

we have

$$\Theta^{**} = \Theta$$

$$\Theta(t) = \sup_b (bt - \Theta^*(b)) = \Theta^{**}(t)$$

calculate  $\Theta^{**}$  at  $t^2$ :

$$\begin{aligned} -\varphi(t) = \Theta(t^2) &= \sup_{b \leq 0} (bt^2 - \Theta^*(b)) \\ &\quad \uparrow \text{replace by } -\frac{1}{2}b \\ &\quad \left( \uparrow \text{if } b > 0 \quad \Theta^*(b) = +\infty \right) \\ &= \sup_{b \geq 0} \left( -\frac{1}{2}bt^2 - \psi(b) \right) \end{aligned}$$

$$\varphi(t) = \inf_{b \geq 0} \left( \frac{1}{2}bt^2 + \psi(b) \right)$$

$$\text{where } \psi(b) = \sup_{t \geq 0} \left( -\frac{1}{2}bt^2 + \varphi(t) \right)$$

$$\begin{aligned} \inf_{b \in \mathbb{R}^2} F(u, b) &= \|Au - v\|^2 + \beta \sum_{i=1}^2 \inf_{b[i]} \left( \frac{b[i]}{2} \|D_i u\|^2 + \psi(b[i]) \right) \\ &= F(u) \qquad \qquad \qquad \varphi(\|D_i u\|) \end{aligned}$$

$$\hat{u}, \hat{b} \Rightarrow \hat{u} \text{ minimizes } F(u)$$

We will compute the value  $\hat{b} \in \mathbb{R}_+$  such that  
 $\varphi(t) = \frac{1}{2} \hat{b} t^2 + \psi(\hat{b})$  (i.e. for which the inf is reached)

$$\hat{b} = \hat{b}(t)$$

Assumptions

$$\varphi \text{ is } C^1 \text{ on } \mathbb{R}_+, \quad \varphi'(0) = 0$$

$$\lim_{t \rightarrow \infty} \varphi(t)/t^2 = 0 \quad \left( \text{classical assumption for edge-preserving regu} \right)$$

$$G(t) = \begin{cases} \frac{\varphi'(t)}{t} & t > 0 \\ \varphi''(0^+) & t = 0 \end{cases}$$

$$F(u, \hat{b}) = F(u) \quad \text{if} \quad \hat{b}[i] = G(\|D_i u\|)$$

F in matrix form

$$F(u, b) = \|Au - v\|^2 + \frac{\beta}{2} (Du)^T B Du + \beta \sum \psi(b[i])$$

where  $B = \text{diag}(b[1] \mathbb{I}_s, \dots, b[r] \mathbb{I}_s)$

s = the dim of  $D_i u$  (e.g. s = 1 or 2)

H

$$\begin{matrix} & b[1] & & & & & \\ & & b[1] & & & & \\ & & & b[2] & & & \\ & & & & b[2] & & \\ & & & & & & \dots \end{matrix} \quad \left[ \begin{matrix} b(1) \\ b(2) \\ \dots \end{matrix} \right]$$

$$H(b) = 2A^T A + \beta D^T B D \quad \text{OR} \quad \text{invertible if } \text{rank} A = p$$

for  $b$  fixed the min of  $F(u, b)$  is reached at  $\hat{u}$  satisfying  $\nabla_u F(\hat{u}, b) = 0$  and hence

$$H(b) \hat{u} = 2A^T v$$

$$\nabla_u F = 2A^T A u - 2A^T v + \beta D^T B D u = 0$$

Therefore the alternate min scheme is

$$\begin{cases} u_k = (H(b_{k-1}))^{-1} 2A^T v \\ b_k[i] = G(\|D_i u_k\|) \quad 1 \leq i \leq r \end{cases}$$

$\hat{b} > 0$  define  $f_{\hat{b}}: \mathbb{R}_+ \rightarrow \mathbb{R}$

$$f_{\hat{b}}(t) = \frac{1}{2} \hat{b} t^2 + \theta(t)$$

then  $\psi(\hat{b}) = - \inf_{t \geq 0} f_{\hat{b}}(t)$

$f_{\hat{b}}$  convex,  $f_{\hat{b}}(0) = 0$ ,  $f_{\hat{b}}(t) \xrightarrow{t \rightarrow \infty} +\infty$

$\Rightarrow f_{\hat{b}}$  has a unique min on  $[0, +\infty)$

let  $\hat{t} > 0$  denote the point where it is reached.

$$\psi(\hat{b}) = -\frac{1}{2} \hat{b} \hat{t}^2 + \varphi(\hat{t}) = -\inf_t f_{\hat{b}}(t)$$

$$\Leftrightarrow \varphi(\hat{t}) = \frac{1}{2} \hat{b} \hat{t}^2 + \psi(\hat{b}) = \inf(\dots)$$

$$\theta'(t) = -\frac{\varphi'(\sqrt{t})}{2\sqrt{t}} \text{ and } \lim_{t \rightarrow 0} \theta'(t) = -\frac{1}{2} \varphi''(0^+)$$

$$f'_{\hat{b}}(t) = \frac{1}{2} \hat{b} + \theta'(t) \text{ increasing because } f_{\hat{b}} \text{ convex}$$

we are looking for a solution <sup>constrained</sup> on  $\mathbb{R}_+$

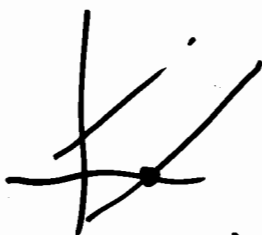
If  $f'_{\hat{b}}(0^+) \geq 0$  then the min is at  $\hat{t} = 0$

$$\Leftrightarrow \hat{b} \geq \varphi''(0^+)$$

otherwise  $\hat{t} > 0$  satisfies

$$f'_{\hat{b}}(\hat{t}) = 0 \Leftrightarrow \hat{b} = -2\theta'(\hat{t})$$

back to  $\varphi \dots \hat{b} = -2\theta'(\hat{t}^2)$



$$H(u) = \mathcal{H} \left( \sum_{i=1}^n \beta (\|D_i u\|)^2 \right) \quad \text{Newton: } \varphi''(\|D_i u\|)$$

$$= 2A^T A + \beta \sum_{i=1}^n \frac{\varphi'(\|D_i u\|)}{\|D_i u\|} D_i^T D_i$$

The method amounts to

$$u_{k+1} = (H(u_{k-1}))^{-1} 2A^T v = \dots$$

$$= u_{k-1} - \underbrace{(H(u_{k-1}))^{-1}} \nabla F(u_{k-1})$$

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(b) Additive form of HQ regularization.

assumptions /  $\varphi$  is convex, continuous

$$\left\{ \begin{array}{l} \frac{1}{2} t^2 - \varphi(t) \text{ is convex } (\Leftrightarrow \varphi'(0)=0) \\ \lim_{t \rightarrow \infty} \frac{\varphi(t)}{t^2} < \frac{1}{2} \end{array} \right.$$

$S = 1$  or  $2$  (each  $D_i$  matrix  $S \times P$ )  $\leftarrow \text{dim of } u$

$b \in \mathbb{R}^S$  below,  $z \in \mathbb{R}^S$

$$\psi(b) = \sup_z \left\{ -\frac{1}{2} \|z - b\|^2 + \varphi(\|z\|) \right\}$$

$$f^* = \sup_z \left\{ -\frac{1}{2} \|z\|^2 - \frac{1}{2} \|b\|^2 + \langle z, b \rangle + \varphi(\|z\|) \right\}$$

$$\psi(b) + \frac{1}{2} \|b\|^2 = \sup_z \left\{ \langle z, b \rangle - \underbrace{\left( \frac{1}{2} \|z\|^2 - \varphi(\|z\|) \right)}_{\text{convex continuous}} \right\}$$

$$\left\{ \begin{array}{l} f^*(x) = \sup_u (\langle u, x \rangle - f(u)) \quad \text{and } f^{**}(u) = f(u) \\ f^{**}(u) = \sup_x (\langle u, x \rangle - f^*(x)) = f(u) \end{array} \right.$$

$$\frac{1}{2} \|z\|^2 - \varphi(\|z\|) = \sup_b \left\{ \langle z, b \rangle - \left( \varphi(b) + \frac{1}{2} \|b\|^2 \right) \right\}$$

$$-\varphi(\|z\|) = \sup_b \left\{ \underbrace{\langle z, b \rangle - \frac{1}{2} \|z\|^2 - \frac{1}{2} \|b\|^2}_{-\|z-b\|^2} - \varphi(b) \right\}$$

$$\varphi(\|z\|) = \inf_b \left\{ \frac{1}{2} \|z-b\|^2 + \varphi(b) \right\}$$

we can compute  $\hat{b}$  for which the inf is reached.

$(\hat{z}, \hat{b})$  are such that the inf in the def. of  $\varphi$  is reached  $(\Leftrightarrow)$  the sup in the def. of  $\Psi$  is reached

$$\varphi(\|\hat{z}\|) = \frac{1}{2} \|\hat{z} - \hat{b}\|^2 + \varphi(\hat{b})$$

$$\Leftrightarrow \varphi(\hat{b}) = -\frac{1}{2} \|\hat{z} - \hat{b}\|^2 + \varphi(\|\hat{z}\|)$$

$\Rightarrow$  the grad with respect to  $\hat{z}$  of the function  $z \rightarrow -\frac{1}{2} \|z - \hat{b}\|^2 + \varphi(\|z\|)$  is null.

$$-(\hat{z} - \hat{b}) + \varphi'(\|\hat{z}\|) \frac{\hat{z}}{\|\hat{z}\|} = 0$$

$$\hat{b} = \hat{z} - \varphi'(\|\hat{z}\|) \frac{\hat{z}}{\|\hat{z}\|}$$

$$F(w) = \|Au - v\|^2 + \beta \sum \varphi(\|D_i u\|) \quad w \in \mathbb{R}^S$$

$$F(u, b) = \|Au - v\|^2 + \beta \sum \left( \frac{1}{2} \|D_i u - b_i\|^2 + \varphi(b_i) \right)$$

$$\inf_b F(u, b) = F(u) \quad \inf_{b_i} (\cdot) = \varphi(\|D_i u\|)$$

For  $u$  fixed

for  $u$  fixed  $\inf_b F(u, b)$  is given reached at

$$\hat{b}_i = D_i u - \varphi'(\|D_i u\|) \frac{D_i u}{\|D_i u\|} \quad 1 \leq i \leq r$$

for  $b$  fixed  $\inf_u F(u, b)$  is reached at  $\hat{u}$

such that  $\nabla_u F(u, b) = 0$

$$D = \begin{bmatrix} D_1 \\ \vdots \\ D_r \end{bmatrix} \quad b = \begin{bmatrix} b_1 \\ \vdots \\ b_r \end{bmatrix}$$

$$F(u, b) = \|Au - \sigma\|^2 + \beta \frac{1}{2} \|Du - b\|^2 + \beta \sum_i \varphi(b_i)$$

$$2A^T(Au - \sigma) + \beta D^T(Du - b) = 0$$

$$\text{or } (2A^T A + \beta D^T D)u = 2A^T \sigma + \beta D^T b$$

Let

$H$

$$\hat{u} = H^{-1} (2A^T \sigma + \beta D^T b)$$

The alternate minimization:

$$\forall i \quad b_i^k \stackrel{\text{iteration}}{=} D_i u^{k-1} - \varphi'(\|D_i u^{k-1}\|) \frac{D_i u^{k-1}}{\|D_i u^{k-1}\|}$$

$$u^k = H^{-1} (2A^T \sigma + \beta D^T b^k)$$

Quasi-Newton  $u^k = u^{k-1} - \underbrace{(H^{-1})}_{\text{approx of } \nabla^2 F(u^{k-1})} \nabla F(u^{k-1})$

$$u^k = u^{k-1} - (H^{-1}) \nabla F(u^{k-1})$$

Corresp. to  $\varphi'' = 1$  all the time

$$2A^T A + \beta D^T \underbrace{\text{diag}(\varphi''(\cdot))}_{\perp} D$$

Both forms (\* and +) convergence established for  $\varphi$  convex smooth, can be interpreted as in terms of ~~the~~ majorizing quadratic func. (involve the inversion of a matrix)

Convergence speed

\* needs less iterations than +  
 cost per iter. for \*  $>$  cost per iter. +  
 Overall + is faster.

both are faster than classical methods.

Matrix  $H(u_k)$ -inversion.

Truncated versions of HQ regu:

the idea: compute  $H(u_k)^{-1}$  only approximately (several iter.)

~~Full~~ ~~NA~~

still convergence proven.