

# INVITATION TO THE DOCTORAL SEMINAR

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**Kha Van Huynh, MSc**

Universität Klagenfurt

**“An extension of minimization based formulations and the projected gradient method with some applications”**



<https://classroom.aau.at/3nh-a2g>  Wednesday, 16 December 2020

 10:00 a.m.

## Abstract

An inverse problem of reconstructing  $x$  such that  $A(x, u) = 0$  from the data  $y$  satisfying  $C(u) = y$  can be written in the minimization form

$$\operatorname{argmin}_{(x,u)} \{ \mathcal{J}(x, u; y) : (x, u) \in M_{\text{ad}}(y) \}.$$

The classical approaches usually consider  $\mathcal{J}(x, u; y) = \frac{1}{2} \|C(x, u) - y\|^2$ ,  $M_{\text{ad}} = \{(x, u) : A(x, u) = 0\}$  or  $\mathcal{J}(x, u; y) = \frac{1}{2} \|A(x, u)\|^2$ ,  $M_{\text{ad}} = \{(x, u) : C(x, u) = y\}$ . Here we follow a new approach assuming that the observation operator  $C$  can be inverted on its range, as is the case, e.g., in the practically relevant setting of a finite dimensional observation space. In this case, we can split  $u$  into two parts: the observed data part  $\tilde{u}$  and the homogeneous data part  $\hat{u}$ ,  $u = \tilde{u} + \hat{u}$  where  $C(\hat{u}) = 0$  and  $C(\tilde{u}) = y$ . Let  $C^{\text{ri}}$  be a right inverse operator of  $C$ , then the problem becomes to find  $x$  such that

$$A(x, C^{\text{ri}}(y) + \hat{u}) = 0, \quad \hat{u} \in \operatorname{Ker}(C) := \{\hat{u} : C(\hat{u}) = 0\}$$

with the new minimization form which is to find

$$(x, \hat{u}) \in \operatorname{argmin}_{(x, \hat{u})} \{ \mathcal{J}(x, \hat{u}; y) : (x, \hat{u}) \in M_{\text{ad}}(y) \}.$$

Under some more conditions on the smoothness of  $\mathcal{J}$  with respect to  $y$ , we can prove the *well-definedness*, *stability* and *convergence* of minimizers. In practice, often an iterative regularization method will be applied to reconstruct  $x$ . Here we want to mention the projected gradient method

$$x_{k+1} = \operatorname{Proj}_{M_{\text{ad}}}(x_k - \mu_k \nabla J(x_k))$$

where  $\operatorname{Proj}_{M_{\text{ad}}}$  is the projection onto  $M_{\text{ad}}$  and  $x$  stands for  $(x, \hat{u})$ , for brevity. This method can be applied to many problems. Here we show numerical results with a Matlab implementation for three examples: inverse ground-water filtration (GWF), impedance acoustic tomography (IAT), and electrical impedance tomography (EIT).

Barbara Kaltenbacher and the Department of Mathematics look forward to seeing you at the talk!

