

1.4 Choice of regularization parameter

$$T \in \mathcal{L}(X, Y)$$

$\{R_\alpha\}_{\alpha \in \mathcal{A}} \subset \mathcal{L}(Y, X)$ induced by filter $\{q_\alpha\}_{\alpha \in \mathcal{A}}$

Ill-posed operator equation $Tf = g$, solution $f^* \in X$, $Tf^* = g$

[$\Rightarrow g \in \mathcal{R}(T)$: exact data attainable]

Recall: if $f^* \in K_{\nu, \beta} := \{f \in X : \|f\|_V \leq \beta\}$ ("source condition")

$\|f^* - R_{\alpha^*} g^\delta\|_X \leq C_V \delta^{\frac{1}{2\nu}} \beta^{\frac{1}{2\nu}}$ if $\{q_\alpha\}$ has qualification $\nu_0 \geq \nu$
and $\alpha^* = \alpha^*(\beta, \delta, \nu) = \left(\frac{C_V \delta}{\nu C_V \beta}\right)^{2/\nu}$

[order optimal a priori parameter choice rule]

Problem: β not known, α^* is bad if β is underestimated

Desired: P.C.R.s that do not rely on knowledge of β and ν

1.4.1 The discrepancy principle

\rightarrow gives $\bar{\alpha} = \bar{\alpha}(\delta, g^\delta)$ a posteriori P.C.R

Def 1.4.1.A (Morozov discrepancy principle)

Pick $\tau > C_T := \sup \{|r_\alpha(\lambda)| : \alpha \in \mathcal{A}, 0 \leq \lambda \leq \|T\|^2\} \geq 1$

(i) If $\frac{\|g^\delta\|_Y}{\delta} \geq \tau$, pick a safety factor $\eta > 1$
and fix $\alpha_d = \alpha(\delta, g^\delta)$ according to

$$\alpha_d = \sup \left\{ \alpha \in \mathcal{A} : \|g^\delta - TR_\alpha g^\delta\|_Y \leq \tau \delta \right\}$$

residual: computable!

(ii) If $\frac{\|g^\delta\|_Y}{\delta} < \tau$ ("useless data")

set " $\alpha_d = \infty$ " : $R_{\alpha_d} := 0$

Is α_d well-defined in case (i)?

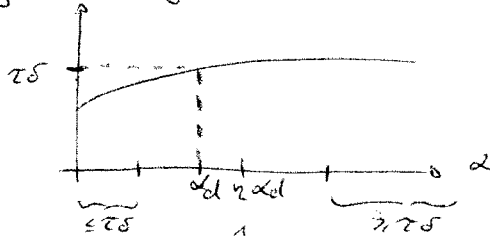
$$\|g^\delta - TR_\alpha g^\delta\|_Y^2 = \sum_j |r_\alpha(\sigma_j)|^2 |(g, g_j)_Y|^2$$

$$\lim_{\alpha \rightarrow \infty} \|g^\delta - TR_\alpha g^\delta\|_Y^2 = \sum_{j: \sigma_j \neq 0} |(g^\delta, g_j)_Y|^2 = \|P_{\overline{\text{ran}(T)}} g^\delta\|_Y^2$$

$$= \|P_{\overline{\text{ran}(T)}} (g - g^\delta)\|_Y^2 \leq \|g - g^\delta\|_Y^2 \leq \delta^2$$

$$\|R_\alpha\| \leq \frac{C_1}{\sqrt{\alpha}} \Rightarrow \lim_{\alpha \rightarrow \infty} \|g^\delta - TR_\alpha g^\delta\|_Y = \|g^\delta\|_Y \approx \tau \delta$$

$$\|g^\delta - TR_\alpha g^\delta\|_Y$$



we can always find an α_δ here

For iterative regularization

$\hat{f}_0 := 0 \rightarrow$ not enough

take first $\hat{f}_k : \|g^\delta - T\hat{f}_k\|_Y \leq \tau \delta$

Regularization error / approximation error

$$\|f^+ - R_{\alpha_\delta} T f^+\|_X = \sum |r_{\alpha_\delta}(\sigma_j^2)|^2 |(f^+, f_j)_X|^2$$

Tool: Hölder inequality for sequences

$$p = v+1, \quad q = \frac{v+1}{v}, \quad \Rightarrow \frac{1}{p} + \frac{1}{q} = 1$$

$$= \sum \left\{ |r_{\alpha}(\sigma_j^2)|^{2/p} \sigma_j^{-2/q} |(f^+, f_j)_X|^{2/p} \right\} \left\{ |r_{\alpha}(\sigma_j^2)|^{2/q} \sigma_j^{2/q} |(f^+, f_j)_X|^{2/q} \right\}$$

$$\leq \left(\sum_j \underbrace{|r_{\alpha}(\sigma_j^2)|^2}_{\leq C_r^2} \sigma_j^{-2v} |(f^+, f_j)_X|^2 \right)^{1/p} \left(\sum_j |r_{\alpha}(\sigma_j^2)|^2 \sigma_j^2 |(f^+, f_j)_X|^2 \right)^{1/q}$$

$$\leq (C_r^2 \|f^+\|_X^2)^{1/p} (\|g - TR_\alpha g\|_Y^2)^{1/q}$$

$$\|g - TR_\alpha g\|_Y^2 = \sum_j |r_{\alpha}(\sigma_j^2)|^2 |(g, g_j)_Y|^2 \stackrel{T^* g_j = \sigma_j f_j}{=} \sum_j \sigma_j^2 |r_{\alpha}(\sigma_j^2)|^2 |(f^+, f_j)_X|^2$$

$$\|g - TR_{\alpha_\delta} g\|_Y \leq \underbrace{\|(\text{Id} - TR_{\alpha_\delta})(g - g^\delta)\|_Y}_{\leq C_r \|g - g^\delta\|_Y} + \underbrace{\|g^\delta - TR_{\alpha_\delta} g^\delta\|_Y}_{\leq \tau \delta}$$

$$\|f^+ - R_{\alpha_\delta} T f^+\|_X \leq \underbrace{C_r^{1/v+1}}_{C_r} \underbrace{\delta^{1/v}}_{\delta^{1/v}} (C_r \delta + \tau \delta)^{v/(v+1)} = C_r^{1/v+1} \delta^{v/(v+1)} (C_r \delta + \tau \delta)^{v/(v+1)}$$

Data noise error

$$\bullet \quad \|R_{\alpha d}(g - g^\delta)\|_X \leq \frac{C_g}{\sqrt{\alpha d}} \delta$$

["Show that α is big enough"]

$$\text{Goal: } \frac{\delta}{\sqrt{\alpha d}} \leq C_g \delta^{1/\nu+1}$$

$$\Leftrightarrow \alpha \geq C_g^2 \delta^{2/\nu+1} \quad (*)$$

Known from discrepancy principle: $\|g^\delta - TR_{\alpha d} g^\delta\| \geq \tau \delta$

$$\Rightarrow \tau \delta \leq \|(\text{Id} - TR_{\alpha d})g\|_Y + \underbrace{\|(\text{Id} - TR_{\alpha d})(g - g^\delta)\|_Y}_{\leq C_r \delta}$$

$$g = Tf^+$$

$$\bullet \quad \begin{aligned} \|(\text{Id} - TR_{\alpha d})g\|_Y^2 &= \sum_j \sigma_j^2 |r_{\alpha d}(\sigma_j^2)|^2 |(f^+, f_j)_X|^2 \\ &\leq \sum_j \sigma_j^{2+2\nu} |r_{\alpha d}(\sigma_j^2)|^2 \\ &\leq \left(\sup_j \sigma_j^{2+2\nu} |r_{\alpha d}(\sigma_j^2)|^2 \right) \|f^+\|_V^2 \end{aligned}$$

Assumption: $\{q_\alpha\}$ has qualification $\geq \nu+1$

$$\Rightarrow \lambda^{\nu+1} |r_\alpha(\lambda)|^2 \leq C_\nu^2 \alpha^{1+\nu} \quad \forall \alpha, 0 \leq \lambda \leq \|T\|^2$$

$$\leq C_\nu^2 (\eta_{\alpha d})^{1+\nu} \|f^+\|_V^2$$

$$\Rightarrow \tau \delta \leq C_\nu^2 (\eta_{\alpha d})^{\frac{1+\nu}{2}} \delta + C_r \delta$$

$$\bullet \quad [\tau > C_r]$$

$$\Rightarrow (\tau - C_r) \delta \leq C_\nu^2 \delta^{\frac{1+\nu}{2}}$$

$$\Rightarrow \alpha_d \geq C_\nu^2 \delta^{2/\nu+1} \quad \Leftrightarrow (*)$$