

# The Bayesian Formulation and Well-Posedness of Fractional Elliptic Inverse Problems

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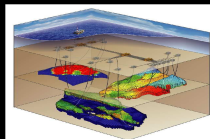
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# Background

# Elliptic Inverse Problem. Dashti, Stuart



## MATHEMATICAL MODEL

$$-\nabla \cdot (e^{u(x)} \nabla p(x)) = f \text{ in } D,$$
$$p = h \text{ on } \partial D.$$

## DATA

$$y_j = p(x_j) + \eta_j, \quad x_j \in D.$$

Uncertainty in input  $u \in \mathcal{H}$ .

Treat it as random.

# Inverse Problems

$$\boxed{u} \xrightarrow{\text{(F)PDE}} \boxed{p} \longrightarrow \boxed{y = \mathcal{O}(p) + \eta}$$

$$\boxed{u \sim \pi} \xrightarrow{\text{BIP}} \boxed{u | y \sim \mu}$$

# Goal

$\pi$

$$\frac{d\mu}{d\pi}(u) \propto g(u; y)$$

$\mu$

Bayesian Inverse Problem

$g$  involves an FPDE.

Formulation, well-posedness?

Setting

# Basic Framework

$$L_A := -\nabla \cdot (A(x)\nabla_x), s \in (0, 1).$$

$$\begin{cases} L_{AP}^s = f, & \text{in } D, \\ \partial_{AP} = 0, & \text{on } \partial D. \end{cases}$$

Set-up: Uncertainty on  $A$  and  $s$ .  $f$  known.

Aim: Learn  $A(x)$  and  $s$  from partial and noisy observations of  $p$

$$y = \mathcal{O}(p) + \eta \in \mathbb{R}^m, \quad \eta \sim N(0, \Gamma).$$

# Why Fractional Order Bayesian Inverse Problems?

- Order uncertainty e.g. in viscoelastic materials (Zayernouri).
- Integer order PDE  $\leftrightarrow$  fractional order in boundary.



# Bayesian Formulation

$$\text{Forward Map } u := (s, A) \mapsto p \quad \begin{cases} L_{AP}^s = f, & \text{in } D, \\ \partial_{AP} = 0, & \text{on } \partial D, \end{cases}$$

$$\text{Data } y = \mathcal{O}(p) + \eta =: \mathcal{G}(u) + \eta.$$

## NON-SMOOTH CASE

- $f \in L_{\text{avg}}^2(D)$ . (Right-hand side)
- $X := (0, 1) \times L^\infty(D)$ . (Prior space)

## SMOOTH CASE

- $f \in L_{\text{avg}}^p(D)$  for some  $p \geq 2$ . (Right-hand side)
- $X := (\frac{d}{2p}, 1] \cap C^1(\bar{D})$ . (Prior space)

# Results

# Main results

## Theorem 1

If  $\pi(X) = 1$ , the BIP is well formulated:

$$\frac{d\mu^y}{d\pi}(u) = \frac{1}{Z} \exp\left(-\frac{1}{2}|y - \mathcal{G}(u)|_{\Gamma}^2\right).$$

## Theorem 2

There is  $C = C(r)$  such that, for all  $y_1, y_2 \in \mathbb{R}^m$  with  $|y_1|, |y_2| \leq r$ ,

$$D_{\text{Hell}}(\mu^{y_1}, \mu^{y_2}) \leq C|y_1 - y_2|.$$

# Two Comments

# Fractional Laplacian

Orthonormal basis of eigenfunctions of  $L_A$ ,  $\psi_k \in H_{\text{avg}}^1(D)$ ,  
Eigenvalues  $0 < \lambda_1 \leq \lambda_2 \leq \dots$ . For  $p(x) = \sum_{k=1}^{\infty} p_k \psi_k(x)$ ,

$$L_A^s p(x) := \sum_{k=1}^{\infty} \lambda_k^s p_k \psi_k(x).$$

Domain of  $L_A^s$ : functions  $p \in L_{\text{avg}}^2(D)$  with

$$\|p\|_{H^s}^2 := \sum_{k=1}^{\infty} \lambda_k^s p_k^2 < \infty.$$

# Extension Problem. Caffarelli, Silvestre

Let  $a := 1 - 2s$ ,

$$B(x) := \begin{pmatrix} A(x) & \mathbf{0} \\ \mathbf{0} & 1 \end{pmatrix} \in \mathbb{R}^{(d+1) \times (d+1)}.$$

We denote by  $P_{s,A} : D \times (0, \infty) \rightarrow \mathbb{R}$  the solution to

$$\begin{cases} \nabla \cdot (\xi^a B \nabla P_{s,A}) = 0, & \text{in } D \times (0, \infty), \\ \partial_A P_{s,A} = 0, & \text{on } \partial D \times [0, \infty), \\ P_{s,A}(x, 0) = p_{s,A}(x), & \text{on } D. \end{cases}$$

Weak form: for  $\phi \in H^1(D \times (0, \infty), \xi^a d\xi dx)$

$$\int_D \int_0^\infty \langle B \nabla P_{s,A}, \nabla \phi \rangle \xi^a d\xi dx = c_s \int_D L_{A,p_{s,A}}^s \phi(x, 0) dx.$$

# Literature and Current Work

# Literature

- BAYESIAN INVERSE PROBLEMS

Kaipio-Somersalo 2006, Marzouk-Xiu 2009, Stuart 2010

- REGULARITY THEORY OF FPDEs

Caffarelli-Silvestre 2007, Stinga 2010, Caffarelli-Stinga 2016

- NUMERICAL SOLUTION OF FPDEs

Nochetto-Otarola-Salgado 2015, Zayernouri-Karniadakis 2015



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Brown: Ainsworth, Darbon, Guzmán, Karniadakis,...

# Current Directions

- Numerical aspects: work in progress with C. Glusa (Brown).
  - Numerical methods for FPDEs via extension problem.
  - Scalable MCMC algorithms: extended pCN.
- Other Laplacians.
- Variable order forward maps, priors, and noise models.



N. Garcia Trillos, D. Sanz-Alonso, *The Bayesian formulation and well-posedness of fractional elliptic inverse problems*.  
Inverse Problems, 2017.