On the existence of global saturation for spectral regularization methods with optimal qualification

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Abstract

A family of real functions \( \{g_\alpha\} \) defining a spectral regularization method with optimal qualification is considered. Sufficient condition on the family and on the optimal qualification guaranteeing the existence of saturation are established. Appropriate characterizations of both the saturation function and the saturation set are found and some examples are provided.

Keywords: Ill-posed, inverse problem, qualification, saturation.

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1 Introduction

Let \( X, Y \) be infinite dimensional Hilbert spaces and \( T : X \to Y \) a bounded linear operator with non-closed range \( \mathcal{R}(T) \). It is well known that under these conditions \( T^\dagger \), the Moore-Penrose generalized inverse of \( T \), is unbounded ([1]) and therefore the linear operator equation

\[
Tx = y
\]

is ill-posed. The Moore-Penrose generalized inverse can be used to define the least squares solutions of ([1]). In fact equation ([1]) has a least squares solution if and only if \( y \in \mathcal{D}(T^\dagger) = \mathcal{R}(T) \oplus \mathcal{R}(T)^\perp \) and in that case, \( x^\dagger = T^\dagger y \) is the least squares solution of minimum norm and the set of all least-squares solutions of ([1]) is given by \( x^\dagger + \mathcal{N}(T) \). Since \( T^\dagger \) is unbounded, \( x^\dagger \) does not depend continuously on the data \( y \). Therefore, if instead of the exact data \( y \), a noisy observation \( y^\delta \) is available, \( y^\delta = Tx + \delta \xi \), where the noise \( \xi \) is assumed to be bounded, \( \|\xi\| \leq 1 \), then it is possible that \( T^\dagger y^\delta \) does not even exist and if it does, it will not necessarily be a good approximation of \( x^\dagger \)

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EXISTENCE OF SATURATION FOR SPECTRAL REGULARIZATION METHODS

This instability becomes evident when trying to approximate \( x^\dagger \) by traditional numerical methods and procedures.

Ill-posed problems must be first regularized if one wants to successfully attack the task of numerically approximating their solutions. Regularizing an ill-posed problem such as (1) essentially means approximating the operator \( T^\dagger \) by a parametric family of bounded operators \( \{ R_\alpha \} \), where \( \alpha \) is the so called “regularization parameter”. If \( y \in D(T^\dagger) \), then the best approximate solution \( x^\dagger \) of (1) can be written as

\[
x^\dagger = \int_0^{\|T\|^2} \frac{1}{\lambda} dE_\lambda T^* y \quad \text{where} \quad \{ E_\lambda \} \text{ is the spectral family associated to the operator } T^* T \text{ (see [1])}.
\]

This is mainly why many regularization methods are based on spectral theory and consist of defining \( R_\alpha \). The importance of choosing the parameter \( \alpha \) appropriately such that for every \( \lambda \in (0, \|T\|^2] \) there holds

\[
\lim_{\alpha \to 0^+} g_\alpha(\lambda) = \frac{1}{\lambda}.
\]

It is important to emphasize however that no mathematical trick can make stable a problem that is intrinsically unstable. Whatever the case, there is always loss of information. All a regularization method can do is to recover the largest possible amount of information about the solution of the problem, maintaining stability. It is often said that the art of applying regularization methods consist always in maintaining an adequate balance between accuracy and stability. Usually accuracy can be improved with increasing assumptions (or information) on the regularity of the exact solution.

In 1994, however, Neubauer ([7]) showed that certain spectral regularization methods “saturate”, that is, they become unable to continue extracting additional information about the exact solution even upon increasing regularity assumptions on it. In his article, Neubauer introduced for the first time the idea of the concept of “saturation” of regularization methods. Saturation is however a rather subtle and complex issue in the study of regularization methods for inverse ill-posed problems and the concept has, for many years, escaped rigorous formalization in a general context. Neubauer’s idea referred to the best order of convergence that a method can achieve independently of the smoothness assumptions on the exact solution and on the selection of the parameter choice rule. In 1997, Neubauer ([8]) showed that this saturation phenomenon occurs for instance in the classical Tikhonov-Phillips method. Later on, in 2004, Mathé ([5]) proposed a general definition of the concept of saturation for spectral regularization methods. However, the concept of saturation defined by Mathé is not applicable to general regularization methods and it is not fully compatible with the original idea of saturation proposed by Neubauer in [7]. In particular, for instance, the definition of saturation given in [5] does not imply uniqueness and therefore, neither a best global order of convergence. More recently, in 2011, Herdman, Spies and Temperini (see [3]) developed a general theory of global saturation for arbitrary regularization methods, formalizing the original and intuitive idea first outlined by Neubauer in 1994 ([7]).

Related in a dual way to the concept of saturation is the concept of qualification of a spectral regularization method, introduced by Mathé and Pereverzev in 2003 ([6]). This concept is strongly related to the optimal order of convergence of the regularization error, under certain “a-priori” assumptions on the exact solution. In 2009 Herdman, Spies and Temperini ([2]) generalized the concept of qualification and introduced three hierarchical levels of it: weak, strong and optimal qualification. There, it was shown that the weak qualification generalizes the definition introduced in [6].

In this work, some light on the existence of saturation for spectral regularization methods with optimal qualification is shed. In particular, sufficient conditions on the family of real functions \( \{ g_\alpha \} \) defining the method and on the optimal qualification \( \rho \), which guarantee the existence of saturation, are established. Moreover, in those cases, appropriate characterizations of both the
saturation function and the saturation set are provided.

2 Preliminaries

In this section we shall recall some basic concepts on global saturation of regularization methods for inverse ill-posed problems theory (for more details see [3]). In the sequel, \( T : X \to Y \) will be a bounded linear operator with non-closed range between two Hilbert spaces \( X \) and \( Y \). Without loss of generality we will assume that the operator \( T \) is invertible (in the context of inverse problems it is customary to work with the Moore-Penrose generalized inverse of \( T \) since one seeks least squares solutions of the problems; therefore the lack of injectivity of \( T \) is never a relevant issue). Also, for simplicity of notation and unless otherwise specified, we shall assume that all subsets of the Hilbert space \( X \) under consideration are not empty and they do not contain \( x = 0 \).

Let \( M \subset X \). We shall say that a function \( \psi : X \times \mathbb{R} \to \mathbb{R} \) belongs to the class \( \mathcal{F}_M \) if there exists \( a = a(\psi) > 0 \) such that \( \psi \) is defined in \( M \times (0, a) \), with values in \( (0, \infty) \) and it satisfies the following conditions:

1. \( \lim_{{\delta \to 0^+}} \psi(x, \delta) = 0 \) for all \( x \in M \), and
2. \( \psi \) is continuous and non-decreasing as a function of \( \delta \) in \( (0, a) \) for each fixed \( x \in M \).

One may think of \( \mathcal{F}_M \) as the collection of all possible \( \delta \)-“orders of convergence” on the set \( M \).

**Definition 2.1.** Let \( \{ R_\alpha \}_{\alpha \in (0, a_0)} \) be a family of regularization operators for the problem \( Tx = y \). The “total error of \( \{ R_\alpha \}_{\alpha \in (0, a_0)} \) at \( x \in X \) for a noise level \( \delta \)” is defined as

\[
\mathcal{E}_\{R_\alpha\}^{\text{tot}}(x, \delta) = \inf_{{\alpha \in (0, a_0)}} \sup_{{y^* \in B_\delta(Tx)}} \| R_\alpha y^* - x \|,
\]

where \( B_\delta(Tx) = \{ y \in Y : \| Tx - y \| \leq \delta \} \).

Note that \( \mathcal{E}_\{R_\alpha\}^{\text{tot}} \) is the error in the sense of the largest possible discrepancy that can be obtained for an observation of \( y \) within noise level \( \delta \), with an appropriate choice of the regularization parameter \( \alpha \).

**Definition 2.2.** Let \( M \subset X \) and \( \psi, \tilde{\psi} \in \mathcal{F}_M \). We say that “\( \psi \) precedes \( \tilde{\psi} \) on \( M \)” and we denote it with \( \psi \preceq \tilde{\psi} \), if there exist a constant \( r > 0 \) and a function \( p : M \to (0, \infty) \) such that \( \psi(x, \delta) \leq p(x) \tilde{\psi}(x, \delta) \) for all \( x \in M \) and for every \( \delta \in (0, r) \).

**Definition 2.3.** Let \( \{ R_\alpha \}_{\alpha \in (0, a_0)} \) be a family of regularization operators for the problem \( Tx = y \), \( M \subset X \) and \( \psi \in \mathcal{F}_M \). We say that \( \psi \) is an “upper bound of convergence for the total error of \( \{ R_\alpha \}_{\alpha \in (0, a_0)} \) on \( M \)” if \( \mathcal{E}_\{R_\alpha\}^{\text{tot}} \preceq \psi \).

With \( \mathcal{U}_M(\mathcal{E}_\{R_\alpha\}^{\text{tot}}) \) we shall denote the set of all functions \( \psi \in \mathcal{F}_M \) that are upper bounds of convergence for the total error of \( \{ R_\alpha \}_{\alpha \in (0, a_0)} \) on \( M \).

The following two definitions formalize certain comparisons of bounds of convergence on different sets of \( X \), which will be needed later to introduce the concept of global saturation.
Definition 2.4. Let $M, \tilde{M} \subset X$, $\psi \in F_M$ and $\tilde{\psi} \in F_{\tilde{M}}$.

i) We say that “$\psi$ on $M$ precedes $\tilde{\psi}$ on $\tilde{M}$”, and we denote it with $\psi \preceq M,\tilde{M} \tilde{\psi}$, if there exist a constant $d > 0$ and a function $k : M \times \tilde{M} \to (0, \infty)$ such that $\psi(x, \delta) \leq k(x, \tilde{x}) \tilde{\psi}(\tilde{x}, \delta)$ for every $x \in M$, $\tilde{x} \in \tilde{M}$ and $\delta \in (0, d)$.

ii) We say that “$\psi$ on $M$ is equivalent to $\tilde{\psi}$ on $\tilde{M}$”, and we denote it with $M,\tilde{M} \approx \psi$, if $\psi \preceq M,\tilde{M} \tilde{\psi}$ and $\tilde{\psi} \preceq M,\tilde{M} \psi$.

Definition 2.5. Let $M \subset X$ and $\psi \in F_M$. We say that “$\psi$ is invariant over $M$” if $\psi \equiv M,\tilde{M} \psi$.

Next we recall the concept of global saturation introduced in [3].

Definition 2.6. Let $M_S \subset X$ and $\psi_S \in U_{M_S}(C^\text{tot}_{\{R_\alpha\}})$. It is said that $\psi_S$ is a “global saturation function of $\{R_\alpha\}$ over $M_S$” if $\psi_S$ satisfies the following three conditions:

$S1$. For every $x^* \in X$, $x^* \neq 0$, $x \in M_S$, $\limsup_{\delta \to 0^+} \frac{C^\text{tot}_{\{R_\alpha\}}(x^*, \delta)}{\psi_S(x, \delta)} > 0$.

$S2$. $\psi_S$ is invariant over $M_S$.

$S3$. There is no upper bound of convergence for the total error of $\{R_\alpha\}$ that is a proper extension of $\psi_S$ (in the variable $x$) and satisfies $S1$ and $S2$, that is, there exist no $\tilde{M} \supset M_S$ and $\tilde{\psi} \in U_{\tilde{M}}(C^\text{tot}_{\{R_\alpha\}})$ such that $\tilde{\psi}$ satisfies $S1$ and $S2$ with $M_S$ replaced by $\tilde{M}$ and $\psi_S$ replaced by $\tilde{\psi}$.

The function $\psi_S$ and the set $M_S$ are refer to as the saturation function and the saturation set, respectively.

This conception of global saturation essentially establishes that in no point $x^* \in X$, $x^* \neq 0$, can exist an upper bound of convergence for the total error of the regularization method that is “strictly better” than the saturation function $\psi_S$ at any point of the saturation set $M_S$.

Let $\{E_\lambda\}_{\lambda \in \mathbb{R}}$ be the spectral family associated to the linear selfadjoint operator $T^*T$ and $\{g_\alpha\}_{\alpha \in (0, a_0)}$ a parametric family of functions $g_\alpha : [0, \|T\|^2] \to \mathbb{R}$, $\alpha \in (0, a_0)$, and consider the following standing hypotheses:

$H1$. For every $\alpha \in (0, a_0)$ the function $g_\alpha$ is piecewise continuous on $[0, \|T\|^2]$.

$H2$. There exists a constant $C > 0$ (independent of $\alpha$) such that $|\lambda g_\alpha(\lambda)| \leq C$ for every $\lambda \in [0, \|T\|^2]$.

$H3$. For every $\lambda \in (0, \|T\|^2)$, there exists $\lim_{\alpha \to 0^+} g_\alpha(\lambda) = \frac{1}{\lambda}$.

If $\{g_\alpha\}_{\alpha \in (0, a_0)}$ satisfies hypotheses $H1$-$H3$, then (see [1], Theorem 4.1) the collection of operators $\{R_\alpha\}_{\alpha \in (0, a_0)}$, where

$$R_\alpha = \int_{0}^{\|T\|^2} g_\alpha(\lambda) dE_\lambda T^* = g_\alpha(T^*T)T^*,$$

is a family of regularization operators for $T^\dagger$. In this case we say that $\{R_\alpha\}_{\alpha \in (0, a_0)}$ is a “family of spectral regularization operators” (FSRO) for $Tx = y$ and $\{g_\alpha\}_{\alpha \in (0, a_0)}$ is a “spectral regularization method” (SRM).

The following definitions will be needed both to recall the concept of qualification as introduced in [2], as well as in the rest of the article.
We denote with $O$ the set of all non-decreasing functions $\rho : \mathbb{R}^+ \rightarrow \mathbb{R}^+$ such that $\lim_{\alpha \rightarrow 0^+} \rho(\alpha) = 0$ and with $S$ the set of all continuous functions $s : \mathbb{R}^+_0 \rightarrow \mathbb{R}^+_0$ satisfying $s(0) = 0$ and such that $s(\lambda) > 0$ for every $\lambda > 0$. Note that if $s \in S$ is non-decreasing, then $s$ is an index function in the sense of Mathé-Pereverzev [6].

**Definition 2.7.** Let $\rho, \tilde{\rho} \in O$. We say that “$\rho$ precedes $\tilde{\rho}$ at the origin” and we denote it with $\rho \preceq \tilde{\rho}$, if there exist positive constants $c$ and $\varepsilon$ such that $\rho(\alpha) \leq c\tilde{\rho}(\alpha)$ for every $\alpha \in (0, \varepsilon)$.

**Definition 2.8.** Let $\rho, \tilde{\rho} \in O$. We say that “$\rho$ and $\tilde{\rho}$ are equivalent at the origin” and we denote it with $\rho \approx \tilde{\rho}$, if there exist positive constants $\varepsilon > 0$, $c_1$, $c_2$, $0 < c_1 < c_2 < \infty$ such that $c_1 \rho(\alpha) \leq \tilde{\rho}(\alpha) \leq c_2 \rho(\alpha)$ for every $\alpha \in (0, \varepsilon)$.

Clearly “$\approx$” is an equivalence relation and it introduces in $O$ a partial ordering. Analogous definitions and notation will be used for $s, \tilde{s} \in S$.

**Definition 2.9.** Let $\{g_\alpha\}_{\alpha \in (0, a_0)}$ be a SRM, $r_\alpha(\lambda) \equiv 1 - \lambda g_\alpha(\lambda), \rho \in O$ and $s \in S$.

i) We say that $(s, \rho)$ is a “weak source-order pair for $\{g_\alpha\}$” if it satisfies

$$
\frac{s(\lambda) | r_\alpha(\lambda)|}{\rho(\alpha)} = O(1) \quad \text{for } \alpha \rightarrow 0^+, \forall \lambda > 0. \quad (2)
$$

ii) We say that $(s, \rho)$ is a “strong source-order pair for $\{g_\alpha\}$” if it is a weak source-order pair and there is no $\lambda > 0$ for which “$O(1)$” in (2) can be replaced by “$o(1)$”. That is, if $(s, \rho)$ is a weak source-order pair for $\{g_\alpha\}$ and

$$
\limsup_{\alpha \rightarrow 0^+} \frac{s(\lambda) | r_\alpha(\lambda)|}{\rho(\alpha)} > 0 \quad \forall \lambda > 0. \quad (3)
$$

iii) We say that $(\rho, s)$ is an “order-source pair for $\{g_\alpha\}$” if there exist a constant $\gamma > 0$ and a function $h : (0, a_0) \rightarrow \mathbb{R}^+$ with $\lim_{\alpha \rightarrow 0^+} h(\alpha) = 0$, such that

$$
\frac{s(\lambda) | r_\alpha(\lambda)|}{\rho(\alpha)} \geq \gamma \quad \forall \lambda \in [h(\alpha), +\infty). \quad (4)
$$

In the context of the previous definitions we refer to the function $\rho$ as the “order of convergence” and to $s$ as the “source function”.

We are now ready to define the concept of qualification in its three different levels as it was introduced in [2].

**Definition 2.10.** Let $\{g_\alpha\}$ be a SRM.

i) We say that $\rho$ is “weak qualification of $\{g_\alpha\}$” if there exists a function $s$ such that $(s, \rho)$ is a weak source-order pair for $\{g_\alpha\}$.

ii) We say that $\rho$ is “strong qualification of $\{g_\alpha\}$” if there exists a function $s$ such that $(s, \rho)$ is a strong source-order pair for $\{g_\alpha\}$.

iii) We say that $\rho$ is “optimal qualification of $\{g_\alpha\}$” if there exists a function $s$ such that $(s, \rho)$ is a strong source-order pair for $\{g_\alpha\}$ and $(\rho, s)$ is an order-source pair for $\{g_\alpha\}$. 
Note that since condition (1) implies condition (3), in the definition of optimal qualification above the requirement that \((s, \rho)\) be strong source-order pair can be replaced by the one that \((s, \rho)\) be a weak source-order pair.

Now given the SRM \(\{g_\alpha\}\) and \(\rho \in \mathcal{O}\), we define

\[
s_\rho(\lambda) = \liminf_{\alpha \to 0^+} \frac{\rho(\alpha)}{|r_\alpha(\lambda)|} \quad \text{for} \quad \lambda \geq 0.
\]

Note that \(s_\rho(0) = 0\) and if \(s_\rho\) is continuous, \(s_\rho \in \mathcal{S}\).

The next theorem provides necessary and sufficient condition, in terms of \(s_\rho\), for an order of convergence \(\rho \in \mathcal{O}\) to be optimal qualification.

**Theorem 2.11.** (2) Let \(\{g_\alpha\}\) be a SRM and \(\rho \in \mathcal{O}\) such that \(s_\rho \in \mathcal{S}\). Then \(\rho\) is optimal qualification of \(\{g_\alpha\}\) if and only if \(s_\rho\) verifies (1) and

\[
0 < s_\rho(\lambda) < +\infty \quad \text{for every} \quad \lambda > 0.
\]

The next theorem shows the uniqueness of the source function.

**Theorem 2.12.** (2) If \(\rho\) is optimal qualification of \(\{g_\alpha\}\) then there exists only one function \(s\) (in the sense of the equivalence classes induced by Definition 2.8) such that \((s, \rho)\) is a strong source-order pair and \((\rho, s)\) is an order-source pair for \(\{g_\alpha\}\). Moreover if \(s_\rho \in \mathcal{S}\), then \(s_\rho\) is such a unique function.

The following converse result, where regularity properties of the exact solution are derived in terms of the rate of convergence of the regularization error, will be needed later. This result states that if the regularization error has order of convergence \(\rho(\alpha)\) and \((\rho, s)\) is an order-source pair, then the exact solution belongs to the source set given by the range of the operator \(s(T^*T)\).

**Theorem 2.13.** (2) Let \(\{g_\alpha\}\) be a SRM and \(R_\alpha = g_\alpha(T^*T)T^*\). If \(\|(R_\alpha - T^*)y\| = O(\rho(\alpha))\) for \(\alpha \to 0^+\) and \((\rho, s)\) is an order-source pair for \(\{g_\alpha\}\), then \(T^*y \in \mathcal{R}(s(T^*T))\).

### 3 Saturation of spectral regularization methods with optimal qualification

The purpose of this section is to shed some light on the saturation of SRM with optimal qualification. More precisely, we will establish sufficient conditions on the family of functions \(\{g_\alpha\}\) and on the optimal qualification \(\rho\) guaranteeing the existence of saturation. Moreover, for those methods we will provide appropriate characterizations of both the saturation function and the saturation set.

Then, let \(\{g_\alpha\}_{\alpha \in (0, a_0)}\) be a SRM and consider the following hypothesis:

\[H_4\]  

\[\exists k > 0 \text{ such that } G_\alpha = \|g_\alpha(\cdot)\|_\infty \leq \frac{k}{\sqrt{\alpha}} \forall \alpha \in (0, a_0).\]

**Lemma 3.1.** Let \(\{g_\alpha\}_{\alpha \in (0, a_0)}\) be a SRM satisfying hypothesis \(H_4\), \(R_\alpha = g_\alpha(T^*T)T^* \) and \((s, \rho)\) a weak source-order pair for \(\{g_\alpha\}_{\alpha \in (0, a_0)}\) where \(\rho\) is continuous. Define \(X^* = \mathcal{R}(s(T^*T)) \setminus \{0\}\), \(\Theta(t) = \sqrt{t} \rho(t)\) for \(t > 0\), \(\psi(x, \delta) = \rho \circ \Theta^{-1}(\delta)\) for \(x \in X^*\) and \(\delta \in (0, \Theta(\alpha_0))\). Then \(\psi \in \mathcal{F}_{X^*}\) and, moreover, \(\psi\) is an upper bound of convergence for the total error of \(\{R_\alpha\}_{\alpha \in (0, a_0)}\) on \(X^*\), that is, \(\psi \in \mathcal{U}_{X^*}(s_{\{R_\alpha\}})\).
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**Proof.** Since \( \rho \) is continuous and non-decreasing and \( \rho(0^+) = 0 \) it follows that \( \Theta(t) \) is continuous and strictly increasing on \((0, +\infty)\) with \( \Theta(0^+) = 0 \). Therefore \( \Theta^{-1} \) exists and has the same properties. It then follows that \( \psi \) is continuous and non-decreasing as a function of \( \delta \) in \((0, \Theta(\alpha_0))\) for each fixed \( x \in X^s \), and \( \psi(x, 0^+) = 0 \) for all \( x \in X^s \). Hence \( \psi \in \mathcal{U}_{X^s} \).

On the other hand, since \((s, \rho)\) is a weak source-order pair for \( \{g_\alpha\}_{\alpha \in (0, \alpha_0)} \), there exist positive constants \( c \) and \( \hat{\alpha} \) such that \( s(\lambda) |r_\alpha(\lambda)| \leq c \rho(\alpha) \) for every \( \alpha \in (0, \hat{\alpha}) \), \( \lambda \in (0, \|T\|^2] \). Moreover, from hypothesis \( H2 \) and the fact that \( \rho \) is non-decreasing it follows that the previous inequality holds (perhaps with a different positive constant \( c \)) for every \( \alpha \in (0, \alpha_0) \), that is,

\[
\forall \alpha \in (0, \alpha_0), \forall \lambda \in (0, \|T\|^2], \quad s(\lambda) |r_\alpha(\lambda)| \leq c \rho(\alpha).
\]  

Now, for every \( p \geq 0 \) we define the source sets \( X^{s,p} \equiv \{ x \in X : x = s(T^*T)\zeta, \|\zeta\| \leq p \} \). Then for each \( x \in X^s \) there exists \( p_x > 1 \) such that \( x \in X^{s,p_x} \). On the other hand, since \( \Theta \) is continuous and strictly increasing in \((0, \alpha_0)\), there exists a unique \( \hat{\alpha}_x \in (0, \alpha_0) \) such that \( x \in X^{s,p_x} \) and \( \Theta(\hat{\alpha}_x) = \frac{\delta}{p_x} \). Therefore,

\[
\mathcal{E}_{\{R_\alpha\}}^{\text{tot}}(x, \delta) = \inf_{\alpha \in (0, \alpha_0)} \sup_{y^\delta \in B_\delta(Tx)} \left\| R_\alpha y^\delta - x \right\|
\leq \sup_{y^\delta \in B_\delta(Tx)} \left\| R_{\hat{\alpha}_x} y^\delta - x \right\|.
\]

Now, since \( y^\delta = Tx + \delta \xi, \|\xi\| \leq 1 \) and \( x = s(T^*T)\zeta \) with \( \|\zeta\| \leq p_x \), it follows immediately that

\[
\left\| R_{\hat{\alpha}_x} y^\delta - x \right\| \leq \| (g_{\hat{\alpha}_x}(T^*T)T^*T - I)s(T^*T)\zeta \| + \delta \| g_{\hat{\alpha}_x}(T^*T)T^*\xi \|
\leq p_x \sup_{\lambda \in (0,\|T\|^2]} \{ s(\lambda) |r_{\hat{\alpha}_x}(\lambda)| \} + \delta \sup_{\lambda \in (0,\|T\|^2]} \{ \sqrt{\lambda} |g_{\hat{\alpha}_x}(\lambda)| \},
\]

where the last inequality follows from properties of functions of a selfadjoint operator, (more precisely, for any piecewise continuous function \( f \) there holds \( \|f(T^*T)\| \leq \sup_\lambda |f(\lambda)| \) and \( \|f(T^*T)T^*\| \leq \sup_\lambda \{ \sqrt{\lambda} |f(\lambda)| \} \), see [11, p. 45]). Using (7) and hypothesis \( H4 \) in [3] it follows that

\[
\left\| R_{\hat{\alpha}_x} y^\delta - x \right\| \leq p_x c \rho(\hat{\alpha}_x) + \delta \frac{k}{\sqrt{\alpha_x}} \|T\|.
\]

Since \( \Theta(\hat{\alpha}_x) = \sqrt{\alpha_x} \rho(\hat{\alpha}_x) = \frac{\delta}{p_x} \), it follows that \( \frac{\delta}{\sqrt{\alpha_x}} = p_x \rho(\hat{\alpha}_x) \). Hence by virtue of (3) one has that

\[
\left\| R_{\hat{\alpha}_x} y^\delta - x \right\| \leq p_x (c + k \|T\|) \rho(\hat{\alpha}_x)
\leq p_x (c + k \|T\|) \rho\left( \Theta^{-1} \left( \frac{\delta}{p_x} \right) \right)
\leq p_x (c + k \|T\|) \rho(\Theta^{-1}(\delta)),
\]

where the last inequality follows from the fact that \( p_x \geq 1 \) and both \( \rho \) and \( \Theta^{-1} \) are non-decreasing functions. From (3) and (10) it follows that for every \( \delta \in (0, \Theta(\alpha_0)) \),

\[
\mathcal{E}_{\{R_\alpha\}}^{\text{tot}}(x, \delta) \leq p_x (c + k \|T\|) \rho(\Theta^{-1}(\delta)) = p_x (c + k \|T\|) \psi(x, \delta).
\]

This proves that \( \psi \in \mathcal{U}_{X^s}(\mathcal{E}_{\{R_\alpha\}}^{\text{tot}}) \). \( \square \)
Definition 3.2. ([3]) Let $k$ be a positive constant and $\rho : (0, k] \to (0, +\infty)$ a continuous non-decreasing function such that $\lim_{t \to 0^+} \rho(t) = 0$. We say that $\rho$ is of local upper type $\beta$ $(\beta \geq 0)$ if there exists a positive constant $d$ such that $\rho(t) \leq ds^{-\beta} \rho(st)$ for every $s \in (0, 1], t \in (0, k]$.

Theorem 3.3. (Saturation for FSRO with optimal qualification.) Let $\{g_\alpha\}_{\alpha \in (0, a_0)}$ be a SRM satisfying hypothesis $H_4$ and having optimal qualification $\rho, r_\alpha(\lambda) = 1 - \lambda g_\alpha(\lambda), R_\alpha = g_\alpha(T^*T)T^*$ and suppose that $s_\rho \in \mathcal{S}$ (where $s_\rho$ is as defined in [3]). Assume further that the following hypotheses hold:

a) The function $\rho$ is of local upper type $\beta$, for some $\beta \geq 0$.

b) There exist positive constants $\gamma_1, \gamma_2, \lambda^*, c_1$, with $\lambda^* \leq \|T\|^2$ and $c_1 > 1$ such that
   i) $0 \leq r_\alpha(\lambda) \leq 1$, for $\alpha > 0$, $0 \leq \lambda \leq \lambda^*$;
   ii) $r_\alpha(\lambda) \geq \gamma_1$, for $0 \leq \lambda < h(\alpha) \leq \lambda^*$, $\alpha \in (0, a_0)$ where $h$ is as in [3]. (Note that by virtue of Theorem 2.12 and the fact that $s_\rho \in \mathcal{S}$, there exists only one function $s \in \mathcal{S}$ satisfying [4], that is, $s = s_\rho$.)
   iii) $|r_\alpha(\lambda)|$ is non-decreasing with respect to $\alpha$ for each $\lambda \in (0, \|T\|^2]$;
   iv) $g_\alpha(c_1 \alpha) \geq \frac{2}{\alpha}$ for $0 < c_1 \alpha \leq \lambda^*$ and
   v) $g_\alpha(\lambda) \geq g_\alpha(\lambda), 0 < \alpha \leq \lambda \leq \lambda^*$.

c) There exist $\{\lambda_n\}_{n=1}^\infty \subset \sigma(TT^*)$ and $c \geq 1$ such that $\lambda_n \downarrow 0$ and $\frac{\lambda_n}{\lambda_{n+1}} \leq c$ for every $n \in \mathbb{N}$.

Let $\Theta(t) = \sqrt{t} \rho(t)$ for $t > 0$ and $X^{s_\rho} = \mathcal{R}(s_\rho(T^*T)) \setminus \{0\}$. Then $\psi(x, \delta) \eqdef \rho \circ \Theta^{-1}(\delta)$ for $x \in X^{s_\rho}$ and $\delta \in (0, \Theta(a_0))$, is saturation function of $\{R_\alpha\}_{\alpha \in (0, a_0)}$ on $X^{s_\rho}$.

To prove this theorem we will need three previous lemmas. The first one is a somewhat technical result, the second one deals with the existence of an a-priori parameter choice rule leading to a worst total error having an appropriate order of convergence, while the third one is a converse result.

Lemma 3.4. Let $\{g_\alpha\}_{\alpha \in (0, a_0)}$ be a SRM, $(\rho, s)$ an order-source pair for $\{g_\alpha\}$ and suppose hypothesis b.ii) of Theorem 3.3 holds. Then for every $\alpha \in (0, a_0)$ the operator $r_\alpha(T^*T)$ is invertible.

Proof. Let $\{E_\lambda\}$ be the spectral family of $T^*T$. It suffices to show that for every $\alpha \in (0, a_0), x \in X$, the function $r_\alpha^{-1}(\lambda)$ is integrable with respect to the measure $d\|E_\lambda x\|^2$. Let $\alpha \in (0, a_0)$ fixed. Since $(\rho, s)$ is an order-source pair for $\{g_\alpha\}$, there exist a constant $\gamma > 0$ and a function $h : (0, a_0) \to \mathbb{R}^+$ with $\lim_{\alpha \to 0^+} h(\alpha) = 0$ such that

$$\frac{s(\lambda) |r_\alpha(\lambda)|}{\rho(\alpha)} \geq \gamma \quad \forall \lambda \in [h(\alpha), +\infty).$$

Therefore

$$\int_{h(\alpha)}^{\|T\|^2+} \frac{1}{r_\alpha^2(\lambda)} d\|E_\lambda x\|^2 \leq \frac{1}{\gamma^2 \rho^2(\alpha)} \int_{h(\alpha)}^{\|T\|^2+} s^2(\lambda) d\|E_\lambda x\|^2 \leq \frac{\|s(T^*T)x\|^2}{\gamma^2 \rho^2(\alpha)} < +\infty. \quad (11)$$

Now, since $\alpha \in (0, a_0)$, it follows from hypothesis b.ii) of Theorem 3.3 that $r_\alpha(\lambda) \geq \gamma_1 > 0$ for every $\lambda \in [0, h(\alpha))$. Then

$$\int_{0}^{h(\alpha)} \frac{1}{r_\alpha^2(\lambda)} d\|E_\lambda x\|^2 \leq \frac{\|x\|^2}{\gamma_1^2} < +\infty. \quad (12)$$
Lemma 3.5. Let \( \{g_\alpha\} \) be a SRM, \( R_\alpha = g_\alpha(T^*T)T^* \), \((\rho, s)\) an order-source pair for \( \{g_\alpha\} \) and assume that hypotheses \( \text{b.ii)} \) and \( \text{b.iii)} \) of Theorem 3.3 hold. Let \( \varphi : (0, +\infty) \rightarrow \mathbb{R}^+ \) be a continuous, non-decreasing function satisfying \( \lim_{\delta \to 0^+} \varphi(\delta) = 0 \) and \( x^* \in X, x^* \neq 0 \).

I) If \( E^t_{\{R_\alpha\}}(x^*, \delta) = o(\varphi(\delta)) \) for \( \delta \to 0^+ \), then there exists an a-priori parameter choice rule \( \tilde{\alpha}(\delta) \) such that

\[
\sup_{y^\delta \in B_3(Tx^*)} \left\| R_{\tilde{\alpha}(\delta)} y^\delta - x^* \right\| = o(\varphi(\delta)) \quad \text{for} \quad \delta \to 0^+.
\]

II) Part I) remains true with \( o(\varphi(\delta)) \) replaced by \( O(\varphi(\delta)) \), that is, if \( E^t_{\{R_\alpha\}}(x^*, \delta) = O(\varphi(\delta)) \) for \( \delta \to 0^+ \), then there exists an a-priori parameter choice rule \( \tilde{\alpha}(\delta) \) such that

\[
\sup_{y^\delta \in B_3(Tx^*)} \left\| R_{\tilde{\alpha}(\delta)} y^\delta - x^* \right\| = O(\varphi(\delta)) \quad \text{for} \quad \delta \to 0^+.
\]

Proof. Let \( \varphi \) and \( x^* \in X \) be as in the hypotheses and suppose that \( E^t_{\{R_\alpha\}}(x^*, \delta) = o(\varphi(\delta)) \) for \( \delta \to 0^+ \). Then by definition of \( E^t_{\{R_\alpha\}} \),

\[
\lim_{\delta \to 0^+} \inf_{\alpha \in (0, \alpha_0)} \sup_{\varphi(\delta)} \left\| R_\alpha y^\delta - x^* \right\| = \lim_{\delta \to 0^+} \sup_{\varphi(\delta)} \left\| R_\alpha y^\delta - x^* \right\| = 0. \tag{13}
\]

For the sake of simplicity we define:

\[
\begin{align*}
    f(\alpha, \delta) & \doteq \sup_{y^\delta \in B_3(Tx^*)} \frac{\left\| R_\alpha y^\delta - x^* \right\|}{\varphi(\delta)} \quad \text{and} \quad q(\delta) \doteq \inf_{\alpha \in (0, \alpha_0)} f(\alpha, \delta),
\end{align*}
\]

so that with this notation \( \text{(13)} \) can be written simply as \( \lim_{\delta \to 0^+} q(\delta) = 0 \) and the objective is to prove the existence of an a-priori parameter choice rule \( \tilde{\alpha}(\delta) \) such that \( \lim_{\delta \to 0^+} f(\tilde{\alpha}(\delta), \delta) = 0 \). It can be easily proved that if for certain \( \delta_0 > 0 \), \( E^t_{\{R_\alpha\}}(x^*, \delta_0) = 0 \) then \( T^1 = 0 \). Hence \( q(\delta) > 0 \) for every \( \delta \in (0, +\infty) \). Also, \( q(\delta) \) is continuous for \( \delta \in (0, +\infty) \) since both \( E^t_{\{R_\alpha\}}(x^*, \delta) \) and \( \varphi(\delta) \) are continuous. Next, for \( n \in \mathbb{N} \) we define

\[
\delta_n = \sup \left\{ d > 0 : q(\delta) \leq \frac{1}{n} \forall \delta \in (0, d) \right\}.
\]

Clearly, \( \delta_n \downarrow 0 \) and since \( q \) is continuous for every \( n \in \mathbb{N} \) and every \( \delta \in (0, \delta_n] \), \( q(\delta) = \inf_{\alpha \in (0, \alpha_0)} f(\alpha, \delta) \leq \frac{1}{n} \). Then, there exists \( \alpha_n = \alpha_n(\delta_n) \in (0, \alpha_0) \) such that for all \( n \in \mathbb{N} \)

\[
f(\alpha_n, \delta) \leq \frac{2}{n} \quad \forall \delta \in (0, \delta_n]. \tag{14}
\]
Now, since \{\alpha_n\} \subset (0, \alpha_0) is a bounded sequence, there exist \alpha^* \in [0, \alpha_0] and \{\alpha_{n_k}\} \subset \{\alpha_n\} such that \lim_{k \to +\infty} \alpha_{n_k} = \alpha^*. We now define \bar{\alpha}(\delta) = \alpha_{n_k} for \delta \in (\delta_{n_k+1}, \delta_{n_k}], k = 1, 2, \ldots, and \bar{\alpha}(\delta) = \bar{\alpha}(\delta_{n_1}) for \delta > \delta_{n_1}. Then

$$\lim_{\delta \to 0^+} \bar{\alpha}(\delta) = \alpha^* \quad (15)$$

and

$$0 \leq \limsup_{\delta \to 0^+} f(\bar{\alpha}(\delta), \delta) \leq \limsup_{k \to +\infty} \left( \sup_{\delta \in (0, \delta_{n_k})} f(\alpha_{n_k}, \delta) \right) \leq \limsup_{k \to +\infty} \frac{2}{n_k} = 0,$$

where the last inequality follows from (14).

Hence,

$$\lim_{\delta \to 0^+} f(\bar{\alpha}(\delta), \delta) = 0. \quad (16)$$

It remains to be shown that \bar{\alpha}(\delta) is an admissible parameter choice rule, for which it suffices to prove that \lim_{\delta \to 0^+} \bar{\alpha}(\delta) = 0, i.e., that \alpha^* = 0. If \alpha^* > 0, it follows from (15) that there exists \delta_0 > 0 such that \bar{\alpha}(\delta) > \frac{\alpha^*}{2} for all \delta \in (0, \delta_0). Hypothesis b.iii) of Theorem 3.3 then implies that for every \delta \in (0, \delta_0), \left| r_{\alpha}(\delta)(\lambda) \right| \geq \left| r_{\alpha^*}(\lambda) \right| \text{ for all } \lambda \in (0, \|T\|^2]. Therefore for every \delta \in (0, \delta_0),

$$\left\| r_{\bar{\alpha}(\delta)}(T^*T)x^* \right\|^2 = \int_0^{\|T\|^2} r_{\bar{\alpha}(\delta)}^2(\lambda) \, d\|E_\lambda x^*\|^2 \geq \int_0^{\|T\|^2} r_{\alpha^*}^2(\lambda) \, d\|E_\lambda x^*\|^2 = \left\| r_{\alpha^*}(T^*T)x^* \right\|^2. \quad (17)$$

Now, for all \delta \in (0, \delta_0),

$$\sup_{y^* \in B_r(T^*T)x^*} \left\| R_{\bar{\alpha}(\delta)}y^* \right\| \geq \left\| R_{\bar{\alpha}(\delta)}Tx^* \right\| = \left\| (I - g_{\bar{\alpha}(\delta)}(T^*T)x^*)x^* \right\| = \left\| r_{\alpha^*}(T^*T)x^* \right\|,$$

where the last inequality follows from (17). Dividing through by \varphi(\delta), taking limit for \delta \to 0^+, and using the definition of \varphi(\delta) and (16) we conclude that \left\| r_{\alpha^*}(T^*T)x^* \right\| = 0. Now since \frac{\alpha^*}{2} < \alpha_0, (\rho, s) is an order-source pair for \{g_{\alpha}\} and hypothesis b.iii) of Theorem 3.3 holds, it follows from Lemma 3.4 that \left\| r_{\alpha^*}(T^*T)x^* \right\| is invertible. Therefore \left\| r_{\alpha^*}(T^*T)x^* \right\| = 0, contradicting the hypothesis that \alpha^* \neq 0. Hence, \alpha^* must be equal to zero, as wanted.

We proceed now to prove the second part of the Lemma. Suppose that there exists \left\| x^* \right\| \in X, \left\| x^* \right\| \neq 0 such that \left\| E_{\left\{R_{\alpha}\right\}}(x^*, \delta) = O(\varphi(\delta)) \right\| as \delta \to 0^+. Then there exist positive constants k and d such that \inf_{\alpha \in (0, \alpha_0)} f(\alpha, \delta) \leq k \text{ for every } \delta \in (0, d), \text{ where } f(\alpha, \delta) \text{ is as previously defined. Let}

$$\left\{\delta_n\right\}_{n \in \mathbb{N}} \subset (0, d) \text{ be such that } \delta_n \downarrow 0 \text{ and } \alpha_n = \alpha_n(\delta_n) \in (0, \alpha_0) \text{ such that }$$

$$f(\alpha_n, \delta) \leq k + \delta_n, \forall \delta \in (0, d), \forall n \in \mathbb{N}. \text{ Without loss of generality we assume that the sequence } \{\alpha_n\} \text{ converges (since if that is not the case, we can take a subsequence which does). Now, like in the previously case, by defining } \bar{\alpha}(\delta) = \alpha_n \text{ for
\[ \delta \in (\delta_{n+1}, \delta_n], n = 1, 2, \ldots, \text{and } \bar{\alpha}(\delta) = \alpha(\delta_1) \text{ for } \delta > \delta_1, \text{since } \delta_n \downarrow 0 \text{ it follows that } f(\bar{\alpha}(\delta), \delta) \leq k + \delta_1 \text{ for every } \delta \in (0, \delta_1) \text{ and therefore} \]

\[ \sup_{y^* \in B_2(Tx^*)} \left\| R_{\bar{\alpha}(\delta)} y^* - x^* \right\| = O(\varphi(\delta)) \text{ as } \delta \to 0^+. \]

Following the same steps as in the proof of Part I we obtain that \( \lim_{\delta \to 0^+} \bar{\alpha}(\delta) = 0 \), i.e. \( \bar{\alpha}(\delta) \) is an admissible parameter choice rule. This concludes the proof of the lemma.

**Lemma 3.6.** Let \( \{g_\alpha\}_{\alpha \in (0, \alpha_0)} \) be a SRM, \( r_\alpha(\lambda) = 1 - \lambda g_\alpha(\lambda), R_\alpha = g_\alpha(T^*T)^T \), \( (\rho, s) \) an order-source pair for \( \{g_\alpha\}, \Theta(t) = \sqrt{t} \rho(t) \) for \( t > 0 \), and suppose that:

\( \alpha \) The function \( \rho \) is of local upper type \( \beta \), for some \( \beta \geq 0 \).
\( b \) There exist positive constants \( \gamma_1, \gamma_2, \lambda^*, c_1 \), with \( \lambda^* \leq \|T\|^2 \) and \( c_1 > 1 \) such that

\( i \) \( 0 \leq r_\alpha(\lambda) \leq 1 \), for \( \alpha > 0 \), \( 0 \leq \lambda \leq \lambda^* \);
\( ii \) \( r_\alpha(\lambda) \geq \gamma_1 \), for \( 0 \leq \lambda < h(\alpha) \leq \lambda^* \), \( \alpha \in (0, \alpha_0) \) where \( h \) is as in (4);
\( iii \) \( |r_\alpha(\lambda)| \) is non-decreasing with respect to \( \alpha \) for each \( \lambda \in (0, \|T\|^2) \);
\( iv \) \( g_\alpha(\alpha \lambda) \geq \frac{2\alpha}{\alpha} \) for \( 0 < \alpha \lambda \leq \lambda^* \) and
\( v \) \( g_\alpha(\lambda) \geq g_\alpha(\tilde{\lambda}) \), for \( 0 < \lambda \leq \lambda \leq \lambda^* \).
\( c \) There exist \( \{\lambda_n\}_{n=1}^\infty \subset \sigma(TT^*) \) and \( \gamma \geq 1 \) such that \( \lambda_n \downarrow 0 \) and \( \frac{\lambda_n}{\lambda_{n+1}} \leq \gamma \) for every \( n \in \mathbb{N} \).

If for some \( x \in X \) we have that

\[ \sup_{y^* \in B_2(Tx)} \inf_{\alpha \in (0, \alpha_0)} \left\| R_\alpha y^* - x \right\| = O(\rho(\Theta^{-1}(\delta))) \text{ when } \delta \to 0^+, \] (18)

then \( x \in \mathcal{R}(s(T^*T)) \). In particular, if \( \rho \) is optimal qualification of \( \{g_\alpha\} \) and \( s_\rho \in \mathcal{S} \), then \( x \in \mathcal{R}(s_\rho(T^*T)) \).

**Proof.** Without loss of generality we may assume that \( \alpha_0 \leq \frac{\lambda^*}{\bar{\alpha}} \) and \( x \neq 0 \) (this is so because hypotheses \( a \) and \( c \) are independent of \( \alpha_0 \) and if \( b \) holds for \( \alpha \in (0, \alpha_0) \) then it holds for \( \alpha \in (0, \bar{\alpha_0}) \) for every \( \bar{\alpha_0} < \alpha_0 \) with the same constants, while if \( x = 0 \) the result of the Lemma is trivial).

Let \( \bar{\alpha} \in \sigma(TT^*) \) be such that \( 0 < \gamma \bar{\alpha} \leq \alpha_0 \) (hypothesis \( c \) guarantees the existence of such \( \bar{\alpha} \)), and define

\[ \tilde{\delta} = \tilde{\delta}(\bar{\alpha}) = \frac{\bar{\alpha}^{1/2}}{\gamma_2} \|R_\bar{\alpha}Tx - x\| = \frac{\bar{\alpha}^{1/2}}{\gamma_2} \|r_\bar{\alpha}(T^*T)x\|. \]

Then, clearly the equation

\[ \|R_\alpha Tx - x\|^2 = \left(\frac{\gamma_2 \tilde{\delta}}{\alpha}\right)^2 \] (19)

in the unknown \( \alpha \), has \( \alpha = \bar{\alpha} \) as a solution. Moreover, since \( \|R_\alpha Tx - x\|^2 = \int_0^{\|T\|^2} r^2_\alpha(\lambda) d\|E_\lambda x\|^2 \)

and \( x \neq 0 \), hypotheses \( b.ii \) and \( b.iii \) imply that the function \( \mu(\alpha) = \alpha \|R_\alpha Tx - x\|^2 \) is strictly increasing for \( \alpha \in (0, \alpha_0) \). Hence, \( \alpha = \bar{\alpha} = \eta(\tilde{\delta}) \) (where \( \eta(\tilde{\delta}) = \mu^{-1}(\tilde{\delta}) \) ) is the unique solution of (19). Note that if \( \bar{\alpha} \to 0^+ \) then \( \tilde{\delta} \to 0^+ \). Moreover, by hypothesis \( b.iii \) and Lemma 3.3 it follows immediately by Fatou’s Lemma that \( \tilde{\delta} \to 0^+ \) only if \( \bar{\alpha} \to 0^+ \).
Now, for $\delta > 0$ define
\[
\bar{y}^\delta \doteq Tx - \delta G_\alpha z,
\] (20)
where $G_\alpha \doteq F_{c_1\bar{\alpha}} - F_{\bar{\alpha}}$ with $\{F_\lambda\}$ being the spectral family associated to $TT^*$ and
\[
z \doteq \begin{cases} \|G_\alpha Tx\|^{-1} Tx, & \text{if } G_\alpha Tx \neq 0, \\ \text{arbitrary with } \|G_\alpha z\| = 1, & \text{in other case.} \end{cases}
\]
Note that since $\bar{\alpha} \in \sigma(TT^*)$ and $c_1 > 1$ it follows that $G_\alpha$ is not the null operator and therefore the definition makes sense. Note also that $\delta > 0$, which implies that $\bar{y}^\delta \in B_\delta(Tx)$.

Now, by using (20) and the fact that $g_\alpha(T^*T)^* = T^*g_\alpha(TT^*)$ it follows that for every $\alpha \in (0, \alpha_0)$ and $\delta > 0$,
\[
\langle R_\alpha Tx - x , R_\alpha(\bar{y}^\delta - Tx) \rangle = \langle g_\alpha(T^*T)^*Tx - x , -g_\alpha(T^*T)^* \delta G_\alpha z \rangle
\]
\[
= \delta \langle g_\alpha(T^*T)^*Tx - x , -T^*g_\alpha(TT^*)G_\alpha z \rangle
\]
\[
= \delta \langle Tg_\alpha(T^*T)^*Tx - x , -g_\alpha(TT^*)G_\alpha z \rangle
\]
\[
= \delta \langle (TT^*g_\alpha(TT^*) - I)Tx , -g_\alpha(TT^*)G_\alpha z \rangle
\]
\[
= \delta \langle -r_\alpha(TT^*)Tx , -g_\alpha(TT^*)G_\alpha z \rangle
\]
\[
= \delta \int_0^{\|T\|^2+} r_\alpha(\lambda)g_\alpha(\lambda) d \langle F_\lambda Tx , G_\alpha z \rangle.
\] (21)

Now by hypothesis $b.ii)$ and since $c_1\bar{\alpha} \leq \lambda^*$ one has that both $g_\alpha(\lambda)$ and $r_\alpha(\lambda)$ are nonnegative for all $\lambda \in [0, c_1\bar{\alpha}]$. Also, from the definitions of $G_\alpha$ and $z$ it follows immediately that the function $m(\lambda) = \langle F_\lambda Tx , G_\alpha z \rangle$ for $\lambda \in [0, c_1\bar{\alpha}]$ is real and non-decreasing and therefore
\[
\int_0^{c_1\bar{\alpha}} r_\alpha(\lambda)g_\alpha(\lambda) d \langle F_\lambda Tx , G_\alpha z \rangle \geq 0.
\] (22)

On the other hand, since $m(\lambda) = \langle Tx , F_\lambda G_\alpha z \rangle$ and $F_\lambda G_\alpha = G_\alpha$ for every $\lambda \geq c_1\bar{\alpha}$, it follows that $m(\lambda)$ is constant for $\lambda \geq c_1\bar{\alpha}$ and therefore
\[
\int_0^{\|T\|^2+} r_\alpha(\lambda)g_\alpha(\lambda) d \langle F_\lambda Tx , G_\alpha z \rangle = 0.
\] (23)

From (22) and (23) we conclude that
\[
\int_0^{\|T\|^2+} r_\alpha(\lambda)g_\alpha(\lambda) d \langle F_\lambda Tx , G_\alpha z \rangle \geq 0,
\]
which, by virtue of (21), implies that
\[
\langle R_\alpha Tx - x , R_\alpha(\bar{y}^\delta - Tx) \rangle \geq 0.
\]
(24)

Hence, for every $\alpha \in (0, \alpha_0)$, $\delta > 0$ and $\bar{\alpha} \in \sigma(TT^*)$ such that $c_1\bar{\alpha} \leq \lambda^*$ we obtain the following
estimate:

\[
\|R_\alpha \bar{y}^\delta - x\|^2 = \|R_\alpha Tx - x\|^2 + \|R_\alpha (\bar{y}^\delta - Tx)\|^2 + 2 \left\langle R_\alpha Tx - x, R_\alpha (\bar{y}^\delta - Tx) \right\rangle \\
\geq \|R_\alpha Tx - x\|^2 + \|T^* \alpha G_\alpha \bar{z}\|^2 + 2 \left\langle R_\alpha Tx - x, R_\alpha (\bar{y}^\delta - Tx) \right\rangle \quad \text{(by (20))}
\]

\[
\geq \|R_\alpha Tx - x\|^2 + \|T^* \alpha G_\alpha \bar{z}\|^2 \\
\geq \|R_\alpha Tx - x\|^2 + \|T^* \alpha G_\alpha \bar{z}\|^2 \\
\geq \|R_\alpha Tx - x\|^2 + \|G_\alpha \bar{z}\|^2.
\]

(25)

We now consider two different cases for \(\alpha \in (0, 0_0)\).

Case I: \(\alpha \leq \bar{\alpha}\). Since \(c_1 \bar{\alpha} \leq \lambda^*\) and \(c_1 > 1\), it follows from hypothesis \(b.v\) that

\[
g_\alpha(\lambda) \geq g_\alpha(c_1 \bar{\alpha}) \geq g_\alpha(\lambda^*) \quad \text{for every} \quad \lambda \in [\bar{\alpha}, c_1 \bar{\alpha}].
\]

(26)

On the other hand, from hypothesis \(b.i)\) it follows that \(\tau_\alpha(\lambda^*) \leq 1\), which implies that \(\lambda^* g_\alpha(\lambda^*) \geq 0\) and therefore, \(g_\alpha(\lambda^*) \geq 0\). It then follows from \(25\) that \(g_\alpha^2(\lambda) \geq \bar{g}_\alpha^2(c_1 \bar{\alpha})\) for every \(\lambda \in [\bar{\alpha}, c_1 \bar{\alpha}]\).

Then,

\[
\int_{c_1 \bar{\alpha}}^{\bar{\alpha}} \lambda g_\alpha^2(\lambda) d \|F_\lambda G_\alpha \bar{z}\|^2 \geq \bar{\alpha} \bar{g}_\alpha^2(c_1 \bar{\alpha}) \int_{c_1 \bar{\alpha}}^{\bar{\alpha}} d \|F_\lambda G_\alpha \bar{z}\|^2 \\
= \bar{\alpha} \bar{g}_\alpha^2(c_1 \bar{\alpha}) \left(\|F_\lambda G_\alpha \bar{z}\|^2 - \|F_\alpha G_\alpha \bar{z}\|^2\right) \\
= \bar{\alpha} \bar{g}_\alpha^2(c_1 \bar{\alpha}) \|G_\alpha \bar{z}\|^2 \\
= \bar{\alpha} \bar{g}_\alpha^2(c_1 \bar{\alpha}).
\]

(27)

where the second to last equality follows from the definition of \(G_\alpha\) and the spectral property \(F_\lambda F_\mu = F_{\min(\lambda, \mu)}\).

At the same time, the hypotheses \(b.i)\) and \(b.iii)\) imply that \(g_\alpha(\lambda)\) is non-increasing as a function of \(\alpha\) for each fixed \(\lambda \in [0, \lambda^*]\). Since \(\alpha \leq \bar{\alpha}\) and \(c_1 \bar{\alpha} \leq \lambda^*\), we then have that

\[
g_\alpha(c_1 \bar{\alpha}) \geq g_\alpha(c_1 \bar{\alpha}),
\]

(28)

and from hypothesis \(b.iv\) we also have that

\[
g_\alpha(c_1 \bar{\alpha}) \geq \frac{72}{\bar{\alpha}} > 0.
\]

(29)

From \(28\) and \(29\) we conclude that

\[
g_\alpha^2(c_1 \bar{\alpha}) \geq \left(\frac{72}{\bar{\alpha}}\right)^2.
\]

(30)

Substituting \(30\) into \(27\) we obtain \(\int_{c_1 \bar{\alpha}}^{\bar{\alpha}} \lambda g_\alpha^2(\lambda) d \|F_\lambda G_\alpha \bar{z}\|^2 \geq \frac{72}{\bar{\alpha}}\), which, by virtue of \(25\) implies that if \(\alpha \leq \bar{\alpha}\), then \(\|R_\alpha \bar{y}^\delta - x\|^2 \geq \frac{(72)^2}{\bar{\alpha}}\).
Case II: \( \alpha > \bar{\alpha} \). In this case, it follows from hypothesis \textit{b.iii)} that \( r^2(\lambda) \geq r^2(\lambda) \) for every \( \lambda \in (0, \|T\|^2) \). Then,

\[
\|R_\alpha Tx - x\|^2 = \int_0^{\|T\|^2} \|r^2(\lambda) d \|E \lambda x\|^2 \geq \int_0^{\|T\|^2} r^2(\lambda) d \|E \lambda x\|^2 = \|R_\alpha Tx - x\|^2,
\]

which, together with (25) imply that \( \|R_\alpha \bar{y}^\delta - x\|^2 \geq \|R_\alpha Tx - x\|^2 \).

Summarizing the results of cases I and II, we obtain that for every \( \alpha \in (0, \alpha_0) \), \( \delta > 0 \), \( \bar{\alpha} \in \sigma(TT^*) \) with \( c_1 \bar{\alpha} \leq \alpha_0 \) and \( \bar{y}^\delta \) as in (20), there holds:

\[
\|R_\alpha \bar{y}^\delta - x\|^2 \geq \begin{cases} 
\frac{(\gamma_2 \delta)^2}{\bar{\alpha}}, & \text{if } 0 < \alpha \leq \bar{\alpha}, \\
\|R_\alpha Tx - x\|^2, & \text{if } \bar{\alpha} < \alpha \leq \alpha_0 
\end{cases} \\
\geq \min \left\{ \|R_\alpha Tx - x\|^2, \frac{(\gamma_2 \delta)^2}{\bar{\alpha}} \right\}.
\]

Then

\[
\min \left\{ \|R_\alpha Tx - x\|, \frac{\gamma_2 \delta}{\sqrt{\bar{\alpha}}} \right\} = \left( \min \left\{ \|R_\alpha Tx - x\|^2, \frac{(\gamma_2 \delta)^2}{\bar{\alpha}} \right\} \right)^{1/2} \\
\leq \inf_{\alpha \in (0, \alpha_0)} \|R_\alpha \bar{y}^\delta - x\| \\
\leq \sup_{\bar{y}^\delta \in B_3(Tx)} \inf_{\alpha \in (0, \alpha_0)} \|R_\alpha \bar{y}^\delta - x\| \quad (\text{since } \bar{y}^\delta \in B_3(Tx)) \\
= \Theta(\rho(\Theta^{-1}(\delta))) \quad \text{for } \delta \to 0^+ \quad (\text{by hypothesis}),
\]

and since \( \bar{\alpha} = \eta(\delta) \) solves equation (19), the previous inequality implies that

\[
\|R_{\eta(\delta)} Tx - x\| = \frac{\gamma_2 \delta}{\sqrt{\bar{\alpha}}} = \Theta(\rho(\Theta^{-1}(\delta))) \quad \text{for } \delta \to 0^+,
\]

and therefore

\[
\frac{\delta}{\rho(\Theta^{-1}(\delta))} = \Theta \left( \sqrt{\eta(\delta)} \right) \quad \text{for } \delta \to 0^+.
\]

Now, since for every \( \delta > 0 \) one has \( \delta = \Theta(\Theta^{-1}(\delta)) \), it follows from the definition of \( \Theta \) that \( \delta = \Theta^{-1}(\delta) \rho(\Theta^{-1}(\delta)) \). Then, from (33) we obtain that \( \sqrt{\Theta^{-1}(\delta)} = \Theta(\sqrt{\eta(\delta)}) \) for \( \delta \to 0^+ \). From this and (32) we then deduce that

\[
\|R_{\eta(\delta)} Tx - x\| = \Theta(\rho(\eta(\delta))) \quad \text{for } \delta \to 0^+, \quad \delta = \mu(\bar{\alpha}), \quad \bar{\alpha} \in \sigma(TT^*), \quad c_1 \bar{\alpha} \leq \alpha_0.
\]

Now let \( L = \max \left\{ \lambda_j : \lambda_j \leq \frac{20}{\bar{\alpha}} \right\} \). Then, since \( \lambda_n \downarrow 0 \), for any \( \alpha \in (0, L] \) there exist a unique \( n = n(\alpha) \in \mathbb{N} \) such that \( \lambda_{n+1} < \alpha \leq \lambda_n \) (note that \( n(\alpha) \to \infty \) if and only if \( \alpha \to 0^+ \)). Then for
\( \alpha \in (0, L] \) and \( n = n(\alpha) \) so defined we have that
\[
\| R_\alpha Tx - x \|^2 = \int_0^\|T\|^2 + r_\alpha^2(\lambda) \, d \| E_\lambda x \|^2 \\
\leq \int_0^\|T\|^2 + r_\alpha^2(\lambda) \, d \| E_\lambda x \|^2, \quad \text{(by hypothesis b.iii)} \\
= \| R_\lambda Tx - x \|^2 \\
= O(\rho^2(\lambda)), \quad \text{(by virtue of (34), with } \bar{\delta} = \mu(\lambda)). \tag{35}
\]
Also, from hypothesis c) we have that \( \lambda_n \leq c \lambda_{n+1} \) for all \( n \in \mathbb{N} \), and since \( \rho \) is non-decreasing and positive (since \( \rho \in \mathcal{O} \)) it follows that
\[
\rho^2(\lambda_n) \leq \rho^2(c \lambda_{n+1}), \quad \forall n \in \mathbb{N}. \tag{36}
\]
Now since \( c \geq 1 \) and by hypothesis a) \( \rho \) is of local upper type \( \beta \), there exists a positive constant \( d \) such that
\[
\rho(c \lambda_{n+1}) \leq d \, c^\beta \rho \left( \frac{1}{c} \lambda_{n+1} \right) = d \, c^\beta \rho(\lambda_{n+1}), \quad \forall n \in \mathbb{N}. \tag{37}
\]
Hence
\[
\rho \left( \lambda_{n(\alpha)} \right) \leq \rho \left( c \lambda_{n(\alpha) + 1} \right) \quad \text{(by (36))} \\
\leq d \, c^\beta \rho \left( \lambda_{n(\alpha) + 1} \right) \quad \text{(by (37))} \\
\leq d \, c^\beta \rho(\alpha) \quad \text{(since } \lambda_{n(\alpha) + 1} < \alpha \text{ and } \rho \in \mathcal{O}). \tag{38}
\]
From (35) and (38) it follows that \( \| R_\alpha Tx - x \| = O(\rho(\alpha)) \) for \( \alpha \to 0^+ \). Since \( T^\dagger T \) is the projection on \( N(T)^\perp = X \) (since \( T \) is invertible), we have that \( x = T^\dagger Tx = T^\dagger y \). Then \( \| (R_\alpha - T^\dagger)y \| = O(\rho(\alpha)) \) for \( \alpha \to 0^+ \). Finally, Theorem 2.13 implies that \( T^\dagger y = x \in \mathcal{R}(s(T^*T)) \).
This concludes the proof of the Lemma. \( \square \)

**Remark 3.7.** Note that since \( \sup_{y^\delta \in B_\delta(Tx)} \inf_{\alpha \in (0, \alpha_0)} \| R_\alpha y^\delta - x \| \leq \mathcal{E}^{\text{tot}}_{\{R_\alpha\}}(x, \delta), \text{ hypothesis (18) of the preceding Lemma holds if } \mathcal{E}^{\text{tot}}_{\{R_\alpha\}}(x, \delta) = O(\rho(\Theta^{-1}(\delta))) \text{ for } \delta \to 0^+. \)

Having stated and proved the previous three lemmas, we are now ready to prove Theorem 3.3.

**Proof of Theorem 3.3.** As in Lemma 3.6 without loss of generality we may assume that \( \alpha_0 \leq \frac{L^*}{c_1} \).
We will show that \( \psi(x, \delta) = \rho \circ \Theta^{-1}(\delta) \) for \( x \in X^{s^p} \) and \( \delta \in (0, \Theta(\alpha_0)) \), is saturation function of \( \{R_\alpha\}_{\alpha \in (0, \alpha_0)} \) on \( X^{s^p} \) (see Definition 2.1).

First we note that since \( \{g_\alpha\} \) satisfies (H4) and \( \rho \) is continuous (\( \rho \) being of local upper type), by virtue of Lemma 3.1 one has that \( \psi \in \mathcal{U}_{X^{s^p}}(\mathcal{E}^{\text{tot}}_{\{R_\alpha\}}) \). Next we will show that \( \psi \) satisfies the SL condition for saturation on \( X^{s^p} \). Suppose that it is not true, i.e. that there exist \( x^* \in X, x^* \neq 0 \) and \( x \in X^{s^p} \) such that \( \limsup_{\delta \to 0^+} \frac{\mathcal{E}^{\text{tot}}_{\{R_\alpha\}}(x^*, \delta)}{\psi(x, \delta)} = 0 \). Then \( \mathcal{E}^{\text{tot}}_{\{R_\alpha\}}(x^*, \delta) = o(\psi(x, \delta)) \) as \( \delta \to 0^+ \) and from Lemma 3.5 I) it follows that there exists an a-priori parameter choice rule \( \hat{\alpha}(\delta) \) such that
\[
\sup_{y^\delta \in B_\delta(Tx^*)} \| R_\hat{\alpha}(\delta)y^\delta - x^* \| = o(\psi(x, \delta)) \quad \text{for } \delta \to 0^+. \tag{39}
\]
On the other hand, from hypothesis c) it follows that there exists \( \bar{\alpha} \in \sigma(TT^*) \) such that 
\[ 0 < c_1 \bar{\alpha} \leq \alpha_0 \text{ and } h(\bar{\alpha}) < \|T\|^2. \]

Now define
\[
\bar{\delta} = \delta(\bar{\alpha}) \triangleq \frac{\bar{\alpha}^{1/2}}{\gamma_2} \|R_{\bar{\alpha}}Tx^* - x^*\| = \frac{\bar{\alpha}^{1/2}}{\gamma_2} \|r_{\bar{\alpha}}(T^*T)x^*\|
\]
and for \( \delta > 0 \),
\[
\bar{y}^\delta \triangleq Tx^* - \delta G_\bar{\alpha}z,
\]

as in (20). Following the same steps as in the proof of Lemma 3.6 we obtain as in (34) that for \( \delta > 0 \), \( \bar{\alpha} \in \sigma(TT^*) \) with \( c_1\bar{\alpha} \leq \alpha_0 \) and \( h(\bar{\alpha}) < \|T\|^2 \), and \( \bar{y}^\delta \) as in (40), there holds:
\[
\|R_{\bar{\alpha}}\bar{y}^\delta - x^*\|^2 \geq \min \left\{ \|R_{\bar{\alpha}}Tx^* - x^*\|^2, \frac{(\gamma_2 \delta)^2}{\bar{\alpha}} \right\}.
\]

Then for \( \delta > 0 \) such that \( \bar{\alpha}(\delta) \in (0, \alpha_0) \),
\[
\min \left\{ \|R_{\bar{\alpha}}Tx^* - x^*\|, \frac{\gamma_2 \delta}{\sqrt{\bar{\alpha}}} \right\} = \left( \min \left\{ \|R_{\bar{\alpha}}Tx^* - x^*\|^2, \frac{(\gamma_2 \delta)^2}{\bar{\alpha}} \right\} \right)^{1/2}
\leq \inf_{\alpha \in (0, \alpha_0)} \|R_{\bar{\alpha}}\bar{y}^\delta - x^*\| \quad \text{(by (41))}
\leq \inf_{\bar{\alpha} \in (0, \alpha_0)} \|R_{\bar{\alpha}}\bar{y}^\delta - x^*\| (since \( \bar{\alpha}(\delta) \in (0, \alpha_0) \))
\leq \sup_{\bar{y}^\delta \in B_{\delta}(T^*x^*)} \|R_{\bar{\alpha}}\bar{y}^\delta - x^*\| (since \( \bar{y}^\delta \in B_\delta(T^*x^*) \))
\leq o(\rho(\Theta^{-1}(\delta))) \quad \text{for } \delta \to 0^+
\]
and since \( \bar{\alpha} = \eta(\bar{\delta}) \) solves equation (19) with \( x = x^* \), the previous inequality implies that
\[
\|R_{\bar{\alpha}}Tx^* - x^*\| = \frac{\gamma_2 \bar{\delta}}{\sqrt{\bar{\alpha}}} = o(\rho(\Theta^{-1}(\delta))) \quad \text{for } \bar{\delta} \to 0^+.
\]

Following analogous steps as in the proof of Lemma 3.6 we obtain, as in (34), that
\[
\|R_{\eta(\bar{\delta})}Tx^* - x^*\| = o(\rho(\eta(\bar{\delta}))) \quad \text{for } \bar{\delta} \to 0^+, \quad \bar{\delta} = \mu(\bar{\alpha}), \bar{\alpha} \in \sigma(TT^*), c_1 \bar{\alpha} \leq \alpha_0, \ h(\bar{\alpha}) < \|T\|^2.
\]

Now,
\[
\|R_{\bar{\alpha}}Tx^* - x^*\|^2 = \int_0^{\|T\|^2} r_{\bar{\alpha}}^2(\lambda) d \|E_\lambda x^*\|^2
\geq \int_{h(\bar{\alpha})}^{\|T\|^2} r_{\bar{\alpha}}^2(\lambda) d \|E_\lambda x^*\|^2, \quad \text{(since } 0 < h(\bar{\alpha}) < \|T\|^2 \text{)}
\geq \gamma^2 \rho^2(\bar{\alpha}) \int_{h(\bar{\alpha})}^{\|T\|^2} s_{\bar{\rho}}^{-2}(\lambda) d \|E_\lambda x^*\|^2,
\]
where the last inequality follows from the fact that \((\rho, s_\rho)\) is an order-source pair for \(\{g_\alpha\}\) (with \(\gamma\) the constant in (4)). Since \( \eta(\bar{\delta}) = \bar{\alpha} \) and \( \bar{\delta} \to 0^+ \) if and only if \( \bar{\alpha} \to 0^+ \), (42) and (43) imply that
\[
\int_{h(\bar{\alpha})}^{\|T\|^2} s_{\bar{\rho}}^{-2}(\lambda) d \|E_\lambda x^*\|^2 = o(1) \quad \text{for } \bar{\alpha} \to 0^+
\]
and therefore $\| s_{\rho}^{-1}(T^*T)x^* \| = 0$. Then $x^* = 0$, contradicting the fact that $x^* \neq 0$. Hence, $\psi(x, \delta) = \rho(\Theta^{-1}(\delta))$ satisfies condition $S1$ on $X^{s_\rho}$.

Also, since $\tilde{\psi}$ is trivially invariant over $X^{s_\rho}$, $\tilde{\psi}$ does not depend on $x$. Thus, it satisfies condition $S2$.

It only remains to be proved that $\tilde{\psi}$ satisfies condition $S3$ on $X^{s_\rho}$. Suppose that is not the case. Then there must exist a set $M, X^{s_\rho} \supsetneq M \subset X \setminus \{0\}$ and $\check{\psi} \in U_M(\mathcal{E}_{\{R_\alpha\}}^{\text{tot}})$ such that $\check{\psi}|_{X^{s_\rho}} = \psi$ and $\check{\psi}$ satisfies $S1$ and $S2$ on $M$. Let $x^* \in M \setminus X^{s_\rho}$. Since $\check{\psi} \in U_M(\mathcal{E}_{\{R_\alpha\}}^{\text{tot}})$, we have that

$$\mathcal{E}_{\{R_\alpha\}}^{\text{tot}} \{x^*\} \lesssim \tilde{\psi}. \tag{44}$$

Also, since $\tilde{\psi}$ is invariant over $M$, we have that $\tilde{\psi} \{x^*\}, X^{s_\rho} \lesssim \tilde{\psi}$, and since $\tilde{\psi}$ coincides with $\psi$ on $X^{s_\rho}$, it follows that $\check{\psi} \{x^*\}, X^{s_\rho} \lesssim \check{\psi}$. This, together with (44) implies that $\mathcal{E}_{\{R_\alpha\}}^{\text{tot}} \lesssim \check{\psi}$ and therefore for every $x \in X^{s_\rho}$, $\mathcal{E}_{\{R_\alpha\}}^{\text{tot}} (x^*, \delta) = O(\psi(x, \delta))$ as $\delta \to 0^+$, that is, $\mathcal{E}_{\{R_\alpha\}}^{\text{tot}} (x^*, \delta) = O(\rho(\Theta^{-1}(\delta)))$ as $\delta \to 0^+$. Lemma 3.3 then implies that there exists an admissible $a$-priori parameter choice rule $\tilde{\alpha}(\delta)$ such that

$$\sup_{y^\rho \in B_{\tilde{\alpha}}(T^*T)} \| R_{\tilde{\alpha}(\delta)} y^\rho - x^* \| = O(\rho(\Theta^{-1}(\delta))) \quad \text{as } \delta \to 0^+$$

and therefore

$$\sup_{y^\rho \in B_{\tilde{\alpha}}(T^*T)} \inf_{\alpha \in (0, \alpha_0)} \| R_{\alpha} y^\rho - x^* \| = O(\rho(\Theta^{-1}(\delta))) \quad \text{as } \delta \to 0^+.$$

Hence, by virtue of Lemma 3.3, $x^* \in R(s_\rho(T^*T))$ and since $x^* \neq 0$, we have that $x^* \in R(s_\rho(T^*T)) \setminus \{0\} = X^{s_\rho}$ which contradicts our original assumption. This concludes the proof of the Theorem 3.3.

### 4 Examples

Although the main results of this article are very theoretical in nature, we provide below two examples of regularization methods with optimal qualification which do possess saturation. In both cases the saturation function and saturation set are found.

**Example 4.1.** The family of Tikhonov-Phillips regularization operators $\{R_\alpha\}_{\alpha \in (0, \alpha_0)}$ where $R_\alpha = g_\alpha(T^*T)T^* \text{ with } g_\alpha(\lambda) = \frac{1}{\lambda + \alpha}$ and $g_\alpha(\lambda) = \lambda \in S$ and $\{g_\alpha\}_{\alpha \in (0, \alpha_0)}$ has optimal qualification $\rho(\alpha) = \alpha$ (32)). It can be easily checked that $\rho$ is of local upper type 1, $s_\rho(\lambda) = \lambda \in S$ and $\{g_\alpha\}_{\alpha \in (0, \alpha_0)}$ satisfies all hypotheses of the Theorem 3.3. Therefore, the function $\tilde{\psi}(x, \delta) = \rho(\Theta^{-1}(\delta)) = \delta^{\frac{3}{4}}$ defined for $x \in X^{s_\rho} \equiv R(T^*T) \setminus \{0\}$ and $\delta \in \left(0, \alpha_0^\frac{3}{4}\right)$ is saturation function of $\{R_\alpha\}_{\alpha \in (0, \alpha_0)}$ on $X^{s_\rho}$.

**Example 4.2.** Given $k \in \mathbb{R}^+$, for $\alpha > 0$ let

$$u_\alpha^k(\lambda) = \begin{cases} \frac{\sqrt{\alpha}}{\lambda^{\frac{3}{2}}}, & \text{for } 0 < \lambda < \alpha, \\ \frac{\sqrt{\alpha}}{\lambda^{\frac{3}{2}}}, & \text{for } \alpha \leq \lambda < 3\alpha, \\ \frac{\sqrt{\alpha}}{\lambda^{\frac{3}{2}}} + \alpha^{k+1}, & \text{for } \lambda \geq 3\alpha, \end{cases}$$

where $\alpha \in (0, \alpha_0)$ and $\alpha_0 \in (0, 1)$.
and $g^k_\alpha(\lambda) = \frac{1}{\lambda} - \alpha^k \sqrt{\lambda} - u^k_\alpha(\lambda)$ for $\lambda > 0$, and for $\lambda = 0$ define $g^k_\alpha(0) = \lim_{\lambda \to 0^+} g^k_\alpha(\lambda) = \frac{1}{\sqrt{\alpha}}$. The family $\{g_\alpha\}_{\alpha \in (0, \alpha_0)}$ is a SRM (\cite{3}).

Now, defining

$$v^k_\alpha(\lambda) = \begin{cases} e^{-\sqrt{\alpha} \lambda}, & \text{for } 0 \leq \lambda < \alpha, \\ e^{-\sqrt{\alpha} \lambda}, & \text{for } \alpha \leq \lambda < 3\alpha, \\ e^{-\sqrt{\alpha} \lambda} + \left(\frac{\alpha}{\lambda}\right)^k, & \text{for } \lambda \geq 3\alpha, \end{cases}$$

it follows that $r^k_\alpha(\lambda) = 1 - \lambda g^k_\alpha(\lambda) = \alpha^k \lambda^{3\alpha} + v^k_\alpha(\lambda)$. If $\rho(\alpha) = \alpha^k$ then $s_\rho(\lambda) = \lambda^k$, then $s_\rho \in \mathcal{S}$ and $s_\rho$ satisfies (b.v). Also, it can be show that $\rho$ and $s_\rho$ verify (4) with $\gamma = 1$ and $h(\alpha) = 3\alpha$. From Theorem 2.11 it then follows that $\rho(\alpha) = \alpha^k$ is optimal qualification of $\{g_\alpha\}$.

On the other hand, for $k \geq 1$ and $\alpha > 0$, the function $g^k_\alpha(\lambda)$ is non-increasing. Thus, hypothesis b.v of Theorem 2.11 holds and $G^k_\alpha = \|g^k_\alpha(\cdot)\|_{\infty} = g^k_\alpha(0) = \frac{1}{\sqrt{\alpha}}$, which implies immediately that also hypothesis $H_4$ is verified. From now on we shall assume $k \geq 1$.

It can be easily checked that $\rho$ is of local upper type $k$ and $\{g_\alpha\}_{\alpha \in (0, \alpha_0)}$ satisfies all hypotheses of the Theorem 3.3. Therefore, the function $\psi(x, \delta) = \rho(\Theta^{-1}(\delta)) = \delta^{\frac{2k^k}{2k+1}}$ defined for $x \in X_{s^\rho} = \mathcal{R}(T^*T)^k \setminus \{0\}$ and $\delta \in \left(0, \alpha_0^{k+\frac{1}{2}}\right)$ is saturation function of $\{R_\alpha\}_{\alpha \in (0, \alpha_0)}$ on $X_{s^\rho}$.

5 Conclusions

In this article families of real functions $\{g_\alpha\}$ defining a spectral regularization methods with optimal qualification were considered. Sufficient conditions on the family and on the optimal qualification guaranteeing the existence of saturation were found. Appropriate characterizations of both the saturation function and the saturation set were given and two examples were provided.

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