

# Geometry of optimal decomposition for the L- functional and duality in convex analysis

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# ROF model

## ROF model

In 1992 Rudin, Osher and Fatemi suggested a denoising model which has made great success. Let  $D = [a, b] \times [c, d]$  be a rectangular domain in  $\mathbb{R}^2$ . Suppose that initial image  $f_* \in BV$  and we observe

$$f_{\text{ob}} = f_* + \eta,$$

where  $\eta \in L^2(D)$  corresponds to noise. In order to reconstruct approximately initial image  $f_*$ , ROF suggested to consider

$$J_{2,1}(t, f_{\text{ob}}, L^2(D), BV(D)) = \inf_{g \in BV} \left( \frac{1}{2} \|f_{\text{ob}} - g\|_{L^2}^2 + t \|g\|_{BV} \right),$$

and to take as approximation to  $f_*$  the function  $f_t$  which minimizes this functional, i.e.,

## ROF model

$$L_{2,1}(t, f_{\text{ob}}, L^2(D), BV(D)) = \frac{1}{2} \|f_{\text{ob}} - f_t\|_{L^2}^2 + t \|f_t\|_{BV},$$

where (for a function  $f$  of class  $C^1$ )

$$\|f\|_{BV} = \iint_D \left( \left| \frac{\partial f}{\partial x}(x, y) \right| + \left| \frac{\partial f}{\partial y}(x, y) \right| \right) dx dy.$$

## Optimal decomposition

The expression

$$f_{\text{ob}} = (f_{\text{ob}} - f_t) + f_t,$$

is called *optimal decomposition* of  $L_{2,1}(t, f_{\text{ob}}, L^2(D), BV(D))$  corresponding to  $f_{\text{ob}}$ .

# ROF model

In 2002, Yves Meyer obtained a mathematical characterization of this optimal decomposition for this couple  $(L^2(D), BV(D))$  by using duality. [Yves Meyer, *Oscillating Patterns in Image Processing and Nonlinear Evolution Equations*, 2002]

# The problem

Let  $(X_0, X_1)$  be a compatible Banach couple. i.e.,  $X_0$  and  $X_1$  are Banach spaces such that  $X_0$  and  $X_1$  are linearly and continuously embedded in some Banach space  $\mathcal{X}$ . Let  $x \in X_0 + X_1$ , let  $1 < p < +\infty$  and  $t > 0$ . We consider the L- functional

$$L_{p,1}(t, x; X_0, X_1) = \inf_{x=x_0+x_1} \left( \frac{1}{p} \|x_0\|_{X_0}^p + t \|x_1\|_{X_1} \right),$$

We give a characterization of *optimal decomposition* for the L- functional. i.e.,  $x = x_{0,\text{opt}} + x_{1,\text{opt}}$  such that

$$L_{p,1}(t, x; X_0, X_1) = \frac{1}{p} \|x_{0,\text{opt}}\|_{X_0}^p + t \|x_{1,\text{opt}}\|_{X_1}$$

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For ROF model,

$$p = 2, X_0 = L^2(D), X_1 = BV(D)$$

## Dual characterization of optimal decomposition

Let  $(X_0, X_1)$  be a regular couple ( $X_0 \cap X_1$  is dense in both  $X_0$  and  $X_1$ ). Then it is a known fact from interpolation theory that  $(X_0^*, X_1^*)$  also form a Banach couple and  $(X_0 \cap X_1)^* = X_0^* + X_1^*$ . The dual spaces are defined by the norm

$$\|y\|_{X_j^*} = \sup \left\{ \langle y, x \rangle : x \in X_j, \|x\|_{X_j} \leq 1 \right\}, \quad j = 0, 1.$$

The spaces  $X_0 + X_1$  and  $X_0 \cap X_1$  are Banach spaces with respect to the following norms

$$\|x\|_{X_0 + X_1} = \inf_{x = x_0 + x_1} \left\{ \|x_0\|_{X_0} + \|x_1\|_{X_1} \right\},$$

where the infimum extends over all representations  $x = x_0 + x_1$  of  $x$  with  $x_0$  in  $X_0$  and  $x_1$  in  $X_1$ , and

$$\|x\|_{X_0 \cap X_1} = \max \left\{ \|x\|_{X_0}, \|x\|_{X_1} \right\}.$$

# Dual characterization of optimal decomposition

## Theorem (Main Theorem)

Let  $1 < p < +\infty$ . The decomposition  $x = x_{0,opt} + x_{1,opt}$  is optimal for  $L_{p,1}(t, x; X_0, X_1)$  if and only if  $\exists y_* \in X_0^* \cap X_1^*$  such that  $\|y_*\|_{X_1^*} \leq t$  and

$$\begin{cases} \frac{1}{p} \|x_{0,opt}\|_{X_0}^p = \langle y_*, x_{0,opt} \rangle - \frac{1}{p'} \|y_*\|_{X_0^*}^{p'}; \\ t \|x_{1,opt}\|_{X_1} = \langle y_*, x_{1,opt} \rangle, \end{cases}$$

where  $\frac{1}{p} + \frac{1}{p'} = 1$ .



## Couple $(\ell^p, X)$

Consider a particular but important case of couple  $(\ell^p, X)$  on  $\mathbb{R}^n$

$$L_{p,1}(t, x; \ell_p, X) = \inf_{x=x_0+x_1} \left( \frac{1}{p} \|x_0\|_{\ell_p}^p + t \|x_1\|_X \right),$$

where  $1 < p < +\infty$ . Consider the following function

$$F_0(u) = \frac{1}{p} \|u\|_{\ell_p}^p, \quad \nabla F_0(v) = \left\{ |v|^{p-1} \operatorname{sgn}(v) \right\}$$

Let us define the set  $\Omega$  by

$$\Omega = \{v \in \mathbb{R}^n : \nabla F_0(v) \in t\mathcal{B}_{X^*}\},$$

## Couple $(\ell^p, X)$

$\Omega = \{v \in \mathbb{R}^n : \nabla F_0(v) \in t\mathcal{B}_{X^*}\}$ . There are two cases

**Case 1: If  $x \in \Omega$**

then the optimal decomposition for  $L_{p,1}(t, x; \ell_p, X)$  is given by

$$x_{0,opt} = x \text{ and } x_{1,opt} = 0$$

**Case 2: If  $x \notin \Omega$**

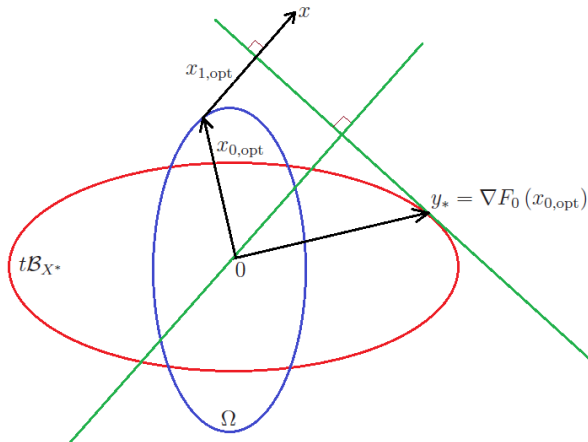
### Theorem

*The decomposition  $x = x_{0,opt} + x_{1,opt}$  is optimal for  $L_{p,1}(t, x; \ell_p, X)$  if and only if*

- (a)  $\|\nabla F_0(x_{0,opt})\|_{X^*} = t$
- (b)  $\langle x_{1,opt}, \nabla F_0(x_{0,opt}) \rangle = t \|x_{1,opt}\|_X$ .

# Geometry of optimal decomposition for couple $(\ell^p, X)$

$x_{1,\text{opt}}$  is orthogonal to the supporting hyperplane to  $t\mathcal{B}_{X^*}$  at  $y_*$



## Case $p = 2$ : Couple $(\ell^2, X)$

$$F_0(u) = \frac{1}{2} \|u\|_{\ell^2}^2, \quad \nabla F_0(v) = v$$

The sets  $\Omega$  and  $t\mathcal{B}_{X^*}$  coincide

$$\Omega = t\mathcal{B}_{X^*} = \{u \in \mathbb{R}^n : \|u\|_{X^*} \leq t\}$$

Corollary (for  $x \notin \Omega$ )

$$\|x_{0,opt}\|_{X^*} = t \text{ and } \langle x_{0,opt}, x - x_{0,opt} \rangle = t \|x - x_{0,opt}\|_X.$$

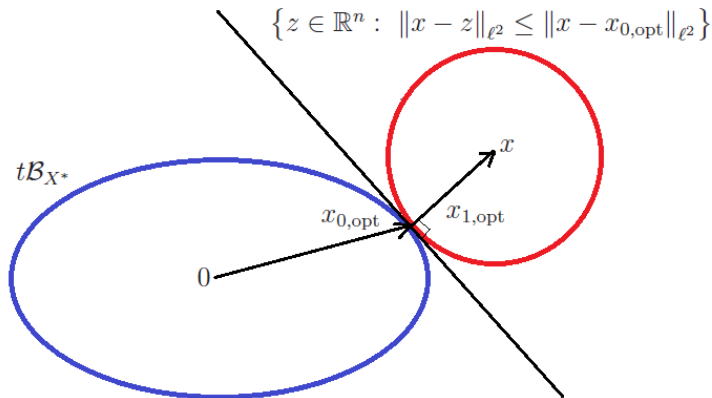
## Case $p = 2$ : Couple $(\ell^2, X)$

### Theorem

Let  $x_{0,opt}$  be an exact minimizer for  $L_{2,1}(t, x; \ell^2, X)$ . Then  $x_{0,opt}$  is the nearest element of  $t\mathcal{B}_{X^*}$  to the point  $x$  in the metric of  $\ell^2$ :

$$E(t, x; \ell^2, X^*) = \inf_{\|x_0\|_{X^*} \leq t} \|x - x_0\|_{\ell^2} = \|x - x_{0,opt}\|_{\ell^2}.$$

# Geometry of optimal decomposition for couple $(\ell^2, X)$



## Illustration in the plane

Consider couple  $(\ell^3, X)$  in the plane where the unit ball of  $X$  is the rotated ball of  $\ell^1$  by the rotation matrix

$$R_\theta = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix},$$

for  $\theta = 30^\circ$ . We have that

$$\|x\|_X = \|R_\theta^{-1}x\|_{\ell^1} = \left| \frac{\sqrt{3}}{2}x_1 - \frac{1}{2}x_2 \right| + \left| \frac{1}{2}x_1 + \frac{\sqrt{3}}{2}x_2 \right|.$$

$$\nabla F_0(u) = \left[ |u_1|^2 \operatorname{sgn}(u_1), |u_2|^2 \operatorname{sgn}(u_2) \right].$$

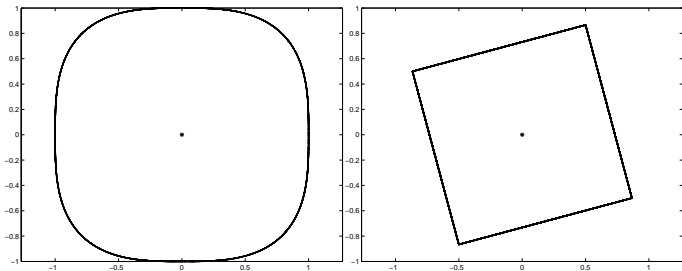
The set  $\Omega$  can be written as

$$\Omega = \left\{ v \in \mathbb{R}^2 : \left\| \left[ |v_1|^2 \operatorname{sgn}(v_1), |v_2|^2 \operatorname{sgn}(v_2) \right]^T \right\|_{X^*} \leq t \right\},$$

# Illustration in the plane

where the norm in  $X^*$  is given by

$$\|y\|_{X^*} = \|R_\theta y\|_{\ell^\infty} = \max \left\{ \left| \frac{\sqrt{3}}{2} y_1 + \frac{1}{2} y_2 \right|, \left| -\frac{1}{2} y_1 + \frac{\sqrt{3}}{2} y_2 \right| \right\}.$$

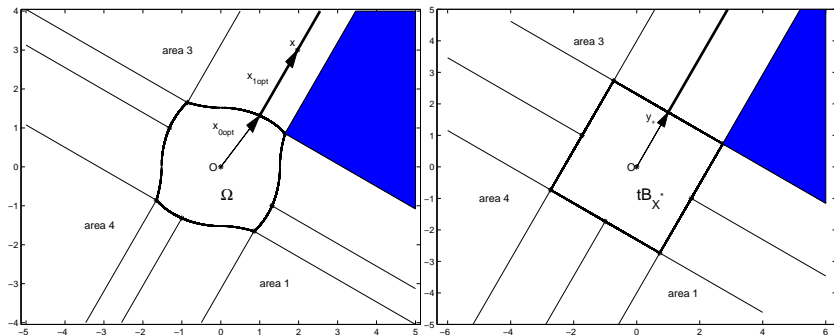


**Figure :** Unit ball of  $\ell^3$  (left) and unit ball of  $X$ : rotated unit ball of  $\ell^1$  (right).



# Illustration in the plane

Geometry of Optimal Decomposition for the Couple  $(\ell_p, X)$  for  $p = 3$ ,  $X = R_\theta(\ell_1)$  and  $\theta = 30^\circ$ . The set  $\Omega$  could be of rather complicated structure.



# Dual characterization of optimal decomposition

## Theorem (general case)

Let  $x \in X_0 + X_1$ ,  $1 < p_0, p_1 < \infty$  and let  $t > 0$  be a fixed parameter. The decomposition  $x = x_{0,opt} + x_{1,opt}$  is optimal for

$$L_{p_0, p_1}(t, x; X_0, X_1) = \inf_{x=x_0+x_1} \left( \frac{1}{p_0} \|x_0\|_{X_0}^{p_0} + \frac{t}{p_1} \|x_1\|_{X_1}^{p_1} \right),$$

if and only if  $\exists y_* \in X_0^* \cap X_1^*$  such that

$$\begin{cases} \frac{1}{p_0} \|x_{0,opt}\|_{X_0}^{p_0} = \langle y_*, x_{0,opt} \rangle - \frac{1}{p_0'} \|y_*\|_{X_0^*}^{p_0'}; \\ \frac{t}{p_1} \|x_{1,opt}\|_{X_1}^{p_1} = \langle y_*, x_{1,opt} \rangle - \frac{t}{p_1'} \left\| \frac{y_*}{t} \right\|_{X_1^*}^{p_1'}. \end{cases}$$

Thank you for your attention!