

# TD 3 – Inverse problems and approximation

Mathematics of data

02/10/24

## Exercise 1.

**Part 1.** Let  $f_0 \in \mathbb{R}^n$  be a discrete signal. We observe the signal  $y := f_0 \star h + \varepsilon$ , where  $h \in \mathbb{R}^n$  is a low-pass filter and  $\varepsilon \in \mathbb{R}^n$  is noise. For some  $\lambda \in \mathbb{R}_+^*$ , we look for

$$\operatorname{argmin}_{f \in \mathbb{R}^n} \frac{1}{2} \|f \star h - y\|^2 + \frac{\lambda}{2} \|f\|^2.$$

1. Using the Fourier decomposition, prove that for an optimal  $f \in \mathbb{R}^n$ ,

$$\hat{f}_k = \frac{\overline{\hat{h}_k}}{|\hat{h}_k|^2 + \lambda} \hat{y}_k,$$

where  $\overline{\hat{h}_k}$  denotes the complex conjugate of  $\hat{h}_k$ .

2. Why does  $\lambda > 0$  improve the deconvolution in presence of noise?

**Part 2.** For  $f \in \mathbb{R}^n$ , denote  $Gf := (f_i - f_{i-1})_i$  (considering indexes modulo  $n$ ).

1. What is the adjoint operator  $G^\top : \mathbb{R}^n \rightarrow \mathbb{R}^n$  for the canonical inner product? (i.e.  $\langle Gf, u \rangle = \langle f, G^\top u \rangle$ )
2. Show that  $G$ ,  $G^\top$  and  $L := GG^\top$  are discrete convolution operators and give their associated filters  $g$ ,  $\tilde{g}$  and  $\ell$ .
3. Compute the discrete Fourier coefficients  $\hat{g}$ . Express  $\hat{g}$  and  $\hat{\ell}$  as a function of  $\hat{g}$ .

**Part 3.** We now consider the problem

$$\operatorname{argmin}_{f \in \mathbb{R}^n} \frac{1}{2} \|f \star h - y\|^2 + \frac{\lambda}{2} \|Gf\|^2.$$

1. Using the Fourier decomposition and the results from part 2, prove that for an optimal  $f \in \mathbb{R}^n$ ,

$$\hat{f}_k = \frac{\overline{\hat{h}_k}}{|\hat{h}_k|^2 + \lambda \hat{\ell}_k} \hat{y}_k.$$

2. How does this expression differ from part 1?

**Exercise 2.** For an arbitrary  $y \in [0, 1]$ , consider  $f := 1_{[y, 1]} \in L^2([0, 1])$ , the indicator function of the interval  $[y, 1]$ . Denote  $\langle f, g \rangle := \int_0^1 fg$  the inner product and  $\|f\|^2 := \langle f, f \rangle$ .

1. For  $M \in \mathbb{N}^*$ , denote  $\theta_k := \sqrt{M} 1_{[\frac{k}{M}, \frac{k+1}{M}]}$  for  $0 \leq k < M$ . Show that  $(\theta_k)_k$  is an orthonormal family and give the expression for the linear approximation

$$f_M := \sum_k \langle f, \theta_k \rangle \theta_k.$$

2. Bound  $\|f - f_M\|$  as a function of  $M$ , independently of  $y$ , using a bound as sharp as possible.
3. Denote  $\theta := 1_{[0, 1]}$  and  $\psi := 1_{[0, 1/2]} - 1_{[1/2, 1]}$  (it is called the Haar wavelet). Denote, for  $j \leq 0$ , and  $0 \leq n < 2^{-j}$  the wavelet functions as  $\psi_{j,n} := \frac{1}{2^{j/2}} \psi(2^{-j}x - n)$ . Draw the wavelets  $\psi_{-1,1}$  and  $\psi_{-2,3}$ .

4. For some  $j_{\min} < 0$ , show that

$$\{\theta\} \cup \{\psi_{j,n} : 0 \geq j \geq j_{\min} \text{ and } 0 \leq n < 2^{-j}\}$$

is an orthogonal family. What is the space spanned by this family ?

5. For each  $j$ , what is the set  $\Sigma_j$  of index  $n$  where  $\langle f, \psi_{j,n} \rangle$  is non-zero? For these  $n \in \Sigma_j$ , bound  $|\langle f, \psi_{j,n} \rangle|$  as a function of  $j$ .

6. For  $T > 0$  define the non-linear approximation of  $f$  as

$$\hat{f} := \langle f, \theta \rangle \theta + \sum_{|\langle f, \psi_{j,n} \rangle| > T} \langle f, \psi_{j,n} \rangle \psi_{j,n}.$$

Bound as a function of  $T$  the number  $M$  of non-zero coefficients

$$M := |\{(j, n) : |\langle f, \psi_{j,n} \rangle| > T\}|.$$

7. Defining  $j_0 := \lfloor \log_2(T^2) \rfloor$  a cutoff scale, we define an approximation using

$$\tilde{f}_T := \langle f, \theta \rangle \theta + \sum_{j \geq j_0, n \in \Sigma_j} \langle f, \psi_{j,n} \rangle \psi_{j,n}.$$

Show that  $\|f - \hat{f}_T\| \leq \|f - \tilde{f}_T\|$ .

8. Bound  $\|f - \tilde{f}_T\|$  as a function of  $j_0$  and then as a function of  $M$ . Compare the decay with  $M$  of  $\|f - f_M\|$  and  $\|f - \hat{f}_T\|$ .

*If you have finished all the exercises, you can move on to the TP3 on [github.com/vcastin/teaching](https://github.com/vcastin/teaching)*