shows that ω cannot be Hölder [Stefanov-Uhlmann 2009]. For our result, use analytic/Gevrey microlocal smoothing of

$$\chi_{\mathcal{L}}R : P(H^{s+\delta}) \rightarrow H^t$$

where P is a microlocal cutoff. Need Weyl law for microlocally elliptic Ψ DOs. Also works for nonlinear IPs.

Geodesic X-ray transform

Let $I f$ be the **geodesic X-ray transform** that integrates f over maximal geodesics in (M, g) .

Theorem. Let (M, g) compact, C^∞ , strictly convex, nontrapping 2-mfld that has **interior conjugate points**. If

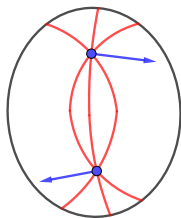
$$\|f\|_{H^s} \leq \omega(\|I f\|_{H^t}), \quad \|f\|_{H^{s+\delta}} \leq 1,$$

then ω **cannot be a Hölder modulus**.

Proof. [Monard-Stefanov-Uhlmann 2015]: I has a microlocal kernel. Consider

$$I : P(H^{s+\delta}) \rightarrow H^t$$

where P is a projection to the microlocal kernel.



Instability

Smoothing implies instability: F maps into smooth functions / (microlocally) smoothing operators (“compresses distances”) \implies IP is **strongly unstable**. Similar results for unique continuation, backward heat equation, ... if ∂M and the coefficients near (a point of) ∂M are **C^∞ /real-analytic**.

This works for general geometries and variable coefficients. Earlier results for balls, half-spaces, and constant coefficients [Hadamard 1923, John 1960, Mandache 2001, ...].

So far we proved strong instability only if the structures are **C^∞ /real-analytic**. What happens for **very rough coefficients**? Could the stability improve?

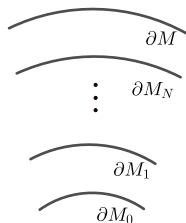
Rough coefficients

Answer: **NO**. For rough coefficients there is a different compression mechanism (“iterated small regularity gain”) \implies instability for Calderón problem etc even with $g \in L^\infty$.

For w solving $\Delta_g w = 0$ near ∂M , factorize

$$\Lambda_q - \Lambda_0 : f \mapsto w|_{\partial M_0} \mapsto w|_{\partial M_1} \mapsto \dots \mapsto w|_{\partial M_N} \mapsto \partial_\nu w|_{\partial M}$$

where $T_j : w|_{\partial M_j} \mapsto w|_{\partial M_{j+1}}$ has **tiny regularity gain** (Meyers estimate) but $\|T_j\|$ is large. Estimate entropy numbers of the composition and optimize w.r.t. N .



Summary

- ▶ **Smoothing** (strong/weak/microlocal) implies **instability**.
- ▶ Instability is due to **compression** properties of forward operator, precisely characterized by **entropy numbers**.
- ▶ Applies to **linear** and **nonlinear** inverse problems with **general geometries and coefficients**.
- ▶ Regularization / wave equations / nonlinear PDE?