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## Sharp converse results for the regularization error using distance functions

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## Abstract

In the analysis of ill-posed inverse problems the impact of solution smoothness on accuracy and convergence rates plays an important role. For linear ill-posed operator equations in Hilbert spaces and with focus on the linear regularization schema we will establish relations between the different kinds of measuring solution smoothness in a point-wise or integral manner. In particular we discuss the interplay of distribution functions, profile functions that express the regularization error, index functions generating source conditions, and distance functions associated with benchmark source conditions. We show that typically the distance functions and the profile functions carry the same information as the distribution functions, and that this is not the case for general source conditions. The theoretical findings are accompanied with examples exhibiting applications and limitations of the approach.

## 1 Introduction

The stable approximate solution of ill-posed inverse problems that can be formulated as linear operator equations

$$Ax = y \tag{1}$$

with an injective and bounded linear operator  $A : X \rightarrow Y$  mapping between Hilbert spaces  $X$  and  $Y$  and possessing a non-closed range  $\mathcal{R}(A)$  requires regularization, since under the above assumptions the (formal) solution mapping  $y \mapsto x = A^{-1}y$  exists for each  $y \in \mathcal{R}(A)$ , however, this dependence is discontinuous. Precisely, this is the *Moore–Penrose inverse*  $A^\dagger$ , which can be defined in a more general context as a densely defined (unbounded) operator. Therefore, the solution theory aims at replacing the discontinuous mapping  $A^\dagger$  by a family of continuous (bounded) regularization operators  $R_\alpha : Y \rightarrow X$  indexed by the regularization parameter  $\alpha > 0$ . This approach is common since the early days of the theory of ill-posed equations, and a seminal treatise along these lines is [4]. The goal is to design families  $R_\alpha$  with the property that  $R_\alpha y \rightarrow A^\dagger y$  as  $\alpha \rightarrow 0$  whenever  $y \in \mathcal{R}(A)$ .

### Regularization error

The deviation of  $R_\alpha y$  from  $A^\dagger y$  is called *regularization error*, and we have for  $y = Ax$  that

$$f_x(\alpha) := \|A^\dagger y - R_\alpha y\| = \|A^\dagger Ax - R_\alpha Ax\| = \|x - R_\alpha Ax\|, \quad \alpha > 0.$$

In most cases the family  $R_\alpha$  is given by a family of piece-wise continuous real functions  $g_\alpha$ . By noticing that  $A^\dagger = (A^*A)^\dagger A^*$  we assign

$$R_\alpha := g_\alpha(A^*A)A^*, \quad (2)$$

where spectral calculus allows to extend the real functions to operator valued ones. In terms of the family  $g_\alpha$  the regularization error reads as  $f_x(\alpha) = \|x - g_\alpha(A^*A)A^*Ax\|$ , and this function was called *profile function* in [12]. A crucial observation in this context is that the decay rate of  $f_x(\alpha)$  as  $\alpha \rightarrow 0$  depends on smoothness properties of the solution element  $x$ . The quantitative relation between smoothness properties of  $x$ , given in terms of spectral information, called distribution function below, and the decay of the profile function for a given regularization was first emphasized in [18, 19]. Here we take a more general point of view, and we shall subsume such properties as different kinds of *solution smoothness* of  $x$  with respect to the operator  $H := A^*A$ .

### Distribution function

The *point-wise characteristics* of the solution smoothness of  $x$  with respect to the spectrum of  $H$ , which contains the complete spectral information of the element, exploits the (right-continuous version of the) spectral *distribution function*

$$F_x^2(t) := \|E_t x\|^2 := \langle \chi_{(0,t]}(H)x, x \rangle = \|\chi_{(0,t]}(H)x\|^2, \quad 0 < t < \infty, \quad (3)$$

or the equivalent representation as  $F_x^2(t) = \int_0^t d\|E_s x\|^2$  for  $t > 0$ . Above, we let  $\chi_{(0,t]}$  be the characteristic function of the interval  $(0, t]$ , and  $E_t = E_t(H)$ ,  $0 \leq t \leq \|H\|$ , be the spectral resolution of the operator  $H$ , i.e., for any (bounded measurable) real function  $h$  we have that

$$\|h(H)x\|^2 = \int_0^{\|H\|} h^2(t) d\|E_t x\|^2,$$

we refer to [20, Chapt. 12] for details on spectral theory of bounded self-adjoint operators in Hilbert space. We should note here that by definition the function  $F_x$  is non-decreasing and right-continuous with  $\lim_{t \rightarrow 0} F_x(t) = 0$ . The latter property is a consequence of zero being an accumulation point of the spectrum of  $H$ . The distribution function may have jumps at the points of the spectrum of  $H$ . In particular, for compact operators  $H$  it will be piece-wise constant. Moreover, an immediate and well-known observation is the following fact.

**Fact 1** ([19, Prop. 2.3]). *Given  $x \in X$ , the increase  $F_x(\alpha) = \mathcal{O}(\alpha^\kappa)$  as  $\alpha \rightarrow 0$ , for some  $0 < \kappa < 1$ , implies that  $x \in \mathcal{R}(H^\nu)$  for every  $0 < \nu < \kappa$ .*

### General smoothness

The smoothness of solutions  $x \in X$  for ill-posed equations can be expressed mathematically in different ways. The most traditional form is characterized by *source conditions*. In its general version it is assumed that there is a non-decreasing continuous

function  $\psi: (0, \|H\|] \rightarrow \mathbb{R}^+ = (0, \infty)$  with  $\lim_{t \rightarrow 0} \psi(t) = 0$  that we will call *index function*, for which

$$x = \psi(H)v, \quad (4)$$

with some *source element*  $v \in X$ . Evidently, (4) can be rewritten as  $x \in \mathcal{R}(\psi(H))$ . Recent results assert that for every  $x \in X$  there is an index function  $\psi$  and a source element  $v \in X$  satisfying (4), see [17, 13].

Although the concept of source conditions (4) proved to be useful in the error analysis of ill-posed operator equations, the point-wise characteristics (3) contains more precise information of the smoothness of  $x$  with respect to the operator  $H$ . We exhibit this in case of Tikhonov regularization, where the family  $R_\alpha$  is given for  $y = Ax$  by  $R_\alpha y = (H + \alpha I)^{-1} Hx$ , and hence the corresponding profile function is

$$f_x(\alpha) := \alpha \|(H + \alpha I)^{-1} Hx\|, \quad 0 < \alpha < \infty, \quad (5)$$

which is an index function, well-defined and increasing for all  $\alpha > 0$ . A celebrated converse result establishes a one-to-one correspondence between the distribution and the profile functions in moderate cases.

**Fact 2** ([19, Thm. 2.1]). *Given  $x \in H$ , we have that  $f_x(\alpha) = \mathcal{O}(\alpha^\kappa)$ , for some  $0 < \kappa < 1$ , if and only if  $F_x(\alpha) = \mathcal{O}(\alpha^\kappa)$  as  $\alpha \rightarrow 0$ .*

The key to the proof of Fact 2 is the following result. It uses the family  $w_\alpha \in X$  of elements given as  $w_\alpha := \chi_{(\alpha, \|H\|]}(H)H^{-1}x$ ,  $\alpha > 0$ , which will prove useful later, see (23) for the general construction.

**Lemma 1** ([19], or the original study [5]). *Let  $w_\alpha$  be defined as above. The behavior  $F_x(\alpha) = \mathcal{O}(\alpha^\kappa)$ , for some  $0 < \kappa < 1$ , yields  $\alpha \|w_\alpha\| = \mathcal{O}(\alpha^\kappa)$  as  $\alpha \rightarrow 0$ .*

In contrast, by using power type source conditions we only have that  $f_x(\alpha) = \mathcal{O}(\alpha^\nu)$  provided that  $x \in \mathcal{R}(H^\nu)$ , again whenever  $0 < \nu < 1$ . This of course is less accurate than the assertion in Fact 2, and it is mentioned in [19] that there are  $x \in \mathcal{R}(H^\nu)$  for which  $f_x(\alpha) = o(\alpha^\nu)$ . We will return to this discussion in Corollary 2.

## Distance function

In the past years, with the study [11] as well as the subsequent studies in [1, 2, 8, 12, 14] and [7], the lack of information occurring when general source conditions are used, was circumvented by using *distance functions*

$$d_\psi(R) := \inf \{ \|x - \psi(H)v\| : \|v\| \leq R \}, \quad 0 \leq R < \infty. \quad (6)$$

Whenever  $x \notin \mathcal{R}(\psi(H))$  such distance functions  $d_\psi(R)$ , which are positive, decreasing, convex and continuous for all  $0 \leq R < \infty$ , moreover tending to zero as  $R \rightarrow \infty$ , measure the degree of violation of  $x$  with respect to the benchmark source condition (4). Most of the mentioned properties of the function  $d_\psi$  are given in [11, Lemma 2.5 and

its proof]. For completeness we indicate the proof of the convexity. If for  $0 < R, S < \infty$  the minimizers for  $d_\psi(R)$  and  $d_\psi(S)$  are called  $v^R, v^S$ , respectively, then  $\|(v^R + v^S)/2\| \leq (R + S)/2$ , and  $d_\psi((R + S)/2) \leq \|x - \psi(H)((v^R + v^S)/2)\| \leq \frac{1}{2}(d_\psi(R) + d_\psi(S))$ .

**Remark 1.** *In the previous study [12] the authors used the distance function  $\rho_\psi$  defined as*

$$\rho_\psi(t) := \inf \{ \|tx - \psi(H)v\| : \|v\| \leq 1 \}, \quad t > 0,$$

*instead of  $d_\psi(R)$ ,  $R > 0$ . The relation*

$$d_\psi(R) = R\rho_\psi(1/R), \quad R > 0, \quad (7)$$

*is evident from the definition. We know from [12, Lemma 5.3] that the functions  $\rho_\psi$  as well as the function  $t \mapsto \rho_\psi(t)/t$ ,  $t > 0$ , are increasing index functions mapping  $(0, \infty)$  onto itself. Some more insight gives [6, Remark 1] indicating that the transformations  $\rho_\psi \mapsto d_\psi$  according to (7) and its inverse  $d_\psi \mapsto \rho_\psi$  are involutions that preserve convexity.*

Here we confine to the usage of  $d_\psi$  as distance function. In the sequel, for short we call the index function  $\psi$  in  $d_\psi$  *benchmark*. If the decay of  $d_\psi(R) \rightarrow 0$  as  $R \rightarrow \infty$  is slow, then  $x$  strongly violates the benchmark source condition, whereas a fast decay corresponds to a weak violation. The distance function approach, also called *method of approximate source conditions* is a third way of expressing solution smoothness. To see this we recall the following result, where as in the sequel we write  $d_\nu(R)$  for short if we mean  $d_\psi(R)$  with monomial benchmark  $\psi(t) = t^\nu$ ,  $\nu > 0$ .

**Fact 3** ([2, Theorem 3.2], [12, §5.2]). *Smoothness of the form  $x \in \mathcal{R}(H^\kappa)$  with  $0 < \kappa < \nu$  implies for the distance function with monomial benchmark  $\psi(t) = t^\nu$ ,  $t > 0$ , a decay rate  $d_\nu(R) = \mathcal{O}\left(R^{-\frac{\kappa}{\nu-\kappa}}\right)$  as  $R \rightarrow \infty$ .*

This assertion will be improved as a result of our analysis in Corollary 2.

At some place we will also use a well-known dual formulation of the distance function

$$d_\psi(R) = \sup \{ \langle x, v \rangle - R\|\psi(H)v\|, \quad \|v\| \leq 1 \}, \quad R > 0, \quad (8)$$

as an alternative to (6), which can be derived from the concept of Fenchel duality, see [22, Sec. 2.7].

## Outline

In this study we will establish relations between the different kinds of measuring solution smoothness, in particular between distance functions  $d_\psi(R)$ , the distribution function  $F_x(\alpha)$ , and a regularization error  $f_x(\alpha)$  in the context of the linear regularization schema introduced in Section 2. The basic question is, whether the distance function carries the same spectral information as the distribution function. General lower and

upper bounds are given in Section 3. The proof of the main general result will be given in Appendix A. For the specific case of power type behavior we even show a one-to-one correspondence between the associated exponents in Section 4, which answers the above basic question in the affirmative. We also highlight the theoretical results by providing examples which show the application and limitations of our findings in Section 5. In particular we discuss relations to other results in this direction, which were previously obtained by several authors.

## 2 The linear regularization schema

Before formulating the main results in Sections 3 and 4 we shall recall the concept of a regularization schema as given in [12], where only linear regularization operators  $R_\alpha$  are under consideration. If the family of operators  $R_\alpha$  was obtained from some family of generator functions  $g_\alpha$ , see (2), then the regularization error constitutes as

$$x - R_\alpha Ax = (I - g_\alpha(H)H)x,$$

and we associate to the function  $g_\alpha$  the residual (bias) function  $r_\alpha(t) := 1 - tg_\alpha(t)$ ,  $0 < t \leq \|H\|$ .

**Definition 1.** A family of piece-wise continuous functions  $g_\alpha(t)$  is called a regularization if  $\lim_{\alpha \rightarrow 0} r_\alpha(t) = 0$  as  $\alpha \rightarrow 0$  for all  $0 < t \leq \|H\|$  and the following estimates hold for all  $0 < t \leq \|H\|$ ,  $0 < \alpha \leq \bar{\alpha}$ , and with constants  $\gamma_0, \gamma_1$  and  $\gamma_*$ :

- 1  $\sqrt{t} |g_\alpha(t)| \leq \gamma_*/\sqrt{\alpha}$ ,
- 2  $|r_\alpha(t)| \leq \gamma_1$ ,
- 3  $t |g_\alpha(t)| \leq \gamma_0$ .

Using the notation of index functions introduced in the initial section we say that an index function  $\varphi$  is a *qualification* of the regularization generated by  $g_\alpha$  with constant  $\gamma$  if

$$|r_\alpha(t)| \varphi(t) \leq \gamma \varphi(\alpha), \quad 0 < t \leq \|H\|, \quad 0 < \alpha \leq \bar{\alpha}.$$

A standard account on the linear regularization schema is [3].

### Regularization from a single function

In many cases the regularization family  $g_\alpha$  can be obtained from a single function, say  $g: (0, \infty) \rightarrow \mathbb{R}$ , accompanied with the function  $r(t) := 1 - tg(t)$ ,  $t > 0$ , where we refer to [21, § 2.3] and to the German textbook [15]. It is easy to see that such function  $g$  gives rise to a regularization by exploiting the dilatation procedure

$$g_\alpha(t) := \frac{1}{\alpha} g\left(\frac{t}{\alpha}\right), \quad t > 0, \quad \alpha > 0,$$

provided that  $g$  obeys

$$4 \quad \sqrt{t}|g(t)| \leq \gamma_*, \quad t > 0,$$

$$5 \quad |r(t)| \leq \gamma_1, \quad t > 0, \text{ and}$$

$$6 \quad t|g(t)| \leq \gamma_0, \quad t > 0.$$

Notice that by construction we have that  $r_\alpha(t) = r(t/\alpha)$  for  $t, \alpha > 0$ , and that  $r_\alpha$  is well defined for every  $\alpha > 0$ , i.e.,  $\bar{\alpha} := \infty$ .

### Profile functions

With respect to any given regularization  $g_\alpha$  we are interested in the profile function (regularization error), which is seen to equal

$$f_x(\alpha) := \|r_\alpha(H)x\|, \quad 0 < \alpha \leq \bar{\alpha}. \quad (9)$$

We shall assume that  $\|r_\alpha(H)x\|$  is a non-decreasing function in  $\alpha$  for simplicity. This is the case if the underlying regularization  $g_\alpha$  is such that the function  $\alpha \mapsto |r_\alpha(t)|$ ,  $0 < \alpha \leq \bar{\alpha}$ , is non-decreasing, which is always satisfied for regularization from a single function  $g$  with non-increasing  $r$ . If the regularization  $g_\alpha$  consists of continuous functions, then  $f_x$  is continuous and hence with  $\lim_{\alpha \rightarrow 0} f_x(\alpha) = 0$  an index function.

**Example 1** (Tikhonov regularization). *This regularization method with the continuous functions  $g_\alpha(t) = 1/(t + \alpha)$  and  $r_\alpha(t) = \alpha/(t + \alpha)$  is obtained from  $g(t) := 1/(t + 1)$ , with  $r(t) = 1/(t + 1)$ . We observe that the function  $r$  is decreasing and moreover that  $\frac{1}{2} = r(1) \leq r(t) \leq 1$ ,  $0 < t \leq 1$ . Any concave index function, in particular any linear, is a qualification of the method.*

**Example 2** (spectral cut-off). *Another important regularization method is spectral cut-off, where  $g_\alpha(t) = 1/t$ ,  $t > \alpha$ , and  $g_\alpha(t) = 0$ , otherwise. This function with a jump at the point  $t = \alpha$  corresponds to  $g(t) = 1/t$ ,  $t > 1$ , and  $g(t) = 0$ , otherwise. The residual function is  $r(t) = \chi_{(0,1]}(t)$ , the characteristic function of the interval  $(0, 1]$ , again a non-increasing function in  $t$ . Spectral cut-off has arbitrary index functions as qualification. We notice the important observation that for spectral cut-off we have that*

$$\|r_\alpha(H)x\| = F_x(\alpha), \quad 0 < \alpha \leq \|H\|. \quad (10)$$

One important observation from [18] relates the distribution function  $F_x$  to the profile function  $f_x$  as follows.

**Proposition 1.** *For every regularization  $g_\alpha$  there is a constant  $0 < c \leq 1$  such that*

$$F_x(c\alpha) \leq 2f_x(\alpha), \quad 0 < \alpha \leq \|H\|.$$

*Proof.* We first notice that by item (i) of Definition 1 with constant  $\gamma_*$  we find a  $0 < c \leq 1$  for which  $|r_\alpha(t)| \geq 1/2$  whenever  $0 < t \leq c\alpha$ . Precisely, setting  $c := \min \left\{ \frac{1}{4\gamma_*^2}, 1 \right\}$  we have

$$|r_\alpha(t)| \geq 1 - t |g_\alpha(t)| \geq 1 - \gamma_* \sqrt{t/\alpha} \geq 1 - \gamma_* \sqrt{c} \geq 1/2.$$

Then we bound

$$\begin{aligned} f_x^2(\alpha) &= \int_0^{\|H\|} |r_\alpha(t)|^2 dF_x^2(t) \\ &\geq \int_0^{c\alpha} |r_\alpha(t)|^2 dF_x^2(t) \geq \frac{1}{4} \int_0^{c\alpha} dF_x^2(t) = \frac{1}{4} F_x^2(c\alpha), \end{aligned}$$

from which the assertion follows.  $\square$

### 3 General results

As already mentioned, both functions, the regularization error  $f_x$  and the distance function  $d_\psi$  reflect smoothness of  $x$  in the sense of certain spectral properties of the involved element with respect to  $H$ .

We shall establish for  $x \notin \mathcal{R}(\psi(H))$  a one-to-one correspondence between distance functions  $d_\psi(R)$  and profile functions  $f_x(\alpha)$ , for an appropriate relation between  $R$  and  $\alpha$ . To this end let

$$\Phi_\psi(R) := \frac{d_\psi(R)}{R} = \rho_\psi \left( \frac{1}{R} \right), \quad R > 0. \quad (11)$$

Since  $d_\psi(R)$  is decreasing and continuous for  $0 \leq R < \infty$ , the function  $\Phi_\psi(R)$  is even a strictly decreasing continuous function for all positive  $R$  mapping  $(0, \infty)$  onto itself. By the above reasoning the equation

$$\Phi_\psi(R) = \psi(\alpha) \quad (12)$$

has a unique solution  $R = R(\alpha)$  for each  $\alpha \in (0, \bar{\alpha}]$ .

The following upper bound was derived by using the function  $\rho_\psi$  in [12, Thm. 5.5]. For the convenience of the reader we recall the proof within the present context, here.

**Proposition 2.** *If the regularization  $g_\alpha$  has qualification  $\psi$  with constant  $\gamma$  and if  $x \notin \mathcal{R}(\psi(H))$ , then*

$$\|r_\alpha(H)x\| \leq (\gamma + \gamma_1) d_\psi \left( \Phi_\psi^{-1}(\psi(\alpha)) \right), \quad 0 < \alpha \leq \bar{\alpha}.$$

*Proof.* Let  $R = R(\alpha)$  be given by solving equation (12), and let  $v$  be the minimizer of the distance function  $d_\psi(R)$ . Then

$$\begin{aligned} \|r_\alpha(H)x\| &\leq \|r_\alpha(H)(x - \psi(H)v)\| + \|r_\alpha(H)\psi(H)v\| \\ &\leq \gamma_1 d_\psi(R) + \gamma \psi(\alpha) R \\ &= (\gamma_1 + \gamma) \max \{ d_\psi(R), \psi(\alpha) R \}, \end{aligned}$$

which allows us to complete the proof.  $\square$

Our goal is to establish a converse to the bound from Proposition 2, and we shall distinguish a *low-benchmark* and a *high-benchmark* case, respectively. The main general result is the following Theorem 1. We establish essential cross connections between distance and profile functions. This will be helpful for discussing the power type case in the subsequent section. Because the proof of this theorem is rather technical, we postpone it to Appendix A.

**Theorem 1.** *Let  $\psi$  be an index function and  $x \notin \mathcal{R}(\psi(H))$ . Moreover let  $g_\alpha$  be a regularization.*

**low-benchmark** *If the function  $t \mapsto \psi^2(t)/t$  is non-increasing, or*

**high-benchmark** *if the function  $t \mapsto \psi^2(t)/t$  is non-decreasing, and the regularization is obtained from a single function  $g$  with non-increasing  $r$ ,  $|r(1)| > 0$ ,*

*then*

$$d_\psi \left( 2 \max \{ \gamma_0, \gamma_* \} \Phi_\psi^{-1}(\psi(\alpha)) \right) \leq C \|r_\alpha(H)x\|, \quad 0 < \alpha \leq \bar{\alpha}, \quad (13)$$

*with constant  $C = 1$  in the low-benchmark case and  $C = \gamma_1/|r(1)|$  in the high benchmark case.*

*If the regularization has qualification  $\psi$ , then*

$$\|r_\alpha(H)x\| \leq 2 \max \{ \gamma, \gamma_1 \} d_\psi \left( \Phi_\psi^{-1}(\psi(\alpha)) \right), \quad 0 < \alpha \leq \bar{\alpha}.$$

**Remark 2.** *In case that the distance function  $d_\psi$  does not decay too quickly we can take out the leading constant  $2 \max \{ \gamma_0, \gamma_* \}$  on the left in (13). Precisely, a non-decreasing function, say  $h: (0, \infty) \rightarrow (0, \infty)$ , is said to obey a  $\Delta_2$ -condition if there is a constant  $C_2 \geq 1$  such that  $h(2t) \leq C_2 h(t)$ ,  $t > 0$ . This restricts the growth rate of the function  $h$  to be sub-exponential. Within the present context this specifies to the following. If the function  $1/d_\psi$  obeys a  $\Delta_2$ -condition, in particular, if  $d_\psi$  decays sub-exponentially, then there is a constant  $0 < c \leq 1$  such that*

$$cd_\psi \left( \Phi_\psi^{-1}(\psi(\alpha)) \right) \leq d_\psi \left( 2 \max \{ \gamma_0, \gamma_* \} \Phi_\psi^{-1}(\psi(\alpha)) \right)$$

*In this case the upper and lower bounds for the profile function in Theorem 1 coincide up to constants.*

We highlight one specific instance of Theorem 1, when we choose spectral cut-off as regularization. In this case the profile function  $f_x(\alpha)$  has a clear interpretation, see the representation (10) in Example 2.

**Corollary 1.** *For an arbitrary index function  $\psi$  we have that*

$$d_\psi \left( 2 \Phi_\psi^{-1}(\psi(\alpha)) \right) \leq F_x(\alpha) \leq 2d_\psi \left( \Phi_\psi^{-1}(\psi(\alpha)) \right), \quad 0 < \alpha < \infty.$$

*Proof.* The proof of the lower bound is the same as for the high-benchmark case of Theorem 1 applied to the spectral cut-off schema, but instead of Lemma 7 here one has to use Remark 5, see Appendix A below. For the upper bound we apply Theorem 1 again to spectral cut-off and take into account that  $\gamma_* = \gamma_0 = 1$ .  $\square$

We close this section with general remarks on techniques for bounding distance functions. The ‘standard’ way for establishing upper bounds is to use the family  $w_\alpha := \chi_{(\alpha, \|H\|]}(H)\psi(H)^{-1}x \in X$ ,  $0 < \alpha \leq \bar{\alpha}$ , since here  $\|x - \psi(H)w_\alpha\| = F_x(\alpha)$ . This family adapts the choice used for proving Lemma 1. Upper bounds are then obtained from this, since  $d_\psi(\|w_\alpha\|) \leq F_x(\alpha)$ , such that upper bounds for  $\|w_\alpha\|$ , typically of the order  $F_x(\alpha)/\psi(\alpha)$ , yield upper bounds for  $d_\psi$  by monotonicity.

It is harder to establish lower bounds. Two ways are worth mentioning. First, Corollary 1 yields, with  $R := \Phi_\psi^{-1}(\psi(\alpha))$  that

$$F_x(\psi^{-1}(\Phi_\psi(R))) \leq 2d_\psi(R), \quad R \geq R_0.$$

In the ‘moderate’ cases, if the smoothness given by the distribution  $F_x$  is far enough from the benchmark  $\psi$ , this yields lower bounds for  $d_\psi$ , see e.g. the reasoning in Example 3(c), below. However, in extremal cases, if the actual smoothness is close to the benchmark, then one apparently must rely on the Fenchel duality (8), and one must guess an appropriate choice of  $v \in X$ , often from some parametric family  $u_\alpha$ ,  $\alpha > 0$ . Here we point at the corresponding choice in Example 3(a).

## 4 Power-type behavior

Here we discuss consequences of Corollary 1 for power-type functions of growth and decay rates expressing the solution smoothness. In general such functions correspond to moderate smoothness situations; other behavior is possible, see Example 4. In the following result we consider power-type benchmark smoothness  $\psi_v(t) := t^v$  for some fixed  $v > 0$  with the corresponding distance functions  $d_v(R) := d_{\psi_v}(R)$  and associated quotient functions  $\Phi_v(R) := \frac{d_v(R)}{R}$ .

Since the results will be asymptotic in nature, we recall the following notion and notation.

**Definition 2.** *Suppose that  $f, g: (0, a) \rightarrow (0, \infty)$  are real functions. Then we denote  $f = \mathcal{O}(g)$  as  $t \rightarrow 0$ , if there are constants  $C < \infty$  and  $0 < \bar{t} \leq a$  such that  $f(t) \leq Cg(t)$ ,  $0 < t < \bar{t}$ . We denote  $f \asymp g$  if  $f = \mathcal{O}(g)$  and  $g = \mathcal{O}(f)$ . Finally, we denote  $f = o(g)$  if  $f(t)/g(t) \rightarrow 0$  as  $t \rightarrow 0$ .*

The above behavior is concerned with functions defined in a right neighborhood of zero, but similar notion and notation applies for positive functions  $f(R), g(R)$ ,  $R \in [M, \infty)$ ,  $M > 0$ , when the limit case  $R \rightarrow \infty$  is under consideration.

The following result extends and reproves [18, Thm. 2.2] by using distance functions. Its proof, as given here, will use the results obtained so far.

**Theorem 2.** Let  $0 < \kappa < \nu$ . For  $x \in X$  satisfying the condition  $x \notin \mathcal{R}(H^\nu)$  the following assertions are equivalent:

- 1 The distribution function for  $x$  behaves like  $F_x(\alpha) \asymp \alpha^\kappa$  as  $\alpha \rightarrow 0$ .
- 2 The distance function for  $x$  behaves like  $d_\nu(R) \asymp R^{-\frac{\kappa}{\nu-\kappa}}$  as  $R \rightarrow \infty$ .
- 3 For an arbitrary regularization  $g_\alpha$ , which has  $\varphi(t) = t^\nu$ ,  $t > 0$ , as qualification, the profile function  $f_x(\alpha) := \|r_\alpha(H)x\|$  for  $x$  behaves like  $f_x(\alpha) \asymp \alpha^\kappa$  as  $\alpha \rightarrow 0$ .

The assertions remain valid when replacing  $\asymp$  by either big-o ‘ $\mathcal{O}$ ’ or little-o ‘ $o$ ’.

We shall see in the examples 3 and 5, given in Section 5, that power type behavior actually occurs for compact and non-compact operators  $H$ .

To prove Theorem 2 we start with the following preliminary result.

**Lemma 2.** Let  $0 < \kappa < \nu$ . We have that  $d_\nu(R) = \mathcal{O}(R^{-\kappa/(\nu-\kappa)})$  as  $R \rightarrow \infty$  implies that  $d_\nu(\Phi_\nu^{-1}(\alpha)) = \mathcal{O}(\alpha^\kappa)$  as  $\alpha \rightarrow 0$ . The assertion remains true if we replace big-O ‘ $\mathcal{O}$ ’ by little-o ‘ $o$ ’.

*Proof.* The assumption on  $d_\nu$  implies that  $\Phi_\nu(R) = \mathcal{O}(R^{-1/(\nu-\kappa)})$ , thus there are  $0 < C < \infty$  and  $0 < R_0 < \infty$  for which  $\Phi_\nu(R) \leq CR^{-1/(\nu-\kappa)}$  whenever  $R \geq R_0$ . This yields that  $\Phi_\nu^{-1}(CR^{-1/(\nu-\kappa)}) \geq R$ , by monotonicity. We assign  $\alpha := CR^{-1/(\nu-\kappa)}$  and thus have that  $\Phi_\nu^{-1}(\alpha) \geq (\alpha/C)^{\kappa-\nu}$ . Therefore, using the monotonicity of  $d_\nu$  we see that

$$d_\nu(\Phi_\nu^{-1}(\alpha)) \leq d_\nu((\alpha/C)^{\kappa-\nu}) \leq C((\alpha/C)^{\kappa-\nu})^{-\kappa/(\nu-\kappa)} = \mathcal{O}(\alpha^\kappa).$$

The assertion for little-o is along the same lines, but more tedious. □

*Proof of Theorem 2.* We first proof the equivalence of items (I) and (II) in either of the asymptotic regimes. By Lemma 2 and Corollary 1 the order of magnitude of  $d_\nu$  yields the corresponding order for  $F_x$ . For the converse we use the approximation  $w_\alpha := \chi_{(\alpha, \|H\|]}(H)H^{-\nu}x$ . By substituting the operator  $H$  in Lemma 1 with  $H^\nu$  we see that  $F_x(\alpha) = \mathcal{O}(\alpha^\kappa)$  yields an inequality of the form  $\alpha^\nu \|w_\alpha\| = \mathcal{O}(\alpha^\kappa)$ . Therefore, there is  $0 < C < \infty$  such that

$$d_\nu(C\alpha^{\kappa-\nu}) \leq d_\nu(\|w_\alpha\|) \leq \|x - H^\nu w_\alpha\| = F_x(\alpha).$$

By letting  $\alpha := (R/C)^{-1/(\nu-\kappa)}$  we obtain that

$$d_\nu(R) \leq F_x\left((R/C)^{-1/(\nu-\kappa)}\right) = \mathcal{O}(R^{-1/(\nu-\kappa)}) \text{ as } R \rightarrow \infty.$$

Again, for little-o the reasoning is similar.

Finally, suppose that  $F_x(\alpha) \asymp \alpha^\kappa$ . Plainly, by the first part of the proof this implies that  $d_\nu(R) = \mathcal{O}(R^{-\kappa/(\nu-\kappa)})$ . If it were true that  $d_\nu(R) = o(R^{-\kappa/(\nu-\kappa)})$ , then this would imply, again by the beginning of the proof, that  $F_x(\alpha) = o(\alpha^\kappa)$ , contradicting the assumption.

Similar applies by assuming that  $d_\nu(R) \asymp R^{-\kappa/(v-\kappa)}$ , and this shows the equivalence of items (I) and (II) in either of the asymptotic regimes.

Next, the asymptotics  $d_\nu(R) = \mathcal{O}(R^{-\frac{\kappa}{v-\kappa}})$  implies that  $f_x(\alpha) = \mathcal{O}(\alpha^\kappa)$ , by Lemma 2 and Proposition 2. The same applies for little-o 'o'.

Finally, if  $F_x(\alpha) \asymp \alpha^\kappa$  then, by the first part of the proof, and by Proposition 2, this implies that  $f_x(\alpha) = \mathcal{O}(\alpha^\kappa)$ . From Proposition 1 we deduce that  $\alpha^\kappa = \mathcal{O}(f_x(\alpha))$  in this case.

Since Item (I) is a special case of Item (III), the proof is complete.  $\square$

This shows that in power type situations, the decay rates of the distance function as well as of the profile function both relate to the 'point-wise' behavior of the spectrum of  $x$  with respect to  $H$ , given by  $F_x$ . In contrast smoothness expressed by source conditions (4), i.e., requiring that  $x$  is in the range of some index function of the operator  $H$ , considers the spectrum of  $H$  in an 'integral' form only. We dwell on this, and highlight some relations to the study [2].

**Lemma 3.** *Let  $\psi$  be an arbitrary index function. If  $x \in \mathcal{R}(\psi(H))$  then  $F_x(\alpha) = o(\psi(\alpha))$  as  $\alpha \rightarrow 0$ .*

*Proof.* First, by the injectivity of the operator  $H$  we have  $F_x^2(\alpha) \rightarrow 0$  as  $\alpha \rightarrow 0$ . Next, by assumption the function  $1/\psi^2 \in L_1((0, \infty), dF_x^2)$ , and hence the measure  $dG_x(t) := 1/\psi^2(t) dF_x^2(t)$  is absolutely continuous with respect to  $dF_x^2$ . Thus for each  $\varepsilon > 0$  there is  $\delta > 0$  such that  $\int_{\mathcal{A}} dF_x^2(t) \leq \delta$  implies that  $\int_{\mathcal{A}} dG_x(t) \leq \varepsilon$  for all Borel sets  $\mathcal{A} \subset (0, \infty)$ . In particular, if  $\alpha_\delta$  is small enough such that  $F_x^2(\alpha_\delta) \leq \delta$ , then

$$\varepsilon \geq \int_0^{\alpha_\delta} \frac{1}{\psi^2(t)} dF_x^2(t) \geq \frac{1}{\psi^2(\alpha_\delta)} F_x^2(\alpha_\delta),$$

from which the proof can be completed.  $\square$

In the light of Theorem 2 this yields the following strengthening and generalization of [2, Thms. 3.1 & 3.2].

**Corollary 2.** *Let  $0 < \eta < v$ , and suppose that  $x \in \mathcal{R}(H^\eta)$  but  $x \notin \mathcal{R}(H^v)$ . Then*

$$d_\nu(R) = o(R^{-\eta/(v-\eta)}) \quad \text{as } R \rightarrow \infty, \quad (14)$$

and

$$f_x(\alpha) = o(\alpha^\eta) \quad \text{as } \alpha \rightarrow 0, \quad (15)$$

for an arbitrary regularization  $g_\alpha$ , which has  $\varphi(t) = t^v$ ,  $t > 0$ , as qualification.

*Proof.* By Lemma 3 we deduce from the assumption that  $F_x(\alpha) = o(\alpha^\eta)$ . Now, Theorem 2, in the little-o 'o' cases, yields both (14) and (15).  $\square$

This explains that optimal decay rates for the regularization error based on general smoothness in terms of index functions  $\psi$  in (4) cannot be obtained, as this was already mentioned after Lemma 1. This can also be seen from [17]. In addition, we refer to the discussion on lower bounds in [12], and also the study [16, § 4].

## 5 Examples

We turn to discussing examples in connection with the theoretical results. The first example is based on a non-compact multiplication operator, in particular it points at the limitations of the bounds from Theorem 1 when the distance functions do not obey some  $\Delta_2$ -condition.

**Example 3.** *The authors in [9, 11] have discussed injective multiplication operators in  $L_2(0, 1)$  with non-closed range, and [9, Example 4.6], or [11, § 3, Example 2], will guide us, here. Precisely, we focus on the non-compact operator*

$$[Hx](s) := m(s)x(s), \quad 0 < s < 1, \quad (16)$$

with multiplier function  $m \in L^\infty(0, 1)$  possessing an essential zero. Below, we shall restrict to cases where  $m$  is strictly increasing and continuous for  $0 < s < 1$  with limit conditions  $\lim_{t \rightarrow 0} m(t) = 0$  and  $\lim_{t \rightarrow 1} m(t) = 1$ . It is evident for such multiplication operators that

$$F_x^2(t) = \int_{\{s \in (0, 1) : 0 < m(s) \leq t\}} x^2(s) ds = \int_0^{m^{-1}(t)} x^2(s) ds, \quad 0 < t \leq 1. \quad (17)$$

This distribution function is continuous with  $F_x(1) = \|x\|$  and can be extended continuously as  $F_x(t) = \|x\|$  for all  $1 < t < \infty$ . So we can easily derive the following representation for an arbitrary function  $h \in L_2((0, \infty), dF_x^2)$ , namely that

$$\|h(H)x\|^2 = \int_0^\infty h^2(t) dF_x^2(t) = \int_0^1 h^2(m(s))x^2(s) ds. \quad (18)$$

We now restrict to the case that  $x \equiv 1$  is a constant function, hence  $F_x^2(t) = m^{-1}(t)$ ,  $0 < t \leq 1$ . We notice that  $1 \in \mathcal{R}(H)$  if and only if  $1/m \in L_2(0, 1)$ , and hence we shall confine ourselves to distance functions  $d_1(R)$ ,  $R > 0$ , with respect to the benchmark  $\psi(t) = t$ ,  $t > 0$ .

We will derive results for the quantities under consideration for different multipliers  $m$  for which  $1 \notin \mathcal{R}(H)$ . In our considerations we shall need the family  $w_\alpha := \chi_{(\alpha, 1]}(H)H^{-1}1$ , for which  $1 - Hw_\alpha = \chi_{(0, \alpha]}(H)1$ , thus  $\|1 - Hw_\alpha\| = F_x(\alpha)$ . Moreover, the representation (18) yields that

$$\|w_\alpha\|^2 = \int_{m^{-1}(\alpha)}^1 \frac{1}{m^2(s)} ds. \quad (19)$$

The following cases for multipliers are of interest.

(a)  $\mathbf{m}(s) = \sqrt{s}$ : This is the limiting case for  $1 \notin \mathcal{R}(H)$ . Here we have  $F_x(\alpha) = \alpha$ ,  $0 < \alpha \leq 1$  and hence  $F_x(\alpha) \asymp \alpha$ , i.e. item (I) in Theorem 2 is satisfied with  $\kappa = 1$ , but we cannot apply this theorem with  $v = 1$  and benchmark  $\psi(t) = t$ ,  $t > 0$ , since the required strict inequality  $\kappa < v$  fails. This is reflected in the following behavior of the distance function. Plainly,  $\|w_\alpha\| = \sqrt{\log(1/\alpha^2)}$  and thus  $d_1\left(\sqrt{\log(1/\alpha^2)}\right) \leq \alpha$ ,  $0 < \alpha < 1$ . By letting  $R := \sqrt{\log(1/\alpha^2)}$  this rewrites as

$$d_1(R) \leq e^{-R^2/2}, \quad R > 0,$$

which expresses a very high decay rate of the distance function as  $R \rightarrow \infty$ .

We obtain some lower bound by using the Fenchel duality from (8) with Ansatz

$$v := u_\alpha(s) = \sqrt{\alpha} s^{-1} \chi_{[\alpha, 1]}(s), \quad 0 < s < 1,$$

for  $\alpha$  from  $\sqrt{\log(1/\alpha)} = \theta R$  and  $\theta > 1$ . This yields  $\|u_\alpha\| \leq 1$ ,  $\|Hu_\alpha\| \leq \sqrt{\log(1/\alpha)}$ , and  $\langle 1, u_\alpha \rangle = \sqrt{\alpha} \log(1/\alpha)$ , such that

$$d_1(R) \geq (\theta - 1)R^2 e^{-\theta \frac{R^2}{2}}, \quad R \geq R_\theta > 0.$$

This yields that for each  $\theta > 1$ , and for some unspecified constants  $0 < \underline{c}(\theta) \leq \bar{c}(\theta) < \infty$  we have that

$$\underline{c} \left( \sqrt{\alpha^2 \log\left(\frac{1}{\alpha^2}\right)} \right)^\theta \leq \|r_\alpha(H)1\| \leq \bar{c} \sqrt{\alpha^2 \log\left(\frac{1}{\alpha^2}\right)}, \quad 0 < \alpha \leq \bar{\alpha}(\theta). \quad (20)$$

Indeed, for the upper estimate we temporarily introduce the function  $\zeta(\alpha) := \sqrt{\log(1/(\alpha^2 \log(1/\alpha^2)))}$ . For this function we have, using the lower bound from above, that

$$\alpha \zeta(\alpha) \leq \alpha^2 \leq e^{-\theta \zeta^2(\alpha)} \leq d_1(\zeta(\alpha)),$$

for  $\alpha$  small enough. This yields  $\Phi_1(\zeta(\alpha)) \geq \alpha$ , and the monotonicity of  $\Phi_1$  implies that  $\Phi_1^{-1}(\alpha) \geq \zeta(\alpha)$ . Then we can use the upper bound in Theorem 1 to deduce that

$$\|r_\alpha(H)1\| \leq d_1(\Phi_1^{-1}(\alpha)) \leq d_1(\zeta(\alpha)) \leq e^{-\zeta^2(\alpha)/2} \leq \sqrt{\alpha^2 \log\left(\frac{1}{\alpha^2}\right)},$$

which proves the right hande side in (20). The lower bound is established similarly, and we omit its proof. We observe from (20) that Theorem 1 does not give precise bounds for the profile function. For Tikhonov regularization we can derive the behavior of the profile function, and it behaves like  $f_x(\alpha) \asymp \sqrt{\alpha^2 \log\left(\frac{1}{\alpha^2}\right)}$  as  $\alpha \rightarrow 0$ . Indeed, in view of the upper bound in (20), we only need a lower bound, and this can be derived as

$$f_x^2(\alpha) = \|r_\alpha(H)1\|^2 = \int_0^1 \frac{\alpha^2}{(\sqrt{s} + \alpha)^2} ds \geq c \alpha^2 \log(1/\alpha),$$

for  $\alpha$  small enough and some constant  $c > 0$ , such that the upper bound in (20) has the right order of magnitude. Notice also the different rates of increase for the distribution function ( $F_x(\alpha) \asymp \alpha$ ) and the profile function.

$m(s)$	$F_x(\alpha)$	$f_x(\alpha)$	$d_1(R)$
$\sqrt{s}$	$\alpha$	$\left(\sqrt{\alpha^2 \log\left(\frac{1}{\alpha^2}\right)}\right)^\theta$	$e^{-\theta R^2/2}$
$s$	$\sqrt{\alpha}$	$\sqrt{\alpha}$	$1/R$
$e^{1-1/\sqrt{s}}$	$1/\log(1/\alpha)$	$1/\log(1/\alpha)$	$1/\log(R)$

Table 1: This table summarizes the asymptotic behavior of the quantities of interest in cases (a)–(c). Notice that the exponent  $\theta$  in the row for  $m(s) = \sqrt{s}$  is not sharp, and that it equals  $\theta = 1$  for the upper bound, while any  $\theta > 1$  can be chosen for the lower bound.

**(b)  $m(s) = s$ :** *This is an intermediate case, and we shall see that the main results are sharp, here. In this case  $F_x(\alpha) = \sqrt{\alpha}$ ,  $0 < \alpha \leq 1$ , and by applying Theorem 2 with  $\nu = 1$  and  $\kappa = 1/2$  we immediately find both, the corresponding rates for the distance function  $d_1$  as*

$$\underline{c} \frac{1}{R} \leq d_1(R) \leq \bar{c} \frac{1}{R}, \quad 0 < R_0 \leq R < \infty,$$

as well as for the profile functions  $f_x$  as

$$\tilde{c} \sqrt{\alpha} \leq \|r_\alpha(H)1\| \leq \hat{c} \sqrt{\alpha}, \quad 0 < \alpha \leq \bar{\alpha}.$$

**(c)  $m(s) = e^{1-1/\sqrt{s}}$ :** *This is an example of the other limiting case, where Theorem 2 also cannot be applied. Such multiplier function  $m$  results in a logarithmic increase for the distribution function, since  $m^{-1}(\alpha) = 1/(1 + \log(1/\alpha))^2$ , and hence  $F_x(\alpha) \asymp 1/\log(1/\alpha)$  as  $\alpha \rightarrow 0$ . To bound the distance function from above we use (19) to see that  $\|w_\alpha\| \leq 1/\alpha$ . Thus,  $d_1(R) \leq 1/\log(R)$ ,  $R > 0$ . The right hand side in Corollary 1 yields a lower bound for the distance function. Indeed, this gives  $\frac{c}{\log(R/d_1(R))} \leq d_1(R)$ , and thus, since  $\log(\xi) \leq c\xi/2$ ,  $\xi \geq 2/c$ ,*

$$\frac{c}{2} \frac{1}{d_1(R)} \leq c \frac{1}{d_1(R)} - \log\left(\frac{1}{d_1(R)}\right) \leq \log(R), \quad R \geq R_0,$$

such that  $d_1(R) \asymp 1/\log(R)$ , which expresses a very low decay rate of the distance function as  $R \rightarrow \infty$ . On the other hand, the function  $1/d_1$  obeys a  $\Delta_2$ -condition in this case, and the bounds from Theorem 1, and Corollary 1, provide the right order of magnitude, such that also in this case we have the exact asymptotics for all quantities of interest.

We summarize the derived asymptotic results in Table 1.

**Example 4.** We use once more the multiplication operator (16), now with  $m(s) = \sqrt{s}$ , but we consider the function

$$x(s) := \sqrt{s^{-1/2} \log(1/s)}, \quad 0 < s < 1.$$

By using (17), and partial integration, we see that

$$F_x^2(t) = \int_0^{t^2} s^{-1/2} \log(1/s) ds = 2t \log(1/t^2) + 2 \int_0^{t^2} s^{-1/2} ds \asymp t \log(1/t).$$

as  $t \rightarrow 0$ . If we now fix the benchmark  $\psi(t) = t$ , then we see that then  $d_1(R) \asymp R^{-1} \log(R)$ . This is a non-polynomial, but still moderate, behavior. Let us consider the quantity

$$\eta_{sup} := \sup \left\{ \eta > 0 : d_1(R) = \mathcal{O} \left( R^{-\eta/(1-\eta)} \right) \right\} \in (0, 1], \quad (21)$$

and we see that for  $x$  from above it holds  $\eta_{sup} = 1/2$ , but the supremum is not attained here. On the other hand, it still holds that  $x \in \mathcal{R}(H^\mu)$  for all  $\mu < 1/2$ . Revisiting [2, Cor. 3.3] we must note that the one-to-one correspondence of exponents in distance function and Hölder type source conditions mentioned there in Corollary 3.3, *ibid.*, has to be rendered more precisely as  $\eta_{sup} = \sup \{ \mu > 0 : x \in \mathcal{R}(H^\mu) \}$ .

**Example 5.** The authors in [10] studied the compact integration operator  $A$  in  $L_2[0, 1]$ , given as

$$[Ax](t) := \int_0^t x(\tau) d\tau, \quad 0 \leq t \leq 1.$$

These authors computed the asymptotics of the distance function for the constant function  $x \equiv 1$  with benchmark  $A^*$ . Since  $\mathcal{R}(A^*) = \mathcal{R}(H^{1/2})$  this corresponds to  $\psi_{1/2}(t) = \sqrt{t}$ , and it was proved there that  $d_{1/2}(R) \asymp 1/R$  as  $R \rightarrow \infty$ . Within the present context this can be seen by evaluating the asymptotics of  $F_x$ , since we know the singular system  $\{\sigma_i, u_i, v_i\}_{i=1}^\infty$  of the integration operator  $A$  with singular values  $\sigma_i = \frac{1}{\pi(i-1/2)} \asymp i^{-1}$  as  $i \rightarrow \infty$  and singular functions  $u_i(t) = \sqrt{2} \cos((i-1/2)\pi t)$ ,  $0 \leq t \leq 1$ ,  $i = 1, 2, \dots$ . Then  $\langle 1, u_i \rangle \asymp i^{-1}$  and consequently

$$\sum_{i=k}^\infty \langle 1, u_i \rangle^2 \asymp \sum_{i=k}^\infty \frac{1}{i^2} \asymp \frac{1}{k} \asymp \sigma_k \quad \text{as } k \rightarrow \infty.$$

We know from [19, Remark 2.2] that for all  $\mu > 0$  the asymptotics  $\sum_{i=k}^\infty \langle x, u_i \rangle^2 \asymp \sigma_k^{4\mu}$  is equivalent to  $F_x(\alpha) \asymp \alpha^\mu$ . Here we apply this for  $\mu = 1/4$  and  $x \equiv 1$ . Then we have  $F_x(\alpha) \asymp \alpha^{1/4}$ , and we can use Theorem 2 with  $\kappa = 1/4$  and  $\nu = 1/2$  to see that  $f_x(\alpha) \asymp \alpha^{1/4}$ , and also that  $d_{1/2}(R) \asymp 1/R$ , which concisely reproves the result from [10].

## A Proof of Theorem 1

We first provide the necessary ingredients to prove Theorem 1. In our followup analysis we will approximate the unknown minimizing element  $v$  in the definition of  $d_\psi(R)$  within some suitable family  $v_\alpha$  and  $w_\alpha$ , respectively, and we introduce these, here. We (formally) assign

$$v_\alpha := g_\alpha(\psi^2(H))\psi(H)x, \quad (22)$$

and

$$w_\alpha := g_\alpha(H)H\psi(H)^{-1}x, \quad 0 < \alpha \leq \bar{\alpha}, \quad (23)$$

where in case of  $w_\alpha$  we have to assume that the operator  $g_\alpha(H)H\psi(H)^{-1}$  is a bounded one. For spectral cut-off the element  $w_\alpha$  is finite without any constraint on the function  $\psi$ , since there  $w_\alpha = \int_\alpha^\infty \psi(t)^{-1} dE_t x$  for  $\alpha > 0$ . Otherwise, for general regularization we provide the following sufficient conditions.

**Lemma 4.** *If the function  $t \mapsto \psi^2(t)/t$  is non-increasing then*

$$\|g_\alpha(H)H\psi(H)^{-1}\| \leq \frac{\max\{\gamma_0, \gamma_*\}}{\psi(\alpha)}, \quad 0 < \alpha \leq \bar{\alpha}.$$

*Thus the element  $w_\alpha$  is well-defined in this case.*

*Proof.* It is enough to show that  $t|g_\alpha(t)|/\psi(t) \leq \max\{\gamma_0, \gamma_*\}/\psi(\alpha)$ . If  $t > \alpha$  then the monotonicity of  $\psi$  allows to conclude that

$$t|g_\alpha(t)|/\psi(t) \leq \gamma_0/\psi(\alpha).$$

Otherwise, if  $t \leq \alpha$  then

$$t|g_\alpha(t)|/\psi(t) = \sqrt{t}|g_\alpha(t)| \frac{\sqrt{t}}{\psi(t)} \leq \frac{\gamma_*}{\sqrt{\alpha}} \frac{\sqrt{\alpha}}{\psi(\alpha)},$$

and the proof is complete. □

**Remark 3.** *A look at Example 3, see (19), reveals that we used exactly the construction  $w_\alpha$  corresponding to (23) and for spectral cut-off.*

**Remark 4.** *Notice that for  $\psi(t) = \sqrt{t}$ ,  $t > 0$ , both elements  $v_\alpha$  and  $w_\alpha$  coincide. The proofs given below will distinguish between the two cases that  $\psi$  tends to zero slower than  $t \mapsto \sqrt{t}$  (low-benchmark case) and faster than  $t \mapsto \sqrt{t}$  (high-benchmark case).*

The following is obvious.

$$\|x - \psi(H)v_\alpha\| = \|r_\alpha(\psi^2(H))x\| \quad (24)$$

and

$$\|x - \psi(H)w_\alpha\| = \|r_\alpha(H)x\|. \quad (25)$$

## A.1 Low-benchmark case

Here we assume that the function  $\psi^2(t)/t$  is non-increasing, and hence the element  $w_\alpha$  from (23) is well-defined.

**Lemma 5.** *If the function  $\psi^2(t)/t$  is non-increasing then*

$$d_\psi\left(2 \max\{\gamma_0, \gamma_*\} \Phi_\psi^{-1}(\psi(\alpha))\right) \leq \|r_\alpha(H)x\|, \quad 0 < \alpha \leq \bar{\alpha}. \quad (26)$$

*Proof.* Given  $\alpha$ , we let  $R$  from  $\psi(\alpha) = \Phi_\psi(R)$ , and denote by  $v$  the minimizing element for  $d_\psi(R)$ . By using the element  $w_\alpha$  from (23) we have by (25) that

$$d_\psi(\|w_\alpha\|) \leq \|r_\alpha(H)x\|,$$

such that it is enough to bound  $\|w_\alpha\|$ , appropriately.

$$\begin{aligned} \|w_\alpha\| &\leq \|g_\alpha(H)H\psi^{-1}(H)(x - \psi(H)v)\| + \|g_\alpha(H)H\psi^{-1}(H)\psi(H)v\| \\ &\leq \max\{\gamma_0, \gamma_*\} \frac{d_\psi(R)}{\psi(\alpha)} + \gamma_0 R \\ &\leq 2 \max\{\gamma_0, \gamma_*\} R \max\left\{\frac{\Phi_\psi(R)}{\psi(\alpha)}, 1\right\} = 2 \max\{\gamma_0, \gamma_*\} R. \end{aligned}$$

From this the proof can easily be completed.  $\square$

## A.2 High-benchmark case

We turn to the the high-benchmark case, where  $\psi$  tends to zero faster than  $t \mapsto \sqrt{t}$ , and we start with the following observation.

**Lemma 6.** *Let  $g_\alpha$  be any regularization. Then*

$$d_\psi((\gamma_0 + \gamma_*)\Phi_\psi^{-1}(\sqrt{\alpha})) \leq \|r_\alpha(\psi^2(H))x\|, \quad 0 < \alpha \leq \bar{\alpha}. \quad (27)$$

*If, in addition the function  $t \mapsto \sqrt{t}$  is a qualification of the regularization then*

$$\|r_\alpha(\psi^2(H))x\| \leq (\gamma_0 + \gamma_*)d_\psi\left(\Phi_\psi^{-1}(\sqrt{\alpha})\right), \quad 0 < \alpha \leq \bar{\alpha}. \quad (28)$$

*Proof.* We consider the family  $v_\alpha$  from (22). In a first step we bound the norm of  $v_\alpha$ . To this end let  $R$  be obtained from  $\sqrt{\alpha} = \Phi_\psi(R)$ , and denote  $v$  the element realizing the distance function  $d_\psi(R)$ . Then we have that

$$\begin{aligned} \|v_\alpha\| &\leq \|g_\alpha(\psi^2(H))\psi(H)(x - \psi(H)v)\| + \|g_\alpha(\psi^2(H))\psi^2(H)v\| \\ &\leq \frac{\gamma_*}{\sqrt{\alpha}}d_\psi(R) + \gamma_0 R = (\gamma_* + \gamma_0)R = (\gamma_* + \gamma_0)\Phi_\psi^{-1}(\sqrt{\alpha}), \end{aligned}$$

by the choice of  $R$ .

Since the distance function is decreasing we obtain that

$$d_\psi((\gamma_0 + \gamma_*)\Phi_\psi^{-1}(\sqrt{\alpha})) \leq d_\psi(\|v_\alpha\|) \leq \|x - \psi(H)v_\alpha\| = \|r_\alpha(\psi^2(H))x\|,$$

which yields (27).

Now, suppose that  $g_\alpha$  has qualification as stated. Then we can argue, with element  $v$  as before, that

$$\begin{aligned} \|r_\alpha(\psi^2(H))x\| &\leq \|r_\alpha(\psi^2(H))(x - \psi(H)v)\| + \|r_\alpha(\psi^2(H))\psi(H)v\| \\ &\leq \gamma_1 d_\psi(R) + \gamma\sqrt{\alpha}R = (\gamma_1 + \gamma)d_\psi(R) \\ &= (\gamma_1 + \gamma)d_\psi(\Phi_\psi^{-1}(\sqrt{\alpha})). \end{aligned}$$

The proof is complete.  $\square$

In order to establish the required bound from Theorem 1 we need to bound  $\|r_\alpha(\psi^2(H))x\|$  from above. To this end we provide some estimate for regularization from a single function.

**Lemma 7.** *Suppose that  $\psi$  is an index function such that  $\psi^2(t)/t$  is non-decreasing. If the function  $g$  gives rise for a linear regularization and if the accompanying function  $r$  has a non-increasing absolute value, and if  $|r(1)| > 0$ , then*

$$\left| r\left(\frac{\psi^2(t)}{\psi^2(\alpha)}\right) \right| \leq \frac{\gamma_1}{|r(1)|} r\left(\frac{t}{\alpha}\right), \quad t, \alpha > 0. \quad (29)$$

Consequently,

$$\left| r_{\psi^2(\alpha)}(\psi^2(t)) \right| \leq \frac{\gamma_1}{|r(1)|} |r_\alpha(t)|.$$

*Proof.* We consider two cases. If  $t \leq \alpha$  then  $0 < \psi^2(t)/\psi^2(\alpha) \leq 1$ , and hence

$$\left| r\left(\frac{\psi^2(t)}{\psi^2(\alpha)}\right) \right| \leq |r(0)| = \gamma_1 \leq \frac{\gamma_1}{|r(1)|} |r(1)| \leq \frac{\gamma_1}{|r(1)|} \left| r\left(\frac{t}{\alpha}\right) \right|.$$

Otherwise, if  $t > \alpha$  then  $\frac{\psi^2(t)}{t} > \frac{\psi^2(\alpha)}{\alpha}$ , and hence  $\frac{\psi^2(t)}{\psi^2(\alpha)} > \frac{t}{\alpha}$ , such that the monotonicity of  $x \mapsto |r(x)|$  yields

$$\left| r\left(\frac{\psi^2(t)}{\psi^2(\alpha)}\right) \right| \leq \left| r\left(\frac{t}{\alpha}\right) \right|,$$

and the proof is complete, since  $1 \leq \gamma_1/|r(1)|$ .  $\square$

For Tikhonov regularization this is fulfilled with  $\gamma_1/|r(1)| = 2$ .

**Remark 5.** *Notice that in case of spectral cut-off we have the equality  $\left| r_{\psi^2(\alpha)}(\psi^2(t)) \right| = |r_\alpha(t)|$  for arbitrary index function  $\psi$ .*

The bound for the high-benchmark case is now given in

**Lemma 8.** *Let  $g_\alpha$  be a regularization which is obtained from a single function  $g$  with non-increasing function  $r$ ,  $|r(1)| > 0$ . Then*

$$d_\psi((\gamma_0 + \gamma_*)\Phi_\psi^{-1}(\psi(\alpha))) \leq \frac{\gamma_1}{|r(1)|} \|r_\alpha(H)x\|, \quad \alpha > 0.$$

*Proof.* We shall apply Lemma 6 with  $\alpha := \psi^2(\alpha)$  and obtain that

$$d_\psi((\gamma_0 + \gamma_*)\Phi_\psi^{-1}(\sqrt{\alpha})) \leq \|r_{\psi^2(\alpha)}(\psi^2(H))x\|, \quad \alpha > 0,$$

such that an application of Lemma 7 allows to complete the proof.  $\square$

*Proof of Theorem 1.* Lemmas 5 and 8 yield the first assertion in Theorem 1. The upper bound was established in Proposition 2, by noticing that  $\gamma_1 + \gamma \leq 2 \max\{\gamma_1, \gamma\}$ .  $\square$

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