

# Part C: Regularization and conditional stability: some case studies in a Hilbert space setting

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Let  $X$  and  $Y$  denote infinite dimensional **Hilbert spaces**, equipped with norms  $\|\cdot\|$  and inner products  $\langle \cdot, \cdot \rangle$ .

We consider also here **linear inverse problems** in form of linear operator equations

$$Ax = y \quad (x \in X, y \in Y) \quad (*)$$

with a **bounded linear forward operator**  $A \in \mathcal{L}(X, Y)$  and **nonlinear inverse problems** in form of nonlinear equations

$$F(x) = y \quad (x \in \mathcal{D}(F) \subseteq X, y \in Y). \quad (**)$$

We denote again by  $x^\dagger$  solutions to equations  $(*)$  and  $(**)$ .



We recall Nashed's ill-posedness concept for (\*)

Definition (▷ NASHED 1987)

We call a linear operator equation (\*) **well-posed in the sense of Nashed** if the range  $\mathcal{R}(A)$  of  $A$  is a closed subset of  $Y$ , consequently **ill-posed in the sense of Nashed** if the range is not closed, i.e.  $\mathcal{R}(A) \neq \overline{\mathcal{R}(A)}^Y$ . In the ill-posed case, the equation (\*) is called **ill-posed of type I** if the range  $\mathcal{R}(A)$  contains an infinite dimensional closed subspace, and **ill-posed of type II** otherwise.

We recall local ill-posedness concept for (\*\*)

Definition (▷ HOFMANN/SCHERZER 1994)

The operator equation (\*\*) is called **locally well-posed** at the solution  $x^\dagger \in \mathcal{D}(F)$  if there is a closed ball  $\mathcal{B}_r(x^\dagger)$  with radius  $r > 0$  and center  $x^\dagger$  such that for every sequence  $\{x_n\}_{n=1}^\infty \subset \mathcal{B}_r(x^\dagger) \cap \mathcal{D}(F)$  the convergence of images  $\lim_{n \rightarrow \infty} \|F(x_n) - F(x^\dagger)\| = 0$  implies the convergence of the preimages  $\lim_{n \rightarrow \infty} \|x_n - x^\dagger\| = 0$ .  
Otherwise (\*\*) is called **locally ill-posed** at  $x^\dagger$ .

**Stable approximate solution** of ill-posed linear problems (\*) and locally ill-posed nonlinear problems (\*\*) requires some kind of **stabilization**. We consider **conditional stability**, **Tikhonov regularization** and **Lavrentiev regularization** as different such approaches.

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# Tikhonov regularization and conditional stability

For ill-posed equations (\*\*\*) and a general situation of  $F$ , classical **Tikhonov regularization** is a very popular method, see, e.g., ▷ ENGL/HANKE/NEUBAUER 1996 (Chapter 10), with regularized solutions  $x_\alpha^\delta$  which are minimizers of

$$\|F(x) - y^\delta\|^2 + \alpha \|x - \bar{x}\|^2 \rightarrow \min, \quad \text{subject to } x \in \mathcal{D}(F). \quad (\text{Tik})$$

For a priori parameter choices  $\alpha_* = \alpha_*(\delta)$  or a posteriori parameter choices  $\alpha_* = \alpha_*(y^\delta, \delta)$  of the regularization parameter  $\alpha > 0$  satisfying

$$\alpha_* \rightarrow 0 \quad \text{and} \quad \frac{\delta^2}{\alpha_*} \rightarrow 0 \quad \text{as} \quad \delta \rightarrow 0, \quad (\searrow)$$

we have, under conditions on  $F$  and  $\mathcal{D}(F)$ , norm convergence in  $X$  of  $x_{\alpha_*}^\delta$  to  $\bar{x}$ -minimum norm solutions  $x^\dagger$  of (\*\*\*) as  $\delta \rightarrow 0$ .

The convergence is based on the **minimizing property**

$$\|F(x_\alpha^\delta) - y^\delta\|^2 + \alpha \|x_\alpha^\delta - \bar{x}\|^2 \leq \|F(x^\dagger) - y^\delta\|^2 + \alpha \|x^\dagger - \bar{x}\|^2 \quad (MP)$$

of the regularized solution  $x_\alpha^\delta \in \mathcal{D}(F)$ , which yields

$$\|x_\alpha^\delta - \bar{x}\| \leq \sqrt{\frac{\delta^2}{\alpha} + \|x^\dagger - \bar{x}\|^2}$$

and

$$\|F(x_\alpha^\delta) - y^\delta\| \leq \sqrt{\delta^2 + \alpha \|x^\dagger - \bar{x}\|^2}.$$

If  $F$  is **weakly sequentially closed**, then we have weak convergence (in subsequences) of  $x_{\alpha_*}^\delta$  to  $\bar{x}$ -minimum norm solutions under  $(\searrow)$  as  $\delta \rightarrow 0$ . Since Hilbert spaces have the **Kadec-Klee** property, this yields even norm convergence.

## Convergence rates for Tikhonov regularization

To obtain convergence rates, solution smoothness with respect to the forward operator is needed, which is expressed by some kind of **source condition**.

We recall a form expressed by **variational inequalities**:

### Variational source condition (VSC)

We assume to have a constant  $0 < \beta \leq 1$  and a **concave** index function  $\varphi$  such that

$$\beta \|x - x^\dagger\|^2 \leq \|x - \bar{x}\|^2 - \|x^\dagger - \bar{x}\|^2 + \varphi(\|F(x) - F(x^\dagger)\|) \quad \forall x \in \mathcal{M}.$$

The function  $\varphi : [0, \infty) \rightarrow [0, \infty)$  is called an **index function** if it is continuous and strictly increasing with  $\varphi(0) = 0$ .

Under variational source conditions (VSC)

$$\alpha_* = \frac{\delta^2}{\varphi(\delta)}$$

is an appropriate a priori choice of the regularization parameter.

For practical use, however, discrepancy principles are more important. Consider for prescribed  $0 < q < 1$  and large  $\alpha_0 > 0$

$$\Delta_q := \left\{ \alpha_j : \alpha_j := q^j \alpha_0, \quad j = 1, 2, \dots \right\}.$$

### Sequential discrepancy principle (SDP)

For prescribed  $\tau > 1$  we say that the regularization parameter  $\alpha_* \in \Delta_q$  is chosen according to the sequential discrepancy principle (SDP) if

$$\|F(x_{\alpha_*}^\delta) - y^\delta\| \leq \tau \delta < \|F(x_{\alpha_*/q}^\delta) - y^\delta\|.$$

## Theorem $\triangleright$ HOFMANN/MATHÉ 2012

Suppose that  $x^\dagger$  obeys (VSC) for some  $0 < \beta \leq 1$  and some concave index function  $\varphi$ .

- (i) Let  $\alpha_* = \alpha_*(\delta) > 0$  be selected according to the a priori parameter choice  $\alpha_* := \frac{\delta^2}{\varphi(\delta)}$ .
- (ii) For prescribed  $\tau > 1$  let  $\alpha_* = \alpha_*(\delta, y^\delta) > 0$  be chosen according to the sequential discrepancy principle (SDP).

Provided that  $x_{\alpha_*}^\delta \in \mathcal{M}$  for all  $0 < \delta \leq \delta_{max}$  and some  $\delta_{max} > 0$  we have for both parameter choices (i) and (ii) the convergence rate

$$\|x_{\alpha_*}^\delta - x^\dagger\|^2 = \mathcal{O}(\varphi(\delta)) \quad \text{as } \delta \rightarrow 0.$$

We mention that both parameter choices in (i) and (ii) satisfy

$$\alpha_* \rightarrow 0 \quad \text{and} \quad \frac{\delta^2}{\alpha_*} \rightarrow 0 \quad \text{as} \quad \delta \rightarrow 0. \quad (\searrow)$$



## Conditional stability

Stable approximate solutions can also be found by having **conditional stability estimates** of the form

$$\|x - x^\dagger\|^2 \leq \varphi(\|F(x) - F(x^\dagger)\|) \quad \text{for all } x \in \mathcal{D}(F) \cap Q \quad (CSE)$$

for a concave index function  $\varphi$  and an appropriate set  $Q \supset \{x^\dagger\}$ .

Often  $Q$  depends on properties of  $x^\dagger$  and is not known a priori. Consequently, (CSE) is not directly applicable for finding stable approximate solutions to (\*\*). Additional tools are needed.

▷ CHENG/YAMAMOTO IP 2000 ▷ HOFMANN/YAMAMOTO IP 2010

## Combine (CSE) and Tikhonov regularization: $Q = B_r(x^\dagger)$

Under  $\|x - x^\dagger\|^2 \leq \varphi(\|F(x) - F(x^\dagger)\|)$  for all  $x \in \mathcal{D}(F) \cap B_r(x^\dagger)$  the sol.  $x^\dagger$  is **unique** in  $B_r(x^\dagger)$ , **(\*\*) is locally well-posed** at  $x^\dagger$ .

An appropriate a priori choice  $\alpha_* = \alpha_*(\delta)$  is here

$$\underline{c} \delta^2 \leq \alpha_*(\delta) \leq \bar{c} \delta^2, \quad 0 < \underline{c} \leq \bar{c} < \infty, \quad \text{satisfying}$$

$$\alpha_* \rightarrow 0 \quad \text{and} \quad 0 < 1/\bar{c} \leq \delta^2/\alpha_* \leq 1/\underline{c} < \infty \rightarrow 0 \quad \text{as} \quad \delta \rightarrow 0. \quad (\longrightarrow)$$

From (MP) we derive  $\|x_\alpha^\delta - \bar{x}\|^2 \leq \frac{\delta^2}{\alpha} + \|x^\dagger - \bar{x}\|^2$ , consequently

$$\|x_\alpha^\delta - x^\dagger\|^2 \leq 2(\|x_\alpha^\delta - \bar{x}\|^2 + \|x^\dagger - \bar{x}\|^2) \leq \frac{2\delta^2}{\alpha} + 4\|x^\dagger - \bar{x}\|^2$$

and hence

$$\|x_{\alpha_*}^\delta - x^\dagger\|^2 \leq \frac{2}{\underline{c}} + 4\|x^\dagger - \bar{x}\|^2.$$

For  $r > \sqrt{\frac{2}{\underline{c}} + 4\|x^\dagger - \bar{x}\|^2}$  we thus have  $x_{\alpha_*}^\delta \in \mathcal{D}(F) \cap B_r(x^\dagger)$ .

(CSE) applies for  $x_{\alpha_*}^\delta \in \mathcal{D}(F) \cap B_r(x^\dagger) = \mathcal{D}(F) \cap Q$  and yields

$$\|x_{\alpha_*}^\delta - x^\dagger\|^2 \leq \varphi(\|F(x_{\alpha_*}^\delta) - y^\delta\| + \|F(x^\dagger) - y^\delta\|),$$

moreover with

$$\|F(x_{\alpha_*}^\delta) - y^\delta\| \leq \sqrt{\delta^2 + \alpha_* \|x^\dagger - \bar{x}\|^2} \leq \sqrt{\delta^2 + \bar{c} \delta^2 \|x^\dagger - \bar{x}\|^2}.$$

For concave index functions  $\varphi$  we find constant  $K > 0$  such that

$$\|x_{\alpha_*}^\delta - x^\dagger\|^2 \leq K \varphi(\delta) = \mathcal{O}(\varphi(\delta)) \quad \text{as } \delta \rightarrow 0.$$

**Using regularization under conditional stability  
is like putting into the hole Q while playing golf!**

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# Oversmoothing regularization in Hilbert scales

If in generalized Tikhonov regularization

$$\|F(x) - y^\delta\|^2 + \alpha \mathcal{R}(x) \rightarrow \min, \quad \text{subject to } x \in \mathcal{D} := \mathcal{D}(F) \cap \mathcal{D}(\mathcal{R}),$$

with convex penalty functionals  $\mathcal{R}$ , solutions  $x^\dagger \in \mathcal{D}(F)$  to (\*\*)  
are **not smooth enough** w.r.t.  $\mathcal{R}$  such that  $\mathcal{R}(x^\dagger) = \infty$ ,

then for regularized solutions  $x_\alpha^\delta$  the minimizing property

$$\|F(x_\alpha^\delta) - y^\delta\|^2 + \alpha \mathcal{R}(x_\alpha^\delta) \leq \|F(x^\dagger) - y^\delta\|^2 + \alpha \mathcal{R}(x^\dagger) \quad (MP)$$

does not help anyway. In  $\triangleright$  HOFMANN/MATHÉ 2017 we found  
situations, where for  $\alpha_* = \alpha_*(y^\delta, \delta)$  satisfying

$$\|F(x_{\alpha_*}^\delta) - y^\delta\| = \tau \delta, \quad \text{for some } \tau > 1, \quad (DP)$$

$$\alpha_* \rightarrow 0 \quad \text{and} \quad \frac{\delta^2}{\alpha_*} \rightarrow \infty \quad \text{as} \quad \delta \rightarrow 0. \quad (\nearrow)$$

Precisely, we have extended the results of the seminal paper by  $\triangleright$  NATTERER 1984 from the linear case (\*) to the nonlinear case (\*\*) formulated in a Hilbert scale setting generated by an operator  $B$  as follows:  
For an unbounded linear operator  $B : \mathcal{D}(B) \subset X \rightarrow X$ , which is self-adjoint and satisfies with some  $m > 0$

$$\|Bx\| \geq m \|x\| \quad \forall x \in \mathcal{D}(B).$$

The Hilbert scale  $\{X_\tau\}_{\tau \in \mathbb{R}}$  is based on norms  $\|x\|_\tau = \|B^\tau x\|$ .

We consider with  $\mathcal{D} := \mathcal{D}(F) \cap \mathcal{D}(B)$  Tikhonov regularization as

$$\|F(x) - y^\delta\|^2 + \alpha \|B(x - \bar{x})\|^2 \rightarrow \min, \quad \text{subject to } x \in \mathcal{D}.$$

We assume that

$$\bar{x} \in \mathcal{D} = \mathcal{D}(F) \cap \mathcal{D}(B) \text{ and } x^\dagger \in \mathcal{D}(F), \text{ but } \underline{x^\dagger \notin \mathcal{D}(B)}$$

and that there is some  $0 < p < 1$  with  $\|B^p x^\dagger\| < \infty$ .

Moreover, let for some  $a > 0$  the nonlinearity condition

$$c_a \|B^{-a}(x - x^\dagger)\| \leq \|F(x) - F(x^\dagger)\| \leq C_a \|B^{-a}(x - x^\dagger)\| \quad \forall x \in \mathcal{D}$$

hold.

## Theorem $\triangleright$ HOFMANN/MATHÉ 2017

Under the assumptions stated we have for all  $0 < p < 1$  the convergence rate

$$\|x_{\alpha_*}^\delta - x^\dagger\| = \mathcal{O}(\delta^{\frac{p}{a+p}}) \quad \text{as } \delta \rightarrow 0$$

whenever the regularization parameter  $\alpha_*$  is chosen according to the discrepancy principle (DP).



## Overview

Classical Tikhonov regularization in Hilbert spaces requires for **finite penalty values** at  $x^\dagger$  to choose  $\alpha_*$  according to

$$\alpha_* \rightarrow 0 \quad \text{and} \quad \frac{\delta^2}{\alpha_*} \rightarrow 0 \quad \text{as} \quad \delta \rightarrow 0. \quad (\searrow)$$

Under **conditional stability** we should choose  $\alpha_*$  according to

$$\alpha_* \rightarrow 0 \quad \text{and} \quad 0 < \underline{C} \leq \frac{\delta^2}{\alpha_*} \leq \overline{C} < \infty \rightarrow 0 \quad \text{as} \quad \delta \rightarrow 0, \quad (\longrightarrow)$$

but in the context of **oversmoothing regularization** as

$$\alpha_* \rightarrow 0 \quad \text{and} \quad \frac{\delta^2}{\alpha_*} \rightarrow \infty \quad \text{as} \quad \delta \rightarrow 0. \quad (\nearrow)$$

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# On Lavrentiev regularization in Hilbert space

For a **monotone operator**  $F : \mathcal{D}(F) \subseteq X \rightarrow X$ , i.e.,

$$\langle F(x) - F(\tilde{x}), x - \tilde{x} \rangle \geq 0 \quad \forall x, \tilde{x} \in \mathcal{D}(F), \quad (\text{Mon})$$

and if  $(**)$  is a model of an inverse problem, then due to local ill-posedness it makes sense to solve a **singularly perturbed well-posed auxiliary equation**

$$F(x_\alpha^\delta) + \alpha(x_\alpha^\delta - \bar{x}) = y^\delta. \quad (\text{Lav}).$$

The analog w.r.t.  $(*)$  (cf. Example 4 in Part A) attains the form

$$Ax + \alpha(x_\alpha^\delta - \bar{x}) = y^\delta. \quad (\text{Lav}).$$

**Lavrentiev regularization** is simpler than the Tikhonov regularization, but its use is restricted to the smaller class of monotone forward operators.

## Useful assumptions

- $F : X \rightarrow X$ ,  $\mathcal{D}(F) = X$  ( $X$  separable real Hilbert space).
- $F$  is a monotone and hemicontinuous operator.

Then  $F$  is even **maximally monotone** and we have a weak-to-norm sequential closedness as

$$x_n \rightharpoonup \tilde{x} \quad \text{and} \quad F(x_n) \rightarrow z_0 \quad \Rightarrow \quad F(\tilde{x}) = z_0.$$

There occur well-posed and ill-posed situations.

The best situation of global well-posedness is characterized by **strong monotonicity**

$$\langle F(x) - F(\tilde{x}), x - \tilde{x} \rangle \geq C \|x - \tilde{x}\|^2 \quad \text{for all } x, \tilde{x} \in X,$$

with some constant  $C > 0$ , which implies the **coercivity** condition

$$\lim_{\|x\| \rightarrow \infty} \frac{\langle F(x), x \rangle}{\|x\|} = \infty.$$

Under the above assumptions  $F : X \rightarrow X$  is **surjective** due to the **Browder-Minty theorem** if the coercivity condition holds. If, moreover,  $F$  is strongly monotone, then  $F$  is **bijective** and  $F^{-1} : X \rightarrow X$  is Lipschitz continuous as

$$\|F^{-1}(y) - F^{-1}(\tilde{y})\| \leq \frac{1}{C} \|y - \tilde{y}\| \quad \text{for all } y, \tilde{y} \in X. \quad (Lip)$$

There are classes of ill-posed inverse problems with monotone  $F$  occurring in natural sciences and engineering, where  $(Lip)$  fails. Then we have operator equations  $(**)$  of the **first kind**, but the associated equations of the **second kind**

$$G(x) = y \quad \text{with} \quad G(x) := F(x) + \alpha x$$

satisfy  $(Lip)$  with  $C = \alpha$  for all  $\alpha > 0$ .

This motivates Lavrentiev regularization  $(Lav)$  for stabilizing  $(**)$ .

As an ill-posed example we consider the identification of the source term  $q$  in the elliptic boundary value problem

$$\begin{aligned} -\Delta u + \xi(u) &= q \text{ in } \mathcal{G} \\ u &= 0 \text{ on } \partial\mathcal{G} \end{aligned}$$

from measurements of  $u$  in  $\mathcal{G}$ , where  $\xi : \mathbb{R} \rightarrow \mathbb{R}$  is some Lipschitz continuously differentiable monotonically increasing function and  $\mathcal{G} \subseteq \mathbb{R}^3$  a smooth domain.

Then the corresponding **nonlinear** forward operator

$$F : X := L^2(\mathcal{G}) \rightarrow H^2(\mathcal{G}) \subseteq L^2(\mathcal{G}),$$

mapping  $q \mapsto u$ , is **monotone and hemicontinuous**.

Lavrentiev regularization is always helpful if bijectivity of  $F$  and hence Lipschitz property of  $F^{-1}$  fails, for example because coercivity fails or well-posedness occurs only in a local sense.

This is the case if  $F$  is **locally strongly monotone**

$$\langle F(x) - F(x^\dagger), x - x^\dagger \rangle \geq C \|x - x^\dagger\|^2 \quad \text{for all } x \in B_r(x^\dagger),$$

with  $C > 0$  and  $r > 0$ , or if  $F$  is **locally uniformly monotone**

$$\langle F(x) - F(x^\dagger), x - x^\dagger \rangle \geq \zeta(\|x - x^\dagger\|) \quad \text{for all } x \in B_r(x^\dagger)$$

with some index function  $\zeta$  and  $r > 0$ .

## Proposition

Let local uniform monotonicity of  $F$  in  $B_r(x^\dagger)$  hold with an index function  $\zeta$  of the form  $\zeta(t) = \theta(t)t$ ,  $t > 0$ , such that  $\theta$  is also an index function. Then we have a conditional stability estimate

$$\|x - x^\dagger\| \leq \theta^{-1}(\|F(x) - F(x^\dagger)\|) \quad \text{for all } x \in B_r(x^\dagger), \quad (\text{CSE})$$

and the operator equation (\*\*) is locally well-posed at  $x^\dagger$ .

In the special case  $\zeta(t) = Ct^2$  of strong monotonicity we find

$$\|x - x^\dagger\| \leq \frac{1}{C} \|F(x) - F(x^\dagger)\| \quad \text{for all } x \in B_r(x^\dagger). \quad (\text{CSE})$$



## One dimensional example

For  $X := \mathbb{R}$  with  $\|x\| := |x|$  we consider (\*\*) with the continuous monotone operator  $F : \mathbb{R} \rightarrow \mathbb{R}$  defined for exponents  $\kappa > 0$  as

$$F(x) := \begin{cases} -1 & \text{if } -\infty < x < -1 \\ -(-x)^\kappa & \text{if } -1 \leq x \leq 0 \\ x^\kappa & \text{if } 0 < x \leq 1 \\ 1 & \text{if } 1 < x < \infty \end{cases},$$

which, however, is not bijective and not coercive.

Then we have local ill-posedness at  $x^\dagger$  if  $x^\dagger \leq -1$  or  $x^\dagger \geq 1$ .  
At  $x^\dagger = 0$  we have local well-posedness for all  $\kappa > 0$  due to a local uniform monotonicity condition with  $\zeta(t) = t^{\kappa+1}$  such that

$$\|x - x^\dagger\| \leq \|F(x) - F(x^\dagger)\|^{1/\kappa} \quad \text{for all } x \in B_1(0). \quad (\text{CSE})$$

Lavrentiev regularization allows for linear convergence if  $\kappa = 1$ , Hölder convergence rates for  $\kappa > 1$ , and there even occurs a superlinear convergence rate at  $x^\dagger = 0$  if  $0 < \kappa < 1$ .

## Convergence rates under variational source conditions

Let us consider a variational source condition adapted to the monotonicity structure:

### Variational source condition (VSC-Lav)

*We assume to have a constant  $0 \leq \beta < 1$  and an index function  $\varphi$  such that for all  $x \in \mathcal{M}$*

$$\langle x^\dagger - \bar{x}, x^\dagger - x \rangle \leq \beta \|x^\dagger - x\|^2 + \varphi(\langle F(x) - F(x^\dagger), x - x^\dagger \rangle).$$

Here,  $\varphi$  must be an index function with  $\lim_{t \rightarrow +0} \frac{\varphi(t)}{\sqrt{t}} \geq \underline{c} > 0$ .

The next theorem is proved by the general Young inequality

$$ab \leq \int_0^a f(t) dt + \int_0^b f^{-1}(t) dt$$

for  $a, b \geq 0$  and an index function  $f$ .

## Theorem $\triangleright$ H./KALTENBACHER/RESMERITA 2016

Let  $(VSC - Lav)$  be satisfied for a solution to  $(**)$  with  $\mathcal{M}$  such that for the choice of  $\alpha > 0$  all regularized solutions  $x_\alpha^\delta$  from  $(Lav)$  belong to  $\mathcal{M}$  for sufficiently small  $\delta > 0$ . Then one has the estimate

$$\|x_\alpha^\delta - x^\dagger\|^2 \leq \frac{1}{(1-\beta)^2} \frac{\delta^2}{\alpha^2} + \frac{2}{1-\beta} \Psi(\alpha),$$

for such  $\delta > 0$ , where  $\Psi(\alpha)$  is introduced as follows:

Let  $f$  be an index function such that its antiderivative

$\tilde{f}(s) := \int_0^s f(t) dt$  satisfies the condition  $\tilde{f}(\varphi(s)) \leq s$  for  $s > 0$

and let  $G(\alpha) \geq \int_0^\alpha f^{-1}(\tau) d\tau$ . Then we set  $\Psi(\alpha) := \frac{G(\alpha)}{\alpha}$ .

## Theorem cont.

Moreover, for the a priori choice

$$\alpha_*(\delta) \sim \Theta^{-1}(\delta^2)$$

with  $\Theta(\lambda) := \lambda^2 \Psi(\lambda)$ , which satisfies

$$\alpha_*(\delta) \rightarrow 0 \quad \text{and} \quad \frac{\delta}{\alpha_*(\delta)} \rightarrow 0 \quad \text{as} \quad \delta \rightarrow 0,$$

this yields the convergence rate

$$\|\mathbf{x}_{\alpha_*(\delta)}^\delta - \mathbf{x}^\dagger\| = \mathcal{O}\left(\frac{\delta}{\Theta^{-1}(\delta^2)}\right) = \mathcal{O}\left(\sqrt{\Psi(\Theta^{-1}(\delta^2))}\right).$$

The best possible rate occurs for  $\varphi(t) \sim \sqrt{t}$  and  $\Psi(t) \sim t$  as

$$\|\mathbf{x}_{\alpha_*(\delta)}^\delta - \mathbf{x}^\dagger\| = \mathcal{O}(\delta^{\frac{1}{3}}) \quad \text{if} \quad \alpha_*(\delta) \sim \delta^{\frac{2}{3}}.$$

## Logarithmic type variational source conditions

We consider (VSC – Lav) with

$$\varphi(t) = \frac{1}{-\ln t}, \quad f(t) = e^{-\frac{1}{t}} \leq \frac{1}{t^2} e^{-\frac{1}{t}} = \varphi^{-1'}(t), \quad f^{-1}(t) = \frac{1}{-\ln t},$$

$$G(\alpha) = \alpha \frac{1}{-\ln \alpha} \geq \int_0^\alpha \frac{1}{-\ln t} dt, \quad \Theta(t) \sim t^2 \frac{1}{-\ln t}, \text{ for } \alpha, t \in (0, 1).$$

This yields the logarithmic rate  $\mathcal{O}\left(\sqrt{\Psi(\Theta^{-1}(\delta^2))}\right)$ .

Since the a priori choice  $\alpha_* = \alpha_*(\delta) \sim \Theta^{-1}(\delta^2)$  cannot be determined explicitly in this logarithmic case, a more convenient choice is  $\alpha_*(\delta) \sim \sqrt{\delta}$  which implies

$$\|x_{\alpha_*}^\delta - x^\dagger\| = \mathcal{O}\left(\frac{1}{\sqrt{-\ln(\delta)}}\right).$$

## Hölder type variational source conditions

For exponents  $\mu \in (0, \frac{1}{2}]$ , we consider (VSC – Lav) with

$$\varphi(t) = t^\mu, \quad f(t) = \frac{1}{\mu} t^{\frac{1-\mu}{\mu}}, \quad f^{-1}(t) = (\mu t)^{\frac{\mu}{1-\mu}},$$

$$\tilde{f}(s) = s^{\frac{1}{\mu}}, \quad G(\alpha) = (1 - \mu)\mu^{\frac{\mu}{1-\mu}} \alpha^{\frac{1}{1-\mu}}, \quad \Phi(t) \sim t^{\frac{2-\mu}{1-\mu}}.$$

According to the theorem this yields the Hölder rate

$$\|x_{\alpha_*}^\delta - x^\dagger\| = \mathcal{O}(\delta^{\frac{\mu}{2-\mu}}) \quad \text{if} \quad \alpha_*(\delta) \sim \delta^{\frac{2(1-\mu)}{2-\mu}}.$$

Note that  $\mathcal{O}(\delta^{\frac{\mu}{2-\mu}}) = \mathcal{O}(\delta^{\frac{\rho}{\rho+1}})$  for  $\mu := \frac{2\rho}{2\rho+1}$ ,  $0 < \rho \leq \frac{1}{2}$ .

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- 3 Oversmoothing regularization in Hilbert scales
- 4 On Lavrentiev regularization in Hilbert space
- 5 References

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