

Construction of admissible regularization strategies.

Definition

Regularization strategy R_α , $\alpha = \alpha(\delta)$ is called admissible if $\alpha(\delta) \rightarrow 0$ and

$$\sup \left\{ \|R_{\alpha(\delta)} y^\delta - x\|; y^\delta \in Y, \|Kx - y^\delta\| \leq \delta \right\} \rightarrow 0$$

as $\delta \rightarrow 0$.

Admissible Regularization strategies by filtering singular systems

$K: X \rightarrow Y$ linear compact operator.

with (y_j, x_j, γ_j) singular system. for K

Solution of $Kx = y$. given by Picard theorem.

$$x = \sum_{j=1}^{\infty} \frac{1}{\gamma_j} (y, y_j) x_j \quad \text{if the series converges.}$$

$y \in R(K)$

Note that if we perturb Y by $Y_\delta = Y + \delta Y_j$.

We obtain $X_\delta = X + \delta w_{j,j}^{-1} X_j$

So $\frac{\|X_\delta - X\|}{\|Y_\delta - Y\|} = \frac{1}{w_{j,j}} \rightarrow \infty$ because $w_{j,j} \rightarrow 0$

Problem is mildly ill-posed if $w_{j,j} \rightarrow 0$ slowly
and is severely ill-posed if $w_{j,j} \rightarrow 0$ fast!

Thus construct regularization strategy by
damping $\frac{1}{w_{j,j}}$

Idea: $R_\alpha Y = \sum_{j=1}^{\infty} \frac{2(\alpha_i w_{j,j})}{w_{j,j}} (Y, Y_j) X_j, Y \in Y$

with $g: (0, \infty) \times (0, \|k\|] \rightarrow \mathbb{R}$

Specially designed

Def Reg. strategy is admissible if $\alpha(\delta) \rightarrow 0$

and $\sup_{\{Y_\delta \in Y, \|kx - Y_\delta\| \leq \delta\}} \|R_{\alpha(\delta)} Y_\delta - Y\| \rightarrow 0$

for every x .

Theorem 1

$K: X \rightarrow Y$ be compact, one to one, with singular system (y_j, x_j, λ_j) and.

$$g: (0, \infty) \times (0, \|K\|] \rightarrow \mathbb{R}$$

be a function with the properties.

(1) $|g(\alpha, y)| \leq 1$ for all $\alpha > 0, 0 < y \leq \|K\|$

(2) For every $\alpha > 0$, there exists $c(\alpha)$ such that

$$|g(\alpha, y)| \leq c(\alpha)y \text{ for all } 0 < y \leq \|K\|$$

(3) $\lim_{\alpha \rightarrow \infty} g(\alpha, y) = 1$ for every $0 < y \leq \|K\|$.

Then $R_\alpha: Y \rightarrow X$ defined by.

$$R_\alpha y = \sum_{j=1}^{\infty} \frac{g(\alpha, \lambda_j)}{\lambda_j} (y, y_j) x_j, \quad y \in Y$$

is a regularisation strategy, with $\|R_\alpha\| \leq c(\alpha)$.

Choice $\alpha = \alpha(\delta)$ admissible if $\alpha(\delta) \rightarrow \infty$ and

$$\delta c(\alpha(\delta)) \rightarrow 0 \text{ as } \delta \rightarrow 0.$$

α -regularisation filter for K

Proof

$$\begin{aligned} \|R_\alpha y\|^2 &= \sum_{j=1}^{\infty} |g(\alpha, y_j)|^2 \frac{1}{y_j^2} |(y, y_j)|^2 \leq \\ &\leq C(\alpha)^2 \|y\|^2 \end{aligned}$$

So $\|R_\alpha\| \leq C(\alpha)$

$$R_\alpha kx = \sum_{j=1}^{\infty} \frac{g(\alpha, y_j)}{y_j} (kx, y_j) x_j \quad (4)$$

$$x = \sum (x, y_j) x_j \quad (5)$$

$$(kx, y_j) = y_j (x, x_j) \quad (6)$$

(6) in (4) together with (5) gives

$$\|R_\alpha kx - x\|^2 = \sum_{j=1}^{\infty} (g(\alpha, y_j) - 1)^2 |(x, x_j)|^2$$

Since $x \in \bar{X}$, for $\epsilon > 0$, $\exists N \in \mathbb{N}$ s.t.

$$\sum_{j=N+1}^{\infty} |(x, x_j)|^2 < \epsilon$$

From (3), $\exists \alpha_0 > 0$ s.t.

$$(g(\alpha, y_j) - 1)^2 < \epsilon, \quad \text{for all } j = 1, \dots, N, \quad 0 < \alpha \leq \alpha_0$$

Thus

$$\begin{aligned} \|R_\alpha E x - x\|^2 &= \sum_{j=1}^N (g(\alpha, y_j) - 1)^2 |(x, x_j)|^2 + \\ &+ \sum_{j=N+1}^{\infty} (g(\alpha, y_j) - 1)^2 |(x, x_j)|^2 < \\ &< N \epsilon \|x\|^2 + 4 \cdot \epsilon. \end{aligned}$$

like in the last estimate we used (1), as well.

Optimal regularisation strategies

Regularisation strategies that converge with the same rate as worst case error!

$$\begin{aligned} \|R_\alpha y^\delta - x\| &\leq \underbrace{\delta \|R_\alpha\|}_{\delta c(\alpha)} + \|R_\alpha E x - x\| \\ &\leq \delta c(\alpha) + \|R_\alpha E x - x\| \end{aligned}$$

If $c(\delta) \rightarrow 0$ with $\delta c(\alpha) \rightarrow 0$ then we have admissibility!

Theorem 2

Let (1), (2) hold as in the Theorem 1.

Let (3) be replaced by

(3*) There exist $c_1 > 0$ with

$$|g(\alpha, y) - 1| \leq c_1 \frac{\sqrt{\alpha}}{y} \quad \forall \alpha > 0, 0 < y \leq \|k\|$$

If $x \in R(K^*)$ then

$$\|R_\alpha Kx - x\| \leq c_1 \sqrt{\alpha} \|z\|.$$

when $x = K^* z$

if (3) replaced by $\exists c_2 > 0$

$$|g(\alpha, y) - 1| \leq c_2 \frac{\alpha}{y^2}, \quad \forall \alpha > 0, 0 < y \leq \|k\|$$

and if $x = K^* Kz$ then

$$\|R_\alpha Kx - x\| \leq c_2 \alpha \|z\|$$

Proof Idea $x = K^* z \Rightarrow (x, x_j) = y_j (z, y_j)$

$$\|R_\alpha Kx - x\|^2 = \sum_{j=1}^{o_1} (g(\alpha, y_j) - 1)^2 y_j^2 |(z, y_j)|^2 \leq c_1^2 \alpha \|z\|^2.$$

Theorems $g_i, i=1,2,3$ satisfy the assumption in Theorem 1, Theorem 2.

$$a) g_1(\alpha, y) = \frac{y^2}{\alpha + y^2}$$

$$b) g_2(\alpha, y) = 1 - (1 - \alpha y^2)^{1/2} \quad \text{for some}$$

$$\alpha \in (0,1) \quad 0 < \alpha < \frac{1}{\|k\|^2} \quad \left(\begin{array}{l} \text{Bernoulli's inequality} \\ (1+x)^q \geq 1+qx \quad \left. \begin{array}{l} q \in \mathbb{R} \\ x \geq -1 \end{array} \right\} \right)$$

$$c) g_3(\alpha, y) = \begin{cases} 1 & y^2 \geq \alpha \\ 0 & y^2 < \alpha \end{cases}$$

Theorem $\gamma_\delta \in Y, \|\gamma_\delta - \gamma\| \leq \delta$ where $\gamma = \mathbb{K}x$
 $\mathbb{K}: X \rightarrow Y$ compact, one to one, with singular system $(\frac{1}{\mu_j}, v_j, \gamma_j)$

$$R_\alpha \gamma = \sum_{\substack{j \\ \mu_j^2 \geq \alpha}} \frac{1}{\mu_j} (\gamma, \gamma_j) v_j, \quad \gamma \in Y$$

define a regularization strategy with $\|R_\alpha\| \leq \frac{1}{\sqrt{\alpha}}$
 The strategy is admissible if $\lambda(\delta) \rightarrow 0$ and $\frac{\delta^2}{\lambda(\delta)} \rightarrow 0$

If $x = \mathbb{K}^* z$ with $\|z\| \leq M$ and $c > 0$ then.

For $\alpha(\delta) = \frac{c\delta}{M}$ we have

$$\|X_{\alpha, \delta} - x\| \leq \left(\frac{1}{\sqrt{c}} + \sqrt{c}\right) \sqrt{\delta M}.$$

If $x = \mathbb{K}^* k$ with $\|k\| \leq M$, $c > 0$ for

$\alpha(\delta) = \frac{c\delta^{2/3}}{M^{2/3}}$ we have

$$\|X_{\alpha, \delta} - x\| \leq \left(\frac{1}{\sqrt{c}} + c\right) M^{1/3} \delta^{2/3}.$$

The spectral cutoff strategy is optimal for the extra approx. information $\|(\mathbb{K}^*)^{-1} x\| \leq M$ or $\|(\mathbb{K}^* k)^{-1} x\| \leq M$.

Tikhonov Regularisation

(1) $K: X \rightarrow Y$ linear and bounded, and $y \in Y$.

$\exists \hat{x} \in X$ with $\|K\hat{x} - y\| \leq \|Kx - y\|$ for all $x \in X$.

if and only if $\hat{x} \in X$ solves.

$$K^* K \hat{x} = K^* y$$

↳ normal equation!

Remark. Short parenthesis about the linear problem in $\mathbb{R}^{m \times n}$.

$$\left[\begin{array}{l} m > n, \quad A \in \mathbb{C}^{m \times n}, \quad b \in \mathbb{C}^m \\ \text{Find } x \in \mathbb{C}^n \text{ s.t. } Ax = b \end{array} \right.$$

If $m > n$, $b \notin \text{R}(A)$ in general.

Project b on $\text{R}(A)$ to minimize $\|b - Ax\|$.

One way to see the equation satisfied by projection

$$(b - A\hat{x}, Az) = 0 \quad \text{for } z \in \mathbb{C}^n.$$

$$\Rightarrow A^* b = A^* A \hat{x}.$$

Different approach

$$J(x) = \langle b - Ax, b - Ax \rangle$$

$J(\hat{x}) = \min_{x \in \mathbb{R}^n} J$, Compute differential of J .

$$\begin{aligned} J(\hat{x} + \delta y) - J(\hat{x}) &= \langle b - A\hat{x} - A\delta y, b - A\hat{x} - A\delta y \rangle - \\ &\quad - \langle b - A\hat{x}, b - A\hat{x} \rangle = \\ &= \delta^2 \langle Ay, Ay \rangle - 2\delta \langle b - A\hat{x}, Ay \rangle. \end{aligned}$$

$$\delta_x = y.$$

$$\frac{J(\hat{x} + \delta y) - J(\hat{x})}{\delta} = \delta \langle Ay, Ay \rangle - 2 \langle b - A\hat{x}, Ay \rangle$$

$$\lim_{\delta \rightarrow 0} \frac{J(\hat{x} + \delta y) - J(\hat{x})}{\delta} = 0 \implies \langle b - A\hat{x}, Ay \rangle = 0$$

$$\implies \boxed{A^* b = A^* A \hat{x}}$$

Proof of (1)

I Projection on to $R(K)$. ?

if $R(K)$ is not closed it may not exist!
Anyway, if it exists is characterized
by.

$$(y - k\hat{x}, kz) = 0 \quad \text{for all } z \in X$$

\Downarrow

$$k^*y - k^*k\hat{x} = 0.$$

Note that $R(K)$ closed $\iff K^{-1} : R(K) \rightarrow X$ bounded

If $k^*y = k^*k\hat{x}$ then from.

$$\begin{aligned} \|kx - y\|^2 &= \|k\hat{x} - y\|^2 = (kx - y, kx - y) - (k\hat{x} - y, k\hat{x} - y) \\ &= \|kx - k\hat{x}\|^2 + (k\hat{x} - y, k(x - \hat{x})) + \\ &\quad + (k(x - \hat{x}), k\hat{x} - y) \\ &= \|kx - k\hat{x}\|^2 + 2(k\hat{x} - y, k(x - \hat{x})) \\ &= \|kx - k\hat{x}\|^2 + 2(k^*k\hat{x} - k^*y, (x - \hat{x})) \end{aligned}$$

In general $K: X \rightarrow Y$ is not s.t. $R(K)$ closed.

So. $K^* K \hat{x} = K^* y$ does not have a solution!!

Penalize the defect, by considering.

x_α solution of.

$$K^* K x_\alpha + \alpha x_\alpha = K^* y$$

x_α will minimize the Tikhonov functional

$$J_\alpha(x) = \|Kx - y\|^2 + \alpha \|x\|^2 \quad \text{for } x \in X$$

Theorem! $K: X \rightarrow Y$, linear, bounded, $\alpha > 0$.

Then J_α has a unique minimum $x_\alpha \in X$, which is the unique solution of

$$\alpha x_\alpha + K^* K x_\alpha = K^* y.$$

Proof. Previous Lemma.

$$\begin{aligned} \|Kx - y\|^2 + \alpha \|x\|^2 - \|Kx_\alpha - y\|^2 - \alpha \|x_\alpha\|^2 = \\ + 2(x - x_\alpha, \alpha x_\alpha + K^*(Kx_\alpha - y)) + \\ + \|K(x - x_\alpha)\|^2 + \alpha \|x - x_\alpha\|^2 \end{aligned} \quad (*)$$

Next show that $\alpha x_\alpha + K^* K x_\alpha = K^* y$ has a unique solution!

$$T: X \rightarrow X, \quad T = \alpha I + K^*K, \quad (2)$$

$$(Tx, x) \geq \alpha \|x\|^2 \quad - \quad \underline{\text{strictly concave.}}$$

Lax - Milgram Lemma

$A: X \rightarrow X$ - linear and strictly concave.
- bounded

Then $A^{-1}: X \rightarrow X$ is bounded.

Thus $T^{-1}: X \rightarrow X$ defined in (2) is bounded.

Second solution for Theorem!

Step (*) as before. Now we show that the functional attains its infimum.

$I = \inf_X J_\alpha(x)$ exists. Let $\{x_n\}_n$ minimizing

sequence, $J_\alpha(x_n) \xrightarrow{n} I$. Show that

$\{x_n\}_n$ is Cauchy.

$$J_\alpha(x_m) + J_\alpha(x_n) = 2J_\alpha\left(\frac{1}{2}(x_n + x_m)\right) + \frac{1}{2}\|k(x_n - x_m)\|^2 +$$

$$\downarrow_{m, n \rightarrow \infty} \quad + \frac{\alpha}{2}\|x_n - x_m\|^2. \quad (3)$$

$$\geq 2I + \frac{\alpha}{2}\|x_m - x_n\|^2$$

Thus, $\|x_m - x_n\|$ can be made arbitrarily small for m, n large.

Let $x_\alpha = \lim_n x_n$, $x_\alpha \in X$.

J_α -continuous, $J_\alpha(x_n) \rightarrow J_\alpha(x_\alpha) = I$.

For (3) we used binomial formula!

$$\|a\|^2 + \|b\|^2 = 2\left\|\frac{a+b}{2}\right\|^2 + 2\left\|\frac{a-b}{2}\right\|^2$$

x_α can be written as $x_\alpha = R_\alpha y$ with

$$R_\alpha: Y \rightarrow X; \quad R_\alpha y = (\alpha I + k^*k)^{-1} k^* y$$

If (y_j, x_j, y_j) is a singular system we see

that $R_\alpha y = \sum_{j=0}^{\infty} \frac{y_j}{\alpha + y_j^2} (y_j, y_j) x_j$

$$g(\alpha, y_j) / y_j, \quad g(\alpha, y_j) = \frac{y_j^2}{\alpha + y_j^2}$$

Theorem 2, $K: X \rightarrow Y$, linear, compact, $\alpha > 0$.

i $(\alpha I + K^* K)^{-1}$, bounded.

$R_\alpha = (\alpha I + K^* K)^{-1} K^* : Y \rightarrow X$ from a regularization strategy. with $\|R_\alpha\| \leq \frac{1}{2\sqrt{\alpha}}$

$R_\alpha y_\delta$ is the unique solution of

$$\alpha x_{\alpha, \delta} + K^* K x_{\alpha, \delta} = K^* y_\delta.$$

$\alpha(\delta) \rightarrow 0$ with $\frac{\delta^2}{\alpha(\delta)} \rightarrow 0$ is admissible

b). if $x = K^* z \in \mathcal{R}(K^*)$ with $\|z\| \leq M$ then

for $\alpha(\delta) = \frac{c\delta}{M}$, $c > 0$, we have

$$\|x_{\alpha, \delta} - x\| \leq \frac{1}{2} \left(\frac{1}{\sqrt{c}} + \sqrt{c} \right) \sqrt{\delta M}$$

c) $x = K^* z$ with $\|z\| \leq M$. Then $\alpha(\delta) = c \left(\frac{\delta}{M} \right)^{2/3}$.

$c > 0$, gives.

$$\|x_{\alpha, \delta} - x\| \leq \left(\frac{1}{2\sqrt{c}} + c \right) M^{1/3} \delta^{2/3}.$$

Tikhonov regularization is optimal for $\|(K^*)^{-1} x\| \leq M$ or

$$\|(K^* K)^{-1} x\| \leq M.$$

Theorem. $k: X \rightarrow Y$ linear, compact, one-to-one
 $R(k)$ infinite dimensional.

Assume there exist $\alpha: [0, \infty) \rightarrow [0, \infty)$ $\alpha(0) = 0$ s.t.

$$\lim_{\delta} \frac{\|X_{\alpha, \delta} - x\|}{\delta^{-2/3}} = 0 \quad \text{for every } \|y_{\delta} - kx\| \leq \delta$$

where $X_{\alpha, \delta} \in X$ solution of.

$$\alpha X_{\alpha, \delta} + k^* k X_{\alpha, \delta} = k^* y_{\delta}.$$

Then $x = 0$

Proof Assume $x \neq 0$.

Show that $\alpha \delta^{-2/3} \rightarrow 0$. $y = kx$. Then.

$$(\alpha I + k^* k) (X_{\alpha, \delta} - x) = k^* (y_{\delta} - y) - \alpha x.$$

$$|\alpha| \|x\| \leq \|k\| \delta + (\alpha + \|k\|^2) \|X_{\alpha, \delta} - x\|$$

~~$$\alpha \cdot \delta^{-2/3} \|x\| \leq \|k\|^{1/3} (\alpha + \|k\|^2).$$~~

$$\alpha \cdot (\|x\| - \|X_{\alpha, \delta} - x\|) \leq \|k\| \delta + \|k\|^2 \|X_{\alpha, \delta} - x\|$$

$$\alpha \cdot \delta^{-2/3} \rightarrow 0$$

(y_j, x_j, γ_j) singular system for k .

$$\delta_j = y_j^3, \quad \gamma_{\delta_j} = \gamma + \delta_j \gamma_j, \quad j \in \mathbb{N}.$$

$$\alpha_j = \alpha(\delta_j)$$

$$\begin{aligned} X_{\alpha_j, \delta_j} - x &= (X_{\alpha_j, \delta_j} - X_{\alpha_j}) + (X_{\alpha_j} - x) = \\ &= (\alpha_j I + k^* k)^{-1} k^* (\delta_j \gamma_j) + (X_{\alpha_j} - x) \\ &= \frac{\delta_j y_j}{\alpha_j + y_j^2} x_j + X_{\alpha_j} - x \end{aligned}$$

Since $\|X_{\alpha_j} - x\| \delta_j^{-2/3} \rightarrow 0$

We have.

$$\frac{\delta_j^{1/3} y_j}{\alpha_j + y_j^2} \rightarrow 0 \quad j \rightarrow \infty.$$

But.

$$\frac{\delta_j^{1/3} y_j}{\alpha_j + y_j^2} = \frac{y_j^2}{\alpha_j + y_j^2} = (1 + \alpha_j \delta_j^{-2/3})^{-1} \rightarrow 1$$

Contr.

Discrepancy Principle of Morozov

A posteriori choice of regularization parameter α !

$K: X \rightarrow Y$ - compact and injective

$$\overline{K(X)} = Y$$

$Kx = y$ for $y \in Y$. (May not have a sol. if $y \notin R(K)$).

with Tikhonov regularization.

Let $x_\alpha = (\alpha I + K^*K)^{-1} K^* y$ be the unique solution

$$\text{of } K^* K x_\alpha + \alpha x_\alpha = K^* y$$

1. x_α depends continuously on α .

2. $\alpha \rightarrow \|x_\alpha\|$ monotonously nonincreasing

3. $\alpha \rightarrow \|Kx_\alpha - y\|$ monotonously nondecreasing.

$$\text{and } \lim_{\alpha \rightarrow 0} Kx_\alpha = y.$$

If $y \notin N(K^*)$ then strict monotonicity is observed in 2 and 3.

proof

Remember x_α is minima of.

$$J_\alpha(x) = \|kx - y\|^2 + \alpha \|x\|^2 \quad \text{on } X$$

Thus.

$$\alpha \|x_\alpha\| \leq J_\alpha(x_\alpha) \leq J_\alpha(0) = \|y\|^2.$$

\Downarrow

$$\|x_\alpha\| \leq \frac{\|y\|}{\sqrt{\alpha}} \Rightarrow x_\alpha \rightarrow 0 \quad \text{as } \alpha \rightarrow \infty$$

let $\alpha, \beta > 0$, let $\varepsilon > 0$.

$$\alpha(x_\alpha - x_\beta) + k^*(k(x_\alpha - x_\beta) + (\alpha - \beta)x_\beta) = 0$$

$$(1) \quad \alpha \|x_\alpha - x_\beta\|^2 + \|k(x_\alpha - x_\beta)\|^2 = (\beta - \alpha)(x_\beta, x_\alpha - x_\beta).$$

\Downarrow

$$\alpha \|x_\alpha - x_\beta\|^2 \leq |\beta - \alpha| \|x_\beta\| \|x_\alpha - x_\beta\|$$

$$\alpha \|x_\beta - x_\alpha\| \leq |\beta - \alpha| \frac{\|y\|}{\sqrt{\beta}} < \varepsilon \frac{\|y\|}{\sqrt{\beta}}$$

choose $\delta = \frac{\varepsilon \|y\|}{\alpha \cdot \sqrt{\beta}}$ and see continuity.

Then let $\beta > \alpha > 0$. Then by using (1)

we obtain

$$(x_\beta, x_\alpha - x_\beta) \geq 0. \quad \text{So, we have.}$$

$$\|x_\beta\|^2 \leq (x_\beta, x_\alpha) \leq \|x_\beta\| \|x_\alpha\|$$

\Downarrow

$$\|x_\beta\| \leq \|x_\alpha\|$$

$\hookrightarrow \alpha - \|x_\alpha\|$ - monotone
nonincreasing

Next;

multiply

$$\beta x_\beta + k^* k x_\beta = k^* y$$

by $(x_\alpha - x_\beta)$ to obtain.

$$\beta (x_\beta, x_\alpha - x_\beta) + (k x_\beta - y, k(x_\alpha - x_\beta)) = 0 \quad (2)$$

Let $\beta > \alpha$. From (1) as before.

$$(x_\beta, x_\alpha - x_\beta) \geq 0$$

From this and (2) we have.

$$(k x_\beta - y, k(x_\alpha - x_\beta)) \leq 0 \Leftrightarrow$$

$$(k x_\beta - y, k x_\alpha - y) \leq \|k x_\beta - y\|^2$$

$$\|Kx_\alpha - y\| \leq \|Kx_\beta - y\|$$

$\alpha \mapsto \|Kx_\alpha - y\|$ - monotonous
nondecreasing.

Next, let $\varepsilon > 0$. Since $\overline{R(K)} = Y$.

$$\exists x \in X, \|Kx - y\|^2 \leq \frac{\varepsilon^2}{2}.$$

$$\|Kx_\alpha - y\|^2 \leq J_\alpha(x_\alpha) \leq J_\alpha(x) \leq \varepsilon^2$$

$$\text{for } \alpha \leq \frac{\varepsilon^2}{2\|x\|^2}$$

$$\text{Thus } \|Kx_\alpha - y\| \leq \varepsilon \quad \text{for } \alpha \leq \frac{\varepsilon^2}{2\|x\|^2}$$

$$\text{So } \lim_{\alpha \rightarrow \infty} Kx_\alpha = y$$

Moore's discrepancy principle

Assume $Kx = y$ has a solution

$K: X \rightarrow Y$ - linear, compact, one-to-one
with dense range! Let $\delta = \delta(\varepsilon)$ be such that

$x_{\delta, \delta}$ solution of

$$\alpha x_{\delta, \delta} + K^* K x_{\delta, \delta} = K^* y_\delta \quad (3)$$

Satisfies $\|kx_{\alpha, \delta} - y_{\delta}\| = \delta. \quad (4)$

Note that (4) is uniquely solvable if

$$\|y_{\delta} - y\| \leq \delta < \|y_{\delta}\| \quad \text{since.}$$

$$\lim_{\alpha \rightarrow \infty} \|kx_{\alpha, \delta} - y_{\delta}\| = \|y_{\delta}\| > \delta.$$

$$\lim_{\alpha \rightarrow \infty} \|kx_{\alpha, \delta} - y_{\delta}\| = 0 < \delta.$$

+ continuity and increasing character of.

$$\alpha \longrightarrow \|kx_{\alpha, \delta} - y_{\delta}\|$$

Theorem. Assume all of the above, i.e., (3), (4). Then

I. $x_{\alpha, \delta} \rightarrow x$ for $\delta \rightarrow 0$

↳ The discrepancy principle is admissible.

II. if $x = k^*z$, $\|z\| \leq M$ then.

$$\|x_{\alpha, \delta} - x\| \leq 2\sqrt{\delta} \cdot \sqrt{M}.$$

When here we assumed $kx = y$ has a solution!

Proof

Recall that $x_{\alpha, \delta}$ minimizes.

$$J(x) = \alpha(\delta) \|x\|^2 + \|kx - y_\delta\|^2$$

So, we have

$$\begin{aligned} \alpha(\delta) \|x_{\alpha, \delta}\|^2 + \delta^2 &= J(x_{\alpha, \delta}) \leq J_\delta(x) = \alpha(\delta) \|x\|^2 + \|y - y_\delta\|^2 \\ &\leq \alpha(\delta) \|x\|^2 + \delta^2 \end{aligned}$$

Thus

$$\|x_{\alpha, \delta}\| \leq \|x\|$$

This implies.

$$\begin{aligned} \|x_{\alpha, \delta} - x\|^2 &= \|x_{\alpha, \delta}\|^2 - 2(x_{\alpha, \delta}, x) + \|x\|^2 \leq \\ &\leq 2(\|x\|^2 - (x_{\alpha, \delta}, x)) = 2(x - x_{\alpha, \delta}, x) \end{aligned}$$

If $x = k^* z$ then.

$$\begin{aligned} \|x_{\alpha, \delta} - x\|^2 &\leq 2(x - x_{\alpha, \delta}, k^* z) = 2(y - kx_{\alpha, \delta}, z) \\ &= 2(y - y_\delta, z) + 2(y_\delta - kx_{\alpha, \delta}, z) \\ &\leq 2\delta \|z\| + 2\delta \|z\| \leq 4\delta M. \end{aligned}$$

↳ Thus II follows.

Observe that since k is one to one

$$N(k) = 0 \Rightarrow \overline{R(k^*)} = X.$$

Thus there exists $\tilde{X} = k^* z$ with

$$\|\tilde{X} - X\| \leq \varepsilon.$$

$$\begin{aligned} \|X_{\alpha, \delta} - X\|^2 &\leq 2(X - X_{\alpha, \delta}, X) = \\ &= 2(X - X_{\alpha, \delta}, X - \tilde{X}) + 2(k^* z, X - X_{\alpha, \delta}) \\ &\leq 2\|X - X_{\alpha, \delta}\| \cdot \varepsilon + 4\delta\|z\|. \end{aligned}$$

$$\left(\|X_{\alpha, \delta} - X\| - \varepsilon\right)^2 \leq \varepsilon^2 + 4\delta\|z\|$$

Choose $\delta \leq \frac{\varepsilon^2}{4\|z\|}$

Computation of $\alpha(\delta)$ such that

$$\|k X_{\alpha, \delta} - Y_\delta\| = \delta. \quad \text{amounts to}$$

applying Newton's method for finding roots of

$$\|k X_{\alpha, \delta} - Y_\delta\|^2 - \delta^2 = 0.$$