

## A discrepancy principle for Tikhonov regularization with approximately specified data

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**Abstract.** Many discrepancy principles are known for choosing the parameter  $\alpha$  in the regularized operator equation  $(T^*T + \alpha I)x_\alpha^\delta = T^*y^\delta$ ,  $\|y - y^\delta\| \leq \delta$ , in order to approximate the minimal norm least-squares solution of the operator equation  $Tx = y$ . We consider a class of discrepancy principles for choosing the regularization parameter when  $T^*T$  and  $T^*y^\delta$  are approximated by  $A_n$  and  $z_n^\delta$  respectively with  $A_n$  not necessarily self-adjoint. This procedure generalizes the work of Engl and Neubauer (1985), and particular cases of the results are applicable to the regularized projection method as well as to a degenerate kernel method considered by Groetsch (1990).

**1. Introduction.** We are concerned with the problem of finding approximations to the minimal norm least-squares solution  $\hat{x}$  of the operator equation

$$(1.1) \quad Tx = y,$$

where  $T : X \rightarrow Y$  is a bounded linear operator between Hilbert spaces  $X$  and  $Y$ , and  $y$  belongs to  $D(T^\dagger) := R(T) + R(T)^\perp$ , the domain of the Moore–Penrose inverse  $T^\dagger$  of  $T$ . It is well known [8] that if the range  $R(T)$  of  $T$  is not closed, then the operator  $T^\dagger$  which associates  $y \in D(T^\dagger)$  to  $\hat{x} := T^\dagger y$ , the unique least-squares solution of minimal norm, is not continuous, and consequently the problem of solving (1.1) for  $\hat{x}$  is ill-posed. A prototype of an ill-posed problem is the Fredholm integral equation of the first kind

$$(1.2) \quad \int_0^1 k(s, t)x(t) dt = y(s), \quad 0 \leq s \leq 1,$$

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with nondegenerate kernel  $k(\cdot, \cdot) \in L^2([0, 1] \times [0, 1])$ , where  $X = Y = L^2[0, 1]$ . Regularization methods are employed to find approximations to  $\hat{x}$ . In Tikhonov regularization one looks for the unique  $x_\alpha$ ,  $\alpha > 0$ , which minimizes the functional

$$x \rightarrow \|Tx - y\|^2 + \alpha\|x\|^2, \quad x \in X.$$

Equivalently, one solves the well-posed equation

$$(1.3) \quad (T^*T + \alpha I)x_\alpha = T^*y$$

for each  $\alpha > 0$ . Since  $T^*T\hat{x} = T^*y$ , it follows that

$$(1.4) \quad \|\hat{x} - x_\alpha\| = \|\alpha(T^*T + \alpha I)^{-1}\hat{x}\| \leq \|\hat{x}\|.$$

It is known ([8], [16]) that

$$(1.5) \quad \|\hat{x} - x_\alpha\| \rightarrow 0 \quad \text{as } \alpha \rightarrow 0$$

and

$$(1.6) \quad \hat{x} \in R((T^*T)^\nu), \quad 0 \leq \nu \leq 1, \quad \text{implies} \quad \|\hat{x} - x_\alpha\| = O(\alpha^\nu).$$

In practical applications the data  $y$  may not be available exactly, instead one may have an approximation  $y^\delta$  with say  $\|y - y^\delta\| \leq \delta$ ,  $\delta > 0$ . Then one solves the equation

$$(1.7) \quad (T^*T + \alpha I)x_\alpha^\delta = T^*y^\delta$$

instead of (1.3) and requires  $\|\hat{x} - x_\alpha^\delta\| \rightarrow 0$  as  $\alpha \rightarrow 0$  and  $\delta \rightarrow 0$ . It follows from (1.3) and (1.7) that

$$\begin{aligned} \|x_\alpha - x_\alpha^\delta\|^2 &= \|(T^*T + \alpha I)^{-1}T^*(y - y^\delta)\|^2 \\ &= \langle (T^*T + \alpha I)^{-1}T^*(y - y^\delta), (T^*T + \alpha I)^{-1}T^*(y - y^\delta) \rangle \\ &= \langle (TT^* + \alpha I)^{-2}TT^*(y - y^\delta), (y - y^\delta) \rangle \\ &\leq \|(TT^* + \alpha I)^{-2}TT^*\| \cdot \|y - y^\delta\|^2 \leq \delta^2/\alpha, \end{aligned}$$

so that

$$(1.8) \quad \|\hat{x} - x_\alpha^\delta\| \leq \|\hat{x} - x_\alpha\| + \delta/\sqrt{\alpha}.$$

Now let  $R_\alpha = (T^*T + \alpha I)^{-1}T^*$  for  $\alpha > 0$ . Then by (1.5) we have

$$\|R_\alpha y - T^\dagger y\| \rightarrow 0 \quad \text{as } \alpha \rightarrow 0$$

for  $y \in D(T^\dagger)$ . Therefore, if  $R(T)$  is not closed, then the family  $\{R_\alpha\}_{\alpha>0}$  is not uniformly bounded so that, as a consequence of the Uniform Boundedness Principle, there exists  $v \in Y$  such that  $\{R_\alpha v\}_{\alpha>0}$  is not bounded in  $Y$ . In particular, if  $y^\delta = y + \delta v/\|v\|$ , then  $\|y - y^\delta\| \leq \delta$  and  $\{R_\alpha y^\delta\}_{\alpha>0}$  is unbounded in  $Y$ . Therefore, the problem of choosing the regularization parameter  $\alpha$  depending on  $y^\delta$  is important. Many works in the literature are devoted to this aspect (cf. [7], [17], [1], [2], [3], [6], [14], [4]).

In order to solve (1.7) numerically, it is required to consider approximations of  $T^*T$  and of  $T^*y^\delta$ . So the problem actually at hand would be of the form

$$(1.9) \quad (A_n + \alpha I)x_{\alpha,n}^\delta = z_n^\delta,$$

where  $(A_n)$  and  $(z_n^\delta)$  are approximations of  $T^*T$  and of  $T^*y^\delta$  respectively.

In the well known regularized projection methods (cf. [10], [2], [3]),

$$A_n = P_n T^* T P_n \quad \text{and} \quad z_n^\delta = P_n T^* y^\delta,$$

where  $(P_n)$  is a sequence of orthogonal projections on  $X$  such that  $P_n \rightarrow I$  pointwise. In this case we have

$$\|T^*T - A_n\| \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

and discrepancy principles are known for choosing the regularization parameter  $\alpha$  in (1.9) (see e.g. [2], [3], [13], [5]).

In the degenerate kernel methods for the integral equation (1.2) with  $k(\cdot, \cdot) \in C([0, 1] \times [0, 1])$ ,  $A_n$  is obtained by approximating the kernel  $\tilde{k}(\cdot, \cdot)$  of the integral operator  $T^*T$  by a degenerate kernel  $\tilde{k}_n(\cdot, \cdot)$  so that  $\|\tilde{k} - \tilde{k}_n\|_\infty \rightarrow 0$  as  $n \rightarrow \infty$ . Then it follows that

$$\|T^*T - A_n\| \leq \|\tilde{k} - \tilde{k}_n\|_2 \leq \|\tilde{k} - \tilde{k}_n\|_\infty \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

(See [11] and [12] for a discussion on degenerate kernel methods for integral equations.) In a degenerate kernel method considered by Groetsch [9] the approximation  $\tilde{k}_n(\cdot, \cdot)$  is obtained from

$$\tilde{k}(s, t) := \int_0^1 k(\tau, s)k(\tau, t) dt, \quad a \leq s, t \leq b.$$

by using a convergent quadrature rule. In this case one has  $\|\tilde{k} - \tilde{k}_n\|_\infty \rightarrow 0$  as  $n \rightarrow \infty$  for nice enough kernels  $k(\cdot, \cdot)$ .

Moreover, for the degenerate kernel method of Groetsch [9] as well as for the regularized projection methods, the operators  $A_n$  are non-negative and self-adjoint.

In this paper we consider the generalized form of a class of discrepancy principles in [1], namely,

$$(1.10) \quad \|A_n x_{\alpha,n}^\delta - z_n^\delta\| = \frac{\delta^p}{\alpha^q}, \quad p > 0, q > 0,$$

for large enough  $n$ , to choose the regularization parameter  $\alpha = \alpha(n, \delta)$  in (1.9), where  $(A_n)$  is a sequence of bounded linear operators on  $X$  and  $(z_n^\delta)$  in  $X$  such that

$$\|T^*T - A_n\| \rightarrow 0 \quad \text{and} \quad \|T^*y^\delta - z_n^\delta\| \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

It has to be observed that we do not assume the operators  $A_n$  to be non-negative and self-adjoint. The consideration of a general  $A_n$ , as has been done recently by Nair [15], is important from the computational point of view, because even if one starts with a non-negative self-adjoint operator as approximation of  $T^*T$ , due to truncation errors etc., one actually may not be dealing with a non-negative self-adjoint operator.

With  $\alpha$  chosen according to (1.10), we show the convergence of the solution  $x_{\alpha,n}^\delta$  of (1.9) to  $\hat{x}$  as  $\delta \rightarrow 0$ ,  $n \rightarrow \infty$ , and also obtain estimates for the error  $\|\hat{x} - x_{\alpha,n}^\delta\|$  whenever  $\hat{x} \in R((T^*T)^\nu)$ ,  $0 < \nu \leq 1$ . Our result on error estimates shows that if  $\nu_0$  is an estimate for the possibly unknown  $\nu$ , with  $0 < \nu \leq \nu_0 \leq 1$ , then taking  $p/(q+1) = 2/(2\nu_0+1)$  one obtains the rate  $O(\delta^{2\nu/(2\nu_0+1)})$ . In particular, prior knowledge of  $\nu$  enables us to obtain the *optimal* rate  $O(\delta^{2\nu/(2\nu+1)})$  (cf. Schock [16]).

If  $A_n = P_n T^* T P_n$  and  $z_n^\delta = P_n T^* y^\delta$  then (1.10) coincides with a discrepancy principle considered by Engl and Neubauer [2] and we recover the optimal result in [2] as a particular case. Thus this paper generalizes the type of results in [2] and [9] for projection methods and degenerate kernel method for integral equations respectively, providing also a parameter choice strategy in the latter case.

**2. Approximate solution and convergence.** Let  $X$  and  $Y$  be Hilbert spaces and  $T : X \rightarrow Y$  be a bounded linear operator with its range  $R(T)$  not necessarily closed in  $Y$ . Let  $y \in D(T^\dagger) := R(T) + R(T)^\perp$ ,  $y \neq 0$ , so that there exists a unique  $\hat{x} \in X$  of minimal norm such that

$$\|T\hat{x} - y\| = \inf\{\|Tx - y\| : x \in X\}.$$

Let  $(A_n)$  be a sequence of bounded linear operators on  $X$  and for  $\delta > 0$ , let  $y^\delta \in Y$  and  $(z_n^\delta)$  in  $X$  be such that

$$\|T^*T - A_n\| \leq \varepsilon_n, \quad \|y - y^\delta\| \leq \delta, \quad \|T^*y^\delta - z_n^\delta\| \leq \eta_n^\delta,$$

where  $(\varepsilon_n)$  and  $(\eta_n^\delta)$  are sequences of nonnegative real numbers such that

$$\varepsilon_n \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

and

$$(2.1) \quad \eta_n^\delta \rightarrow 0 \quad \text{as } n \rightarrow \infty \text{ and } \delta \rightarrow 0.$$

Throughout the paper we denote the operator  $T^*T$  by  $A$ , and  $c, c', c_1, c_2$ , etc., denote positive constants which may assume different values in different contexts.

**THEOREM 2.1.** *If  $\varepsilon_n \leq c_0\alpha$  with  $0 < c_0 < 1$ , then  $A_n + \alpha I$  is bijective and*

$$\|(A_n + \alpha I)^{-1}\| \leq 1/(\alpha(1 - c_0)).$$

Moreover, if  $x_\alpha^\delta$  and  $x_{\alpha,n}^\delta$  are the unique solutions of (1.7) and (1.9) respectively, then

$$(2.2) \quad \|\widehat{x} - x_{\alpha,n}^\delta\| \leq c \left( \|\widehat{x} - x_\alpha^\delta\| + \frac{\eta_n^\delta}{\alpha} + \frac{\varepsilon_n}{\alpha} \right).$$

In particular, if  $\alpha := \alpha(\delta, n)$  is chosen in such a way that

$$\alpha(\delta, n) \rightarrow 0, \quad \frac{\delta}{\sqrt{\alpha(\delta, n)}} \rightarrow 0, \quad \frac{\varepsilon_n}{\alpha(\delta, n)} \rightarrow 0 \quad \text{and} \quad \frac{\eta_n^\delta}{\alpha(\delta, n)} \rightarrow 0$$

as  $\delta \rightarrow 0$  and  $n \rightarrow \infty$ , then

$$\|\widehat{x} - x_{\alpha,n}^\delta\| \rightarrow 0 \quad \text{as } \delta \rightarrow 0 \text{ and } n \rightarrow \infty.$$

*Proof.* Since  $A$  is non-negative and self-adjoint, it follows from spectral theory that for each  $\alpha > 0$ ,  $(A + \alpha I)^{-1}$  exists as a bounded linear operator on  $X$  and

$$\|(A + \alpha I)^{-1}\| \leq 1/\alpha.$$

Therefore, if  $\|A - A_n\| < 1/\|(A + \alpha I)^{-1}\|$  then, by results on perturbation of operators,  $(A_n + \alpha I)^{-1}$  exists and is a bounded operator, and

$$\begin{aligned} \|(A_n + \alpha I)^{-1}\| &\leq \frac{\|(A + \alpha I)^{-1}\|}{1 - \|A - A_n\| \cdot \|(A + \alpha I)^{-1}\|} \\ &\leq \frac{1/\alpha}{1 - \varepsilon_n/\alpha} \leq \frac{1}{\alpha(1 - c_0)}. \end{aligned}$$

Now let  $w_{\alpha,n}^\delta$  be the unique solution of the equation (1.9) with  $T^*y^\delta$  in place of  $z_n^\delta$ , i.e.,

$$(2.3) \quad (A_n + \alpha I)w_{\alpha,n}^\delta = T^*y^\delta.$$

Then from (1.7), (1.9) and (2.3), we have

$$x_{\alpha,n}^\delta - w_{\alpha,n}^\delta = (A_n + \alpha I)^{-1}(z_n^\delta - T^*y^\delta)$$

and

$$w_{\alpha,n}^\delta - x_\alpha^\delta = (A_n + \alpha I)^{-1}(A - A_n)x_\alpha^\delta.$$

Since  $\varepsilon_n \leq c_0\alpha$ , it follows that

$$\|x_{\alpha,n}^\delta - w_{\alpha,n}^\delta\| \leq c_1\eta_n^\delta/\alpha$$

and

$$\|w_{\alpha,n}^\delta - x_\alpha^\delta\| \leq c_2(\|\widehat{x} - x_\alpha^\delta\| + \varepsilon_n/\alpha),$$

so that

$$\|\widehat{x} - x_{\alpha,n}^\delta\| \leq c(\|\widehat{x} - x_\alpha^\delta\| + \eta_n^\delta/\alpha + \varepsilon_n/\alpha).$$

Now the assumptions on  $\alpha := \alpha(\delta, n)$  together with (1.6) and (1.8) imply the convergence  $\|\widehat{x} - x_{\alpha,n}^\delta\| \rightarrow 0$  as  $\delta \rightarrow 0$  and  $n \rightarrow \infty$ . ■

**3. The discrepancy principle.** By our assumption (2.1) on  $(\eta_n^\delta)$  and the fact that  $0 \neq y \in D(T^\dagger)$ , we have  $c_1 \leq \|z_n^\delta\| \leq c_2$  for all large enough  $n$ , say  $n \geq n_0(\delta)$  and for each  $\delta \in (0, \delta_0]$  for some  $\delta_0$ . Therefore by Theorem 2.1,

$$(3.1) \quad \|A_n x_{\alpha,n}^\delta - z_n^\delta\| = \|\alpha x_{\alpha,n}^\delta\| = \|\alpha(A_n + \alpha I)^{-1} z_n^\delta\| \leq \gamma_1$$

for some constant  $\gamma_1$  and for all  $\alpha \geq \varepsilon_n/c_0$ . Moreover, if

$$\alpha \geq \gamma_0 := \max\{\varepsilon_n/c_0 : n = 1, 2, \dots\} \quad \text{and} \quad \delta \leq \delta_0,$$

then

$$(3.2) \quad \|A_n x_{\alpha,n}^\delta - z_n^\delta\| \geq \|\alpha(A_n + \alpha I)^{-1} z_n^\delta\| \geq \frac{\gamma_0 \|z_n^\delta\|}{\|A_n\| + \alpha} \geq \gamma_2$$

for some  $\gamma_2 > 0$ , since  $(A_n)$  is uniformly bounded.

Now to choose the regularization parameter  $\alpha$  in (1.9), we consider the discrepancy principle (1.10).

For simplicity of presentation we assume that

$$(3.3) \quad \eta_n^\delta \leq c_3 \delta^r \quad \text{and} \quad \varepsilon_n \leq c_4 \delta^k$$

for some positive reals  $r$  and  $k$ , and for all  $n \geq n_0(\delta)$ .

**THEOREM 3.1.** *Let  $p$  and  $q$  be positive integers. Then for each  $\delta \in (0, \delta_0]$ , there exists a positive integer  $n_1(\delta)$  and for each  $n \geq n_1(\delta)$ , there exists  $\alpha := \alpha(\delta, n)$  such that (1.10) is satisfied. Moreover,*

$$(3.4) \quad \alpha \leq c_1 \delta^{p/(q+1)} \quad \text{and} \quad \delta^p/\alpha^q \leq c_2 \delta^\mu, \quad n \geq n_1(\delta),$$

where

$$\mu = \min \left\{ r, \frac{p}{(q+1)}, 1 + \frac{p}{2(q+1)} \right\}.$$

**Proof.** Let  $\delta \in (0, \delta_0]$ . For  $\alpha \geq \varepsilon_n/c_0$  and  $n = 1, 2, \dots$ , define

$$f_n(\alpha) = \alpha^q \|A_n x_{\alpha,n}^\delta - z_n^\delta\|.$$

Then from (3.1) it follows that  $f_n(\alpha) \leq \gamma_1 \alpha^q$  so that

$$f_n(\varepsilon_n/c_0) \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Let  $n_1(\delta) \geq n_0(\delta)$  be the smallest positive integer such that for all  $n \geq n_1(\delta)$ ,

$$\varepsilon_n \leq c_0 \min\{(\delta^p/\gamma_2)^{1/q}, (\delta^p/\gamma_1)^{1/q}\}.$$

Then taking  $\alpha_0 = \max\{\gamma_0, (\delta^p/\gamma_2)^{1/q}\}$ , we obtain  $\varepsilon_n \leq c_0 \alpha_0$  and  $\alpha_0 \geq \gamma_0$  so that by (3.1) and (3.2), we have

$$f_n(\varepsilon_n/c_0) \leq \delta^p \leq f_n(\alpha_0).$$

Therefore by the Intermediate Value Theorem, there exists  $\alpha := \alpha(\delta, n)$  such that

$$\varepsilon_n/c_0 \leq \alpha \leq \alpha_0 \quad \text{and} \quad \|A_n x_{\alpha,n}^\delta - z_n^\delta\| = \delta^p/\alpha^q$$

for all  $n \geq n_1(\delta)$ . We also note that

$$x_{\alpha,n}^\delta = \frac{1}{\alpha}(z_n^\delta - A_n x_{\alpha,n}^\delta)$$

so that for all  $n \geq n_1(\delta)$  and  $\alpha = \alpha(\delta, n)$ ,

$$\|z_n^\delta\| - \delta^p/\alpha^q = \|z_n^\delta\| - \|A_n x_{\alpha,n}^\delta - z_n^\delta\| \leq \|A_n x_{\alpha,n}^\delta\| \leq \|A_n\| \delta^p/\alpha^{q+1}.$$

Therefore  $\alpha^{q+1} \leq \delta^p(\alpha + \|A_n\|)/\|z_n^\delta\| \leq c\delta^p$  and consequently

$$\alpha(\delta, n) \leq c_1 \delta^{p/(q+1)}, \quad n \geq n_1(\delta).$$

Now, using the estimates in (1.4), (1.8) and (2.2), we have

$$\begin{aligned} \delta^p/\alpha^q &= \|A_n x_{\alpha,n}^\delta - z_n^\delta\| = \alpha \|x_{\alpha,n}^\delta\| \leq \alpha(\|\hat{x}\| + \|\hat{x} - x_{\alpha,n}^\delta\|) \\ &\leq c\alpha(\|\hat{x}\| + \|\hat{x} - x_{\alpha,n}^\delta\| + \eta_n^\delta/\alpha + \varepsilon_n/\alpha) \\ &\leq c'(\alpha + \delta\sqrt{\alpha} + \eta_n^\delta) \leq c_2 \delta^\mu, \end{aligned}$$

where  $\mu = \min\{r, p/(q+1), 1 + p/2(q+1)\}$ . ■

**4. Error estimates under the discrepancy principle.** In order to prove the convergence of  $x_{\alpha,n}^\delta$  to  $\hat{x}$  and to obtain the estimates for the error  $\|\hat{x} - x_{\alpha,n}^\delta\|$  under the discrepancy principle (1.10), we impose certain restrictions on the parameters  $p$  and  $q$  appearing in (1.10) in terms of the error levels  $\eta_n^\delta$  and  $\varepsilon_n$  of the data  $A_n$  and  $z_n^\delta$  respectively. More precisely, we assume that

$$(4.1) \quad \frac{p}{q+1} \leq \min\{2, r, k\},$$

where  $r$  and  $k$  are as in (3.3).

**THEOREM 4.1** *Let  $\alpha := \alpha(\delta, n)$  be chosen according to (1.10). Then:*

- (i)  $\|\hat{x} - x_{\alpha,n}^\delta\| \rightarrow 0$  as  $n \rightarrow \infty$  and  $\delta \rightarrow 0$ .
- (ii) If  $\hat{x} \in R(A^\nu)$ ,  $0 < \nu \leq 1$ , then for all large enough  $n$  and small enough  $\delta$ ,

$$\|\hat{x} - x_{\alpha,n}^\delta\| \leq c\delta^s,$$

where

$$s = \min \left\{ \frac{p\nu}{q+1}, 1 - \frac{p}{2(q+1)}, r - \frac{p}{q+1}, k - \frac{p}{q+1} \right\}.$$

(iii) In particular, if

$$\min\{r, k\} \geq \frac{2\nu + 2}{2\nu + 1} \quad \text{and} \quad \frac{p}{q+1} = \frac{2}{2\nu + 1},$$

then

$$\|\hat{x} - x_{\alpha,n}^\delta\| \leq c\delta^{2\nu/(2\nu+1)}.$$

Proof. Using (3.4), we have

$$\delta^l/\alpha^m = \delta^{l-mp/q}(\delta^p/\alpha^q)^{m/q} \leq c\delta^{l-m(p-\mu)/q}$$

for every  $l \geq 0$  and  $m \geq 0$ , where  $\mu$  is as in Theorem 3.1. But by the assumption (4.1),  $\mu = p/(q+1)$ , so that

$$\delta^l/\alpha^m \leq c\delta^{l-mp/(q+1)}.$$

Therefore

$$\delta/\sqrt{\alpha} \leq c_1\delta^{1-p/2(q+1)}, \quad \eta_n^\delta/\alpha \leq c_2\delta^{r-p/(q+1)} \quad \text{and} \quad \varepsilon_n/\alpha \leq c_3\delta^{k-p/(q+1)}.$$

Using this, the result in (i) follows from (1.5), (1.8) and (2.2), the estimate in (ii) follows from (1.6), (1.8) and (2.2), and (iii) is a consequence of (ii). ■

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