## Regularization of rough linear functionals & Adaptivity

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Motivation (PDEs with rough input data)

Projection in dual norms

Numerical results

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## Motivation

(PDEs with rough input data)



### **Motivation**

Consider your favorite well-posed variational formulation of a PDE ...

 $\begin{cases} \text{Find } u \in \mathbb{U} \text{ such that:} \\ b(u, v) = \ell(v), \quad \forall v \in \mathbb{V}. \end{cases}$ 

R. ARAYA, E. BEHRENS, R. RODRÍGUEZ. A posteriori error estimates for elliptic problems with Dirac delta source terms. NUMER. MATH. (2006) 105:193–216.

J.P. AGNELLI, E.M. GARAU, P. MORIN. A posteriori error estimates for elliptic problems with Dirac measure terms in weighted spaces. ESAIM: MATH. MODEL. NUMER. ANAL. (2014) 48:1557–1581.

## **Motivation**

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## Other examples of rough linear functionals

Singular functions: 
$$\ell(v) = \int_{\Omega} f v \dots$$
 with a singular  $f$ 

• Action over the derivatives: 
$$\ell(v) = \int_{\Omega} \vec{F} \cdot \nabla v$$
.  
 $(\vec{F} \text{ could be singular as well!})$ 

• Point sources: 
$$\ell(\mathbf{v}) = \langle \delta_{\mathbf{x}_0}, \mathbf{v} \rangle = \mathbf{v}(\mathbf{x}_0).$$

• Line sources: 
$$\ell(v) = \int_C \varphi v$$
 ..... where C is a contour.

Surface sources: 
$$\ell(v) = \int_{S} \psi v$$
 ..... where *S* is a surface.

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#### **Regularization idea:**

Find a regularized RHS  $\ell_H \approx \ell$  such that  $\|\ell - \ell_H\|_{V^*}$  is controlled.

• Using  $\ell_H$  we define the regularized problem:

 $\begin{cases} \text{Find } u_H \in \mathbb{U} \text{ such that:} \\ b(u_H, v) = \ell_H(v), \quad \forall v \in \mathbb{V}. \end{cases}$ 

The regularized problem can be approached numerically using AFEM:

 $\begin{cases} \text{Find } u_h \in \mathbb{U}_h \text{ such that:} \\ b(u_h, v_h) = \ell_H(v_h), \quad \forall v_h \in \mathbb{V}_h. \end{cases}$ 

Error estimation:



#### 2-step adaptive algorithm:

- Set initial mesh  $\mathcal{T}_0$  and tolerance  $\varepsilon > 0$ .
- $\blacktriangleright [\mathcal{T}_{H}, \ell_{H}] = \mathsf{RHS}(\mathcal{T}_{0}, \ell, \gamma \varepsilon/2)$
- $\blacktriangleright [\mathcal{T}_h, u_h] = \mathsf{PDE}(\mathcal{T}_H, \ell_H, \varepsilon/2)$

$$\|u - u_h\|_{\mathbb{U}} \leq \frac{1}{\gamma} \underbrace{\|\ell - \ell_H\|_{\mathbb{V}^*}}_{\text{regularization}} + \underbrace{\|u_H - u_h\|_{\mathbb{U}}}_{\text{discretization}} \leq \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon$$

#### Desirable features for constructing the regularized RHS $\ell_H$ :

- Must be adaptively built.
- ▶ If so, we need to localize the error  $\|\ell \ell_H\|_{V^*}$  with some indicator.
- We want l<sub>H</sub> to be a standard function in some piece-wise polynomial space such that:

$$\langle \ell_H, \mathbf{v}_h \rangle = \int_{\Omega} \ell_H \mathbf{v}_h = \sum_j w_j \, \ell_H(x_j) \mathbf{v}_h(x_j)$$

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# Projection in dual norms

**Projection is a minimization problem.** Let  $V_H$  a piecewise polynomial space defined over an affine simplicial mesh  $\mathcal{T}_H$ . Given a rough  $\ell \in \mathbb{V}^*$ , we aim to find  $\ell_H \in V_H$  such that:

$$\ell_{H} = \underset{g_{H} \in V_{H}}{\operatorname{argmin}} \|\ell - g_{H}\|_{\mathbb{V}^{*}}$$
(Min)

where

$$\|\cdot\|_{\mathbb{V}^*} := \sup_{\mathbf{v}\in\mathbb{V}}rac{\langle\,\cdot\,,\mathbf{v}
angle_{\mathbb{V}^*,\mathbb{V}}}{\|\mathbf{v}\|_{\mathbb{V}}}$$

Fortunately, (Min) is equivalent to the mixed (saddle point) formulation:

$$\left\{ \begin{array}{rcl} {\sf Find}\;(r,\ell_{H})\in\mathbb{V}\times V_{H} & {\sf such that:} \\ \\ \left\langle J_{\mathbb{V}}(r),v\right\rangle_{\mathbb{V}^{*},\mathbb{V}}+\int_{\Omega}\ell_{H}\,v & =\ell(v) & \forall v\in\mathbb{V} \\ \\ \\ \int_{\Omega}g_{H}\,r & =0 & \forall g_{H}\in V_{H} \end{array} \right.$$

In particular,  $\|r\|_{\mathbb{V}}^{q-1} = \|J_{\mathbb{V}}(r)\|_{\mathbb{V}^*} = \|\ell - \ell_H\|_{\mathbb{V}^*}$  (loc. error representative).

#### Discrete dual norm projection

Recall the mixed (saddle point) formulation:

$$\begin{cases} \quad \mathsf{Find} \ (r,\ell_H) \in \mathbb{V} \times V_H \quad \text{such that:} \\ \\ \left\langle J_{\mathbb{V}}(r), v \right\rangle_{\mathbb{V}^*,\mathbb{V}} + \int_{\Omega} \ell_H v &= \ell(v) \qquad \forall v \in \mathbb{V} \\ \\ \\ \int_{\Omega} g_H r &= 0 \qquad \forall g_H \in V_H \end{cases}$$

Given a discrete (conforming) piecewise polynomial space  $V_h \subset \mathbb{V}$  defined over an affine simplicial mesh  $\mathcal{T}_h$ , we propose to solve:

 $\begin{cases} \text{Find } (r_h, \ell_H) \in V_h \times V_H \quad \text{such that:} \\ \left\langle J_{\mathbb{V}}(r_h), v_h \right\rangle_{\mathbb{V}^*, \mathbb{V}} + \int_{\Omega} \ell_H v_h &= \ell(v_h) \quad \forall v_h \in V_h \\ \int_{\Omega} g_H r_h &= 0 \quad \forall g_H \in V_H \end{cases}$ 

## **Questions:**

- 1. Is this fully-discrete formulation well-posed and stable?
- 2. How good is this new approximation  $\ell_H$ ?
- 3. Is this new  $\ell_H$  a minimizer in some sense?
- 4. Can we use  $||\mathbf{r}_h||_{\mathbb{V}}$  as an error indicator for adaptivity?

**Under Fortin compatibility condition:** There exists a continuous operator  $\Pi : \mathbb{V} \to V_h$  such that:

 $\int_{\Omega} g_H \, \Pi v = \int_{\Omega} g_H \, v \qquad \forall v \in V_h, \ \forall g_H \in V_H$  $\| \Pi v \|_{\mathbb{V}} \leq C_{\Pi} \| v \|_{\mathbb{V}}, \qquad \forall v \in \mathbb{V}, \ \text{ with mesh-independent } C_{\Pi} > 0$ 

#### Example of compatible pairs $V_H/V_h$ :

- ▶  $\mathbb{P}_0/(\mathbb{P}_1 + \text{bubbles})$  is a compatible pair discretizing  $W^{-1,\rho}(\Omega)/W^{1,q}_0(\Omega)$ .
- $\mathbb{P}_1/\mathbb{P}_2$  is a compatible pair discretizing  $W^{-1,p}(\Omega)/W_0^{1,q}(\Omega)$ .
- F. MILLAR, I. MUGA, S. R. & K.G. VAN DER ZEE. Projection in negative norms and the regularization of rough linear functionals. NUMERISCHE MATHEMATIK (2022) 150:1087–1121.

#### Fortin compatibility implies ...

1. Is this discrete formulation well-posed and stable? **YES!** 

 $\|r_h\|_{\mathbb{V}} \leq \|\ell\|_{\mathbb{V}^*} \quad \text{and} \quad \|\ell_H\|_{\mathbb{V}^*} \lesssim C_{\Pi} \|\ell\|_{\mathbb{V}^*}$ 

2. How good is this new approximation? It's quasi-optimal!

$$\|\ell - \ell_H\|_{\mathbb{V}^*} \leq (1 + 2C_{\Pi}) \inf_{g_H \in V_H} \|\ell - g_H\|_{\mathbb{V}^*}$$

Is this new ℓ<sub>H</sub> a minimizer in some sense?
 Indeed, it's the minimizer in the discrete dual norm:

$$\sup_{\mathbf{v}_h \in \mathbf{V}_h} \frac{\langle \ell - \ell_H, \mathbf{v}_h \rangle_{\mathbb{V}^*, \mathbb{V}}}{\|\mathbf{v}_h\|_{\mathbb{V}}} \leq \sup_{\mathbf{v}_h \in V_h} \frac{\langle \ell - g_H, \mathbf{v}_h \rangle_{\mathbb{V}^*, \mathbb{V}}}{\|\mathbf{v}_h\|_{\mathbb{V}}} \qquad \forall g_H \in V_H$$

4. Can we use  $||r_h||_{\mathbb{V}}$  as an error indicator for adaptivity? **YES!** 

$$\|r_h\|_{\mathbb{V}}^{q-1} \leq \|\ell - \ell_H\|_{\mathbb{V}^*} \leq C_{\Pi} \|r_h\|_{\mathbb{V}}^{q-1} + \underbrace{\sup_{v \in \mathbb{V}} \frac{\langle \ell, v - \Pi v \rangle_{\mathbb{V}^*, \mathbb{V}}}{\|v\|_{\mathbb{V}}}}_{\operatorname{osc}(\ell)}$$

## Numerical Results

 $W^{-1,p}$  projection of Dirac delta in 1D ( $\mathbb{P}_1/\mathbb{P}_2$  compatible pair)



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### Elliptic ODE with projected Dirac delta source (p = 2, $\mathbb{P}_0/\mathbb{P}_2$ C.P.)



## 2D Dirac delta projection in $W^{-1,p}$ ( $\mathbb{P}_1/\mathbb{P}_2$ C.P.)



Sequence of mesh refinements for p = 1.2



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## Elliptic PDE on a L-shape domain with Dirac source ( $\mathbb{U}_h = \mathbb{P}_1$ )



$$u(x) = \frac{1}{2\pi} \log \left( \|x - x_0\| \right) + \|x\|^{\frac{2}{3}} \sin \left( \frac{2}{3} (\pi - \theta_x) \right)$$



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## Elliptic PDE on a L-shape domain with Dirac source $(\mathbb{U}_h = \mathbb{P}_1)$

 $\begin{cases} -\Delta u = \delta_{x_0} + f \\ +BCs \end{cases} \text{ such that the exact solution is:} \\ u(x) = \frac{1}{2\pi} \log \left( \|x - x_0\| \right) + \|x\|^{\frac{2}{3}} \sin \left( \frac{2}{3} (\pi - \theta_x) \right) \end{cases}$ 



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## Line source in 2D ( $\mathbb{P}_0/\mathbb{P}_1$ + bubbles C.P. )





## **Future directions**

- Rougher RHS (e.g., derivatives of Dirac delta)
- ► Non-conforming FEM.

#### Main reference:

F. Millar, I. Muga, S. Rojas & K.G. Van der Zee.

Projection in negative norms and the regularization of rough linear functionals. NUMERISCHE MATHEMATIK (2022) 150:1087–1121.

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