

Inverse Problems - Exercise Sheet 1

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Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be Banach spaces and $K: X \rightarrow Y$ be a linear continuous operator. For a given datum $f \in Y$ consider the inverse problem of finding $u \in X$ such that

$$Ku = f. \quad (1)$$

We say that the inverse problem (1) is *well-posed* in the sense of Hadamard if it admits unique solution for all $y \in Y$ and if solutions are continuous with respect to perturbations, i.e., if it holds

$$\|Ku_j - f\|_Y \rightarrow 0 \implies \|u_j - u\|_X \rightarrow 0.$$

Problem (1) is *ill-posed* if it is not well-posed.

Exercise 1.1 - Matrix inversion (30 pts)

Let $n \geq 1$ and consider $X = Y = \mathbb{R}^n$. Suppose that $K \in \mathbb{R}^{n \times n}$ is a positive definite symmetric matrix. Then by the spectral theorem

$$K = \sum_{j=1}^n \lambda_j k_j \otimes k_j$$

with $\{k_j\}_{j=1}^n$ orthonormal basis of \mathbb{R}^n and $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n > 0$ eigenvalues, where the tensor product between two vectors $a, b \in \mathbb{R}^n$ is the matrix $a \otimes b := ab^T$, that is, $(a \otimes b)_{ij} = a_i b_j$.

(a) Suppose that $Ku = f$ and $Ku^\delta = f^\delta$. Show that

$$\|u - u^\delta\| \leq \frac{1}{\lambda_n} \|f - f^\delta\|,$$

where $\|\cdot\|$ is the Euclidean norm in \mathbb{R}^n .

Hint: Note that $(a \otimes b)u = (b \cdot u)a$, where \cdot is the scalar product in \mathbb{R}^n .

(b) In the setting of (a), prove the relative error bound

$$\frac{\|u - u^\delta\|}{\|u\|} \leq \frac{\lambda_1}{\lambda_n} \frac{\|f - f^\delta\|}{\|f\|}.$$

(c) Is the inverse problem $Ku = f$ well-posed?

(d) $\kappa := \lambda_1 \lambda_n^{-1}$ is called *condition number* of the matrix K . In terms of real-world reconstructions, i.e., in the presence of noise, is it better for κ to be small or large?

Exercise 1.2 - Differentiation (30 pts)

Let $X = C([0, 1])$ be the space of continuous functions equipped with the supremum norm

$$\|u\|_{\infty} = \sup_{x \in [0, 1]} |u(x)|, \quad \text{for all } u \in X.$$

Let $Y = \{f \in C^1([0, 1]) : f(0) = 0\}$, with $C^1([0, 1])$ the space of continuously differentiable functions. Define the linear operator $K : X \rightarrow Y$ by

$$(Ku)(x) = \int_0^x u(y) dy,$$

for all $u \in X$ and $x \in [0, 1]$.

- (a) Show that the inverse problem $Ku = f$ admits a unique solution for all $f \in Y$.
- (b) Prove that the inverse problem $Ku = f$ is ill-posed when Y is equipped with the supremum norm.

Hint: Consider noisy data of the form $f^{\delta} = f + n^{\delta}$ for some suitable noise $n^{\delta} \in Y$.

- (c) Show that the inverse problem $Ku = f$ is well-posed when Y is equipped with the norm $\|\cdot\|_{C^1}$, where $\|u\|_{C^1} = \|u\|_{\infty} + \|u'\|_{\infty}$ for all $u \in C^1([0, 1])$.

Exercise 1.3 - Closed range (20 pts)

Let $(X, \|\cdot\|_X)$, $(Y, \|\cdot\|_Y)$ be Banach spaces and $K : X \rightarrow Y$ be a bounded linear operator. On X we define the equivalence relation

$$x_1 \sim x_2 \iff x_1 - x_2 \in \ker(K).$$

Define $\hat{X} = X/\sim$ to be the quotient space w.r.t. \sim , with equivalence classes denoted by $[x]$. Introduce

$$\|[x]\|_{\hat{X}} := \inf_{y \in [x]} \|y\|_X,$$

and the operator $\hat{K} : \hat{X} \rightarrow \text{rg}(K)$ defined by $\hat{K}[x] := Kx$, where $\text{rg}(K)$ is the range of K .

- (a) Show that $(\hat{X}, \|\cdot\|_{\hat{X}})$ is a Banach space.

Hint: Use that a normed space is complete if and only if every absolutely convergent series is convergent

- (b) Show that \hat{K} is well-defined, linear and bounded.

- (c) Show that if $\text{rg}(K)$ is closed then \hat{K}^{-1} is continuous.

Hint: Note that \hat{K} is bijective and use the open mapping theorem

- (d) Prove that if \hat{K}^{-1} is continuous then $\text{rg}(K)$ is closed.

On the space $L^2([-\pi, \pi]^2)$ we define the scalar product

$$\langle u, v \rangle := \int_{[-\pi, \pi]^2} u(x) \overline{v}(x) dx.$$

With respect to such scalar product, an orthonormal basis of $L^2([-\pi, \pi]^2)$ is given by the functions

$$e_l(x_1, x_2) := c_l e^{i(l_1 x_1 + l_2 x_2)},$$

for all $l = (l_1, l_2) \in \mathbb{Z}^2$, with $c_l \in \mathbb{R}$ suitable normalization constant. Any function $u \in L^2([-\pi, \pi]^2)$ admits a representation in terms of its Fourier series

$$u = \sum_{l \in \mathbb{Z}^2} \hat{u}_l e_l, \quad \hat{u}_l := \langle u, e_l \rangle.$$

Recall that $\hat{u} := (\hat{u}_l)_l \in \ell^2(\mathbb{Z}^2)$. Also, an operator is compact if it is the limit of finite-range operators in operator norm. Recall that a space is finite dimensional if and only if its closed unit ball is compact.

Exercise 1.4 - Convolution (20 pts)

For $k \in L^2([-\pi, \pi]^2)$ define the convolution operator $K: L^2([-\pi, \pi]^2) \rightarrow L^2([-\pi, \pi]^2)$ by setting

$$Ku := k * u, \quad (k * u)(x) := \int_{[-\pi, \pi]^2} k(x - y) u(y) dy,$$

where we implicitly assume that k and u are extended periodically to the whole \mathbb{R}^2 .

(a) Show that

$$\widehat{(Ku)}_l = \frac{1}{c_l} \hat{k}_l \hat{u}_l \quad \text{for all } l \in \mathbb{Z}^2.$$

Moreover, provide the inverse of K in case $\hat{k}_l \neq 0$ for all $l \in \mathbb{Z}^2$.

(b) Prove that K is compact. Deduce that, in case $\hat{k}_l \neq 0$ for infinitely many $l \in \mathbb{Z}^2$, $\text{rg}(K)$ is not closed.

Note: Thanks to Exercise 1.3, this shows that de-convolution is an ill-posed inverse problem

Inverse Problems - Exercise Sheet 2

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Motivation: compact operators often arise in inverse problems applications. In this worksheet we highlight a few of their properties.

Let X be a Banach space with norm $\|\cdot\|_X$ and denote by B_X its unit ball, that is,

$$B_X := \{x \in X : \|x\|_X \leq 1\}.$$

Let Y be a Banach space with norm $\|\cdot\|_Y$. We denote by $\mathcal{L}(X, Y)$ the space of linear continuous operators $K: X \rightarrow Y$. Recall that $\mathcal{L}(X, Y)$ is a Banach space with the operator norm. We set $\mathcal{L}(X) := \mathcal{L}(X, X)$. For a subset $M \subset X$ we denote the distance of $x \in X$ to M by

$$\text{dist}(x, M) := \inf_{m \in M} \|x - m\|_X.$$

Compact operators: We say that $K \in \mathcal{L}(X, Y)$ is a *compact operator* if the closure of $K(B_X)$ is compact in Y . We denote the space of compact operators from X to Y by $\mathcal{K}(X, Y)$, and $\mathcal{K}(X) := \mathcal{K}(X, X)$.

Finite rank operators: We say that $K \in \mathcal{L}(X, Y)$ has *finite rank* if $K(X)$ is finite dimensional. Note that finite rank operators are clearly compact.

Adjoint: Let $K \in \mathcal{L}(X, Y)$. The *adjoint* of K is the linear operator $K^*: Y^* \rightarrow X^*$ defined by

$$\langle K^* y^*, x \rangle_{X^*, X} = \langle y^*, Kx \rangle_{Y^*, Y} \quad \text{for all } x \in X, y^* \in Y^*.$$

It is well-known that $K^* \in \mathcal{L}(Y^*, X^*)$, with $\|K\| = \|K^*\|$.

Summable sequences: For $p \geq 1$ denote by ℓ_p the space of square summable sequences with norm $\|\cdot\|_{\ell_p}$, that is,

$$\ell_p := \left\{ x = (x_j)_{j \in \mathbb{N}} : \sum_{j=1}^{\infty} |x_j|^p < \infty \right\}, \quad \|x\|_{\ell_p} := \left(\sum_{j=1}^{\infty} |x_j|^p \right)^{1/p}.$$

Recall that ℓ_2 is a Hilbert space with scalar product

$$\langle x, y \rangle_{\ell_2} := \sum_{j=1}^{\infty} x_j y_j.$$

Exercise 2.1 (30 pts) - The identity is not compact in infinite dimensions. Let X be a normed space. Suppose $M \subset X$ is a closed linear subspace with $M \neq X$.

(a) Let $x \in X \setminus M$. Prove that $\text{dist}(x, M) > 0$.

(b) (Riesz's Lemma) Prove that for all $\varepsilon > 0$ there exists $x \in X$ such that

$$\|x\|_X = 1, \quad \text{dist}(x, M) \geq 1 - \varepsilon.$$

(c) Suppose in addition that X is a Banach space. Prove that the identity map $I: X \rightarrow X$ is compact if and only if $\dim X < \infty$.

Hint: You can use point (b).

Exercise 2.2 (20 pts) - Image through compact operators. Let X, Y be Banach spaces and assume that X is reflexive. Let $K \in \mathcal{L}(X, Y)$ and $M \subset X$ be closed, convex and bounded.

- (a) Show that $K(M)$ is closed in Y .

Hint: Recall that a linear operator $K: X \rightarrow Y$ between normed spaces is continuous if and only if it is weak-to-weak continuous. Also recall that if $M \subset X$ is convex then the closure of M coincides with its weak closure.

- (b) In addition, assume that $K \in \mathcal{K}(X, Y)$. Show that $K(M)$ is compact.

Exercise 2.3 (20 pts) - Compact operators on Hilbert spaces. Let X, Y be real Hilbert spaces and $K \in \mathcal{L}(X, Y)$. Let $x_n, x \in X$ for $n \in \mathbb{N}$.

- a) Show that $x_n \rightarrow x$ strongly in X if and only if

$$x_n \rightharpoonup x \text{ weakly in } X \text{ and } \|x_n\|_X \rightarrow \|x\|_X.$$

- b) Show that K is compact if and only if the following condition holds:

$$\text{If } x_n \rightharpoonup x \text{ weakly in } X, \text{ then } Kx_n \rightarrow Kx \text{ strongly in } Y.$$

Hint: The statements in Exercise 2.2 could be useful.

Exercise 2.4 (10 pts) - Range of compact operators.

- a) Let X be a Banach space with $\dim X = +\infty$ and let $K \in \mathcal{K}(X)$. Show that K cannot be surjective, that is, there exists $y \in Y$ such that the equation

$$Kx = y$$

has no solution in X .

- b) Let X, Y be Banach spaces. Let $K \in \mathcal{K}(X, Y)$. Show that $\text{rg}(K)$ is closed if and only if $\dim \text{rg}(K) < \infty$.

Hint: You might find Exercise 2.1 point (c) and Exercise 1.3 useful. Also recall that the composition (in whichever order) of a compact operator with a bounded operator is compact.

Exercise 2.5 (20 pts) Define the operator $K: \ell_2 \rightarrow \ell_2$ by

$$(Kx)_j := \frac{x_j}{j}, \quad \text{for all } j \in \mathbb{N}.$$

- (a) Show that K is well-defined, linear and compact.

Hint: ℓ_2 is a Hilbert space, therefore you can use Exercise 2.3 to prove compactness. Also, you could use the dominated convergence Theorem in the ℓ_1 setting.

- (b) Show that range of K is not closed by finding some element $y \in \overline{\text{rg}(K)} \setminus \text{rg}(K)$.

Inverse Problems - Exercise Sheet 3

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Exercise 3.1 (25 pts) - Compact operators are not coercive in infinite dimensions. Let X, Y be Banach spaces, $K \in \mathcal{L}(X, Y)$. Assume there exists a constant $c > 0$ such that

$$\|Kx\|_Y \geq c\|x\|_X \quad \text{for all } x \in X. \quad (1)$$

Show that K is compact if and only if $\dim X < +\infty$.

Remark: An operator K satisfying (1) is called coercive. Notice that coercivity implies injectivity.

Exercise 3.2 (25 pts) - Non-injective compact operators form a dense set. Let X, Y be Banach spaces, with $\dim X = \infty$. Suppose that $K \in \mathcal{K}(X, Y)$

- (a) Show that there exists a sequence $\{x_n\}_n$ in X such that $\|x_n\|_X = 1$ for all $n \in \mathbb{N}$ and $Kx_n \rightarrow 0$ strongly.
- (b) Using (a), prove that the set

$$\{K \in \mathcal{K}(X, Y) : K \text{ is not injective}\}$$

is dense in $\mathcal{K}(X, Y)$ with respect to the operator norm.

Exercise 3.3 (25 pts) - Existence of minimal norm elements. Let X be a Banach space. Let $A \subset X$ be convex, closed and non-empty.

- (a) Prove that the function

$$x \in X \mapsto \text{dist}(x, A)$$

is lower semicontinuous with respect to the weak topology on X .

- (b) Assume that X is reflexive. Prove that A has a minimal norm element, that is, show that there exists $\hat{x} \in A$ such that

$$\|\hat{x}\| = \inf_{x \in A} \|x\|$$

- (c) Assume that X is reflexive. Let $M \subset X$ be a closed linear subspace. Prove that in Riesz's Lemma of Exercise 2.1 point (b) one can choose $\varepsilon = 0$, that is, show that there exists $x \in X$ such that

$$\|x\|_X = 1, \quad \text{dist}(x, M) \geq 1.$$

Exercise 3.4 (25 pts) - Computation of Moore-Penrose Inverse For the following operators T in $\mathcal{L}(X, Y)$ compute the Moore-Penrose inverse of T , and check whether $\text{Dom}(T^\dagger) = Y$.

- (a) For Hilbert spaces X, Y and fixed $u \in X$ with $u \neq 0$ and $v \in Y$ with $v \neq 0$ let $T: X \rightarrow Y$ be defined according to $Tx = v\langle u, x \rangle$ (where $\langle \cdot, \cdot \rangle$ denote the scalar product on X).
- (b) Let $T: L^2([0, 1]) \rightarrow L^2([0, 1])$ according to

$$[Tx](t) = \int_0^t x(s) ds \quad \text{for almost every } t \in [0, 1].$$

Inverse Problems - Exercise Sheet 4

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Exercise 4.1 (50 pts) - Radon transform. For $f: B \rightarrow \mathbb{R}$ continuous, where $B := \{x \in \mathbb{R}^2 : \|x\| \leq 1\}$, we define the Radon transform as

$$\mathcal{R}[f](\theta, s) = \int_{-\infty}^{\infty} f(s \mathbf{w}(\theta) + t \mathbf{w}^{\perp}(\theta)) dt, \quad (\theta, s) \in [0, 2\pi] \times \mathbb{R},$$

with f extended by 0 outside B and $\mathbf{w}(\theta) = (\cos(\theta), \sin(\theta))^t$, $\mathbf{w}^{\perp}(\theta) = (-\sin(\theta), \cos(\theta))^t$.

- (a) Show that the Radon transform can be extended to a linear continuous operator from $L^2(B)$ into $L^2(Q)$ with $Q := [0, 2\pi] \times [-1, 1]$
- (b) Prove that the adjoint of the Radon transform \mathcal{R}^* , also called **backprojection operator**, has the form

$$\mathcal{R}^*[g](x) = \int_0^{2\pi} g(\theta, x \cdot \mathbf{w}(\theta)) d\theta. \quad (1)$$

Note: Adjoint means that $\langle \mathcal{R}[f], g \rangle_{L^2(Q)} = \langle f, \mathcal{R}^*[g] \rangle_{L^2(B)}$, $\forall f \in L^2(B), g \in L^2(Q)$.

Instructions for next exercise: Please bring your code on a pendrive on the day of the class

Exercise 4.2 (50 pts) - Numerical implementation of the Radon transform. We will now work in the discrete case where functions defined on $[-1, 1] \times [-1, 1]$ will be represented by square matrices.

- (a) Using the provided template in Matlab `radontrans.m`, implement the Radon transform.

Hint: Consider using the `imrotate` function. If the code is taking more than 10 lines, reconsider your approach.

- (b) The output of a Radon transform is called a sinogram, denoted by

$$g(\theta, s) = \mathcal{F}[f](\theta, s).$$

From the sinogram at the left hand side of Figure 1, which could be the underlying source f that produces it?

- (c) Using the provided template in Matlab `backprojection.m`, implement the adjoint of the Radon transform defined in (1).

Hint: Notice that for fixed θ , the function $G(x) = g(\theta, x \cdot \mathbf{w}(\theta))$, $x \in \mathbb{R}^2$, is constant along lines parallel to $\mathbf{w}^{\perp}(\theta)$.

Hint 2: Consider using the `imrotate` function. If the code is taking more than 10 lines, reconsider your approach.

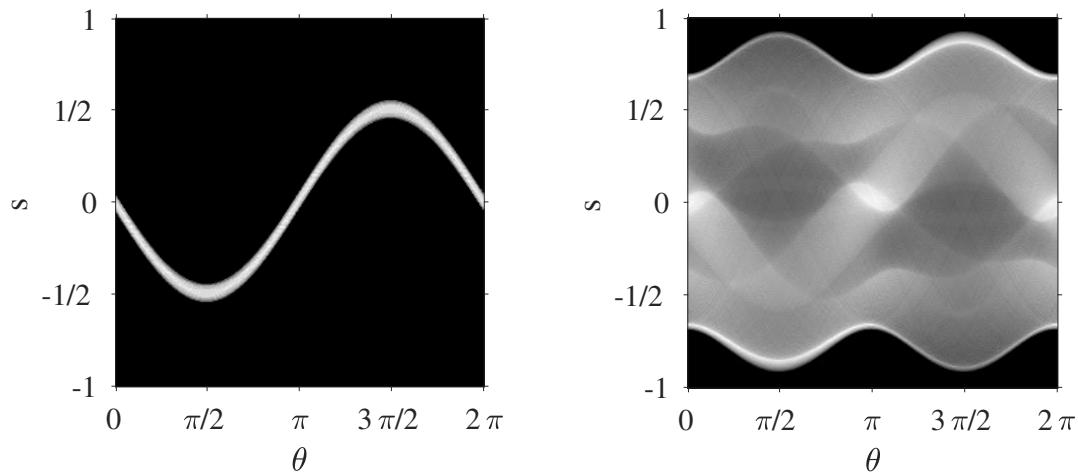


Figure 1: Example sinograms of unknown sources

- (d) The **backprojection algorithm** corresponds to simply applying the adjoint operator \mathcal{R}^* to the sinogram data. Write a script where you apply your implemented adjoint to the provided data `sinogram1.mat` and `sinogram2.mat`, that corresponds to the ones presented in Figure 1. What do you obtain?

Remark: It is required to use the same family of angles θ for the Radon transform and its adjoint.
For the provided data $\theta = \text{linspace}(0, 360, 256)$.

- (e) Could you explain or interpret what is the backprojection algorithm $\mathcal{R}^*[\mathcal{R}[f]](x)$ doing?

Inverse Problems - Exercise Sheet 5

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Exercise 5.1 (50 pts) - Singular values of the Radon transform For this exercise we fix the weight function $w(s) = \sqrt{1-s^2}$. Furthermore, we will require the Chebyshev polynomials

$$U_m(s) = \frac{\sin[(m+1)\arccos(s)]}{\sin(\arccos(s))}.$$

They satisfy the following properties:

- They are orthogonal in the $L^2([-1, 1], w)$, the w -weighted space. That is

$$\int_{-1}^1 w(s) U_m(s) U_{m'}(s) ds = \frac{\pi}{2} \delta_{m, m'}.$$

- They form an orthogonal basis for $L^2([-1, 1], w)$. Consequently, the functions $U_m(s)w^{-1}(s)$ form a basis of $Z = L^2([-1, 1], w^{-1})$.
- They satisfy the following integral property

$$\int_{-w(s)}^{w(s)} U_m(\mathbf{w}(\theta') \cdot (s\mathbf{w}(\theta) + t\mathbf{w}^\perp(\theta))) dt = \frac{2}{m+1} \frac{\sin((m+1)(\theta - \theta')) \sin((m+1)\arccos(s))}{\sin(\theta - \theta')}$$

- a) Prove that the Radon transform is continuous from $L^2(B)$ to $Z = L^2(\Omega, w(s)^{-1})$, the L^2 space weighted with w^{-1} , that is, show the following inequality

$$\|\mathcal{R}f\|_{L^2(\Omega, w^{-1})} = \int_0^{2\pi} \int_{-1}^1 |\mathcal{R}[f](\theta, s)|^2 w(s)^{-1} ds d\theta \leq 4\pi \|f\|_{L^2(B)}$$

Hint 1: Notice that the interval $[-w(s), w(s)]$ corresponds to the effective interval in which the Radon transform integrates, for any direction θ and displacement s .

Hint 2: To integrate, reuse the same idea presented in the last exercise sheet.

- b) Under this new metric on the image space of the Radon transform, find the adjoint operator \mathcal{R}^* .

Hint: you can use the results from the previous exercise class.

- c) Consider functions of the form $g(s, \theta) = U(s)w(s)v(\theta)$, for $U \in L^2([-1, 1])$, $v \in L^2([0, 2\pi))$. Check that

$$(\mathcal{R}\mathcal{R}^*g)(\theta, s) = \int_{-w(s)}^{w(s)} \int_0^{2\pi} U(\mathbf{w}(\theta') \cdot (s\mathbf{w}(\theta) + t\mathbf{w}^\perp(\theta))) v(\theta') d\theta' dt$$

- d) Check that

$$(\mathcal{R}\mathcal{R}^*g_m)(\theta, s) = \frac{4\pi}{m+1} w(s) U_m(s) \bar{v}(\theta),$$

where U_m are the Chebyshev polynomials, and $\bar{v}(\theta) = \frac{1}{2\pi} \int_0^{2\pi} \frac{\sin((m+1)(\theta-\theta'))}{\sin(\theta-\theta')} v(\theta') d\theta'$.

- e) Consider the family $Y_l(\theta) = \frac{1}{\sqrt{2\pi}} e^{-il\theta}$ of functions that form a basis for $L^2([0, 2\pi))$ and satisfy

$$\bar{Y}_l = \begin{cases} Y_l & \text{if } -m \leq l \leq m \text{ and } l - m \text{ is even} \\ 0 & \text{otherwise.} \end{cases}$$

Write a basis for Z . From it, find the eigenfunctions and nonzero eigenvectors of $\mathcal{R}\mathcal{R}^*$.

- f) Obtain the singular values for the Radon transform \mathcal{R} . How ill-posed is the respective inverse problem? Why there seem to be many zero singular values?

Exercise 5.2 (50 pts) - Singular values of the Radon transform and the limited angle problem.

- a) Write a code that returns the matrix associated to the Radon transform of a function with support in the unitary ball. For it use the provided template `RadonMatrix.py` (in Python, feel free to do an equivalent Matlab version). Take into consideration the following points
- The Radon transform goes from images in two dimensions to a function with two variables. Since it is written as a matrix, it will take vectors into vectors. Take care to be consistent in these transformations.
 - The obtained matrix will be a $|S||\Theta| \times N^2$ sized matrix. It is recommended that you use a sparse format for it.
 - The Radon transform is such that it acts only on objects that are supported in the unitary ball, so no weight should be given to pixels outside that area.
 - Feel free to consider reasonable approximations. One suggested example could be to approximate the line integral by an uniform Riemann sum.
 - Save the output results in the following sections, as computations could be a bit length (order of minutes).
- b) In the template `RadonMatrix.py` there is a script to test the Matrix with an image composed of a non-centered disk, use it to check your implementation.
- c) In the following points, fix $N = 40$, $S = 40$, $T = 60$. Use template `InvertData.py` to invert the operator for the cases $\theta \in [0, \pi]$, $\theta \in [0, \pi/4]$, and noiseless and 8% of noise level. What do you observe when plotting the reconstructed images?
- d) Write a script in which you plot the singular values (in a non-increasing fashion) of the matrix for the cases $\theta \in [0, 2\pi]$, $[0, \pi]$, $[0, \pi/2]$, $[0, \pi/4]$. What do you see and what would it explain?
- e) Write a script, in which for the case of $\theta \in [0, \pi/4]$ and 10% of noise, an approximate inversion is obtained by using all the singular values above: 0.001, 0.01, and 0.05. Comment on the results.

Inverse Problems - Exercise Sheet 6

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Exercise 6.1 (20 pts) - Tikhonov regularization with semi-norms. Let X, Y , and Z be Banach spaces, where X is additionally reflexive and Z is separable. Further, let $T : X \rightarrow Y$ be a linear and continuous operator, and $A : \text{dom}(A) \rightarrow Z^*$ be a densely defined, linear operator on the Banach space X . Moreover, assume that the range of A is closed in Z^* , the kernel of A is finite-dimensional, and that A is weakly-weakly* closed, which means that $x_n \rightharpoonup x$ and $Ax_n \xrightarrow{*} z$ implies $x \in \text{dom}(A)$ and $Ax = z$. For each $y^\delta \in Y, p, q \geq 1$, and $\alpha > 0$, show that there exists a solution of the minimization problem

$$\min_{x \in X} \frac{1}{q} \|Tx - y^\delta\|_Y^q + \frac{\alpha}{p} \|Ax\|_{Z^*}^p.$$

Hint: You may exploit: if the linear operator $S : \text{dom}(S) \rightarrow Y$ between the dense subspace $\text{dom}(S) \subset X$ and Y is closed and bijective, then S^{-1} is continuous. Further, note that the finite-dimensional subspace $\ker(A)$ is complemented in X .

Convex analysis. Let X be a Banach space, X^* its dual, and $f : X \rightarrow \mathbb{R} \cup \{-\infty, +\infty\}$ a function. We define the convex conjugate of f as the function $f^* : X^* \rightarrow \mathbb{R} \cup \{-\infty, +\infty\}$ defined by

$$f^*(x^*) = \sup_{x \in X} \langle x^*, x \rangle - f(x), \quad (1)$$

where $\langle x^*, x \rangle$ denotes the dual pairing between X and X^* . The convex conjugate satisfies the following properties:

- f^* is always convex and lower-semicontinuous.
- The convex biconjugate f^{**} (that is to apply the convex conjugate twice), corresponds to the closed convex hull of f , i.e. the largest lower semi-continuous convex function such that $f^{**} \leq f$. If f is proper, convex and lower semicontinuous, then $f^{**} = f$.
- **Fenchel duality:** For $f : X \rightarrow \mathbb{R} \cup \{\pm\infty\}$, $g : Y \rightarrow \mathbb{R} \cup \{\pm\infty\}$ convex functions and $A : X \rightarrow Y$ bounded linear operator, the following inequality holds

$$\inf_{x \in X} f(Ax) + g(x) \geq \sup_{y^* \in Y^*} -f^*(-y^*) - g^*(A^*y^*).$$

Equality holds under certain regularity conditions. The above inequality is also named the weak duality of the *primal* and *dual* optimization problems. In the case of equality of the two optimization problems, it is said that there is strong duality.

Exercise 6.2 (30 pts) - The convex conjugate.

- a) Consider a Banach space X with norm $\|\cdot\|_X$. Prove that the convex conjugate (1) of the norm is the convex indicator function over the unitary ball in the dual space, that is,

$$(\|\cdot\|_X)^*(x^*) = \mathbb{1}_{B^*}(x^*) = \begin{cases} 0 & \text{if } \|x^*\|_{X^*} \leq 1, \\ \infty & \text{if } \|x^*\|_{X^*} > 1. \end{cases}$$

- b) Consider the space $X = \mathbb{R}^n \times \mathbb{R}^n$ endowed with the norm $\|x\| = \sum_{i=1}^n \sqrt{y_i^2 + z_i^2}$, for $x = (y, z)$. Prove that for $x^* = (y^*, z^*) \in X$ one has

$$(\|\cdot\|)^*(x^*) = \begin{cases} 0 & \text{if } \max_i \sqrt{(y_i^*)^2 + (z_i^*)^2} \leq 1, \\ \infty & \text{if } \max_i \sqrt{(y_i^*)^2 + (z_i^*)^2} > 1, \end{cases} \quad (2)$$

The primal-dual algorithm. We want to compute a solution to the problem

$$\min_{x \in X} f(Ax) + g(x), \quad (3)$$

where:

- X, Y are two finite-dimensional real vector spaces equipped with an inner product $\langle \cdot, \cdot \rangle$ and norm.
- $A : X \rightarrow Y$ is a continuous linear operator
- $g : X \rightarrow [0, \infty)$ and $f : Y \rightarrow [0, \infty)$ are proper, convex, lower-semicontinuous functions.

For a function h the *resolvent operator* is defined as

$$(I + \tau \partial h)^{-1}(y) = \arg \min_x \left\{ \frac{\|x - y\|^2}{2\tau} + h(x) \right\}.$$

Then, for selected constants $\tau, \sigma > 0, \theta \in [0, 1]$, initial guess $(x^0, y^0) \in X \times Y$, f^* the convex conjugate of f , and A^* the adjoint operator of A , the *primal-dual algorithm* stopped at N iterations is

Algorithm 1 Primal-dual Algorithm

```
1: procedure PRIMAL-DUAL( $\tau, \sigma, \theta, N, x^0, y^0$ )
2:    $\bar{x}^0 = x^0$ 
3:   for  $n = 0, 1, 2, \dots, N$  do
4:      $y^{n+1} = (I + \sigma \partial f^*)^{-1}(y^n + \sigma A \bar{x}^n)$ 
5:      $x^{n+1} = (I + \tau \partial g)^{-1}(x^n - \tau A^* y^{n+1})$ 
6:      $\bar{x}^{n+1} = x^{n+1} + \theta(x^{n+1} - x^n)$ 
7:   end for
8:   return  $x^{N+1}$ .
9: end procedure
```

The ROF model: We now introduce the ROF model for image denoising. We model a 2D image u by a $N \times N$ matrix, $N > 0$ integer. Thus $u \in \mathbb{R}^{N \times N}$. The discrete gradient of u with Neumann boundary conditions is defined by $\nabla u = (\nabla_x u, \nabla_y u)$, where $\nabla_x u, \nabla_y u \in \mathbb{R}^{N \times N}$ are matrices defined by

$$(\nabla_x u)_{i,j} := \begin{cases} u_{i+1,j} - u_{i,j} & \text{if } i < N-1, \\ 0 & \text{if } i = N-1, \end{cases} \quad (\nabla_y u)_{i,j} := \begin{cases} u_{i,j+1} - u_{i,j} & \text{if } j < N-1, \\ 0 & \text{if } j = N-1. \end{cases} \quad (4)$$

For $u \in \mathbb{R}^{N \times N}$, $p = (p^x, p^y)$, $p^x, p^y \in \mathbb{R}^{N \times N}$ we define

$$\|u\|_2 := \sqrt{\sum_{i,j=1}^N u_{i,j}^2}, \quad \|p\|_{1,2} := \|p^x\|_2 + \|p^y\|_2. \quad (5)$$

Given a noisy image $D \in \mathbb{R}^{N \times N}$, the denoised image $u \in \mathbb{R}^{N \times N}$ according to the ROF model is a solution to the minimization problem

$$\min_{u \in \mathbb{R}^{N \times N}} \|\nabla u\|_{1,2} + \frac{\lambda}{2} \|u - D\|_2^2, \quad (6)$$

where $\lambda > 0$ is a fixed regularization parameter.

Exercise 6.3 (50 pts) - Implementing the primal-dual algorithm for ROF. In this exercise we will implement the primal-dual algorithm to compute solutions of the ROF model for image denoising at (6). We define the following:

- $X := \mathbb{R}^{N \times N}$ normed by $\|\cdot\|_X := \|\cdot\|_2$ as defined in (5). Notice that such norm is induced by the Hilbert product

$$\langle u, v \rangle_X := \langle u, v \rangle_2 := \sum_{i,j=1}^N u_{i,j} v_{i,j},$$

for $u, v \in X$.

- $Y := \mathbb{R}^{N \times N} \times \mathbb{R}^{N \times N}$ normed by $\|\cdot\|_Y := \|\cdot\|_{1,2}$ as defined in (5). This norm is induced by the Hilbert product

$$\langle p, q \rangle_Y := \langle p, q \rangle_{1,2} := \langle p^x, q^x \rangle_2 + \langle p^y, q^y \rangle_2$$

for $p = (p^x, p^y), q = (q^x, q^y) \in Y$.

- $A: X \rightarrow Y$ linear continuous operator defined by $Au := \nabla u$ where $\nabla = (\nabla_x, \nabla_y)$ is as in (4).
- $f: Y \rightarrow [0, \infty)$ defined by $f(p) := \|p\|_{1,2}$.
- Given $\lambda > 0$ and $D \in X$, define $g_\lambda: X \rightarrow [0, \infty)$ as $g_\lambda(u) := \frac{\lambda}{2} \|u - D\|_2^2$.

With the above definitions, please address the following questions.

- Compute A^* , the adjoint of the operator A .
- Following the provided template `primal_dual.py`, implement the operators A and A^* .
- By Exercise 6.2 we have that

$$f^*(p) = \begin{cases} 0 & \text{if } \max_{i,j} \sqrt{(p_{i,j}^x)^2 + (p_{i,j}^y)^2} \leq 1, \\ \infty & \text{if } \max_{i,j} \sqrt{(p_{i,j}^x)^2 + (p_{i,j}^y)^2} > 1. \end{cases}$$

Prove that the resolvent of f^* has the following form:

$$p = (I + \sigma \partial f^*)^{-1}(\tilde{p}) \iff p_{i,j} = \frac{\tilde{p}_{i,j}}{\max\left(1, \sqrt{(\tilde{p}_{i,j}^x)^2 + (\tilde{p}_{i,j}^y)^2}\right)}.$$

d) Following the provided template `primal_dual.py`, implement the resolvent of f^* .

e) Prove that the resolvent of g_λ has the following form:

$$(I + \tau \partial g_\lambda)^{-1}(\tilde{u}) = \frac{\tilde{u} + \lambda \tau D}{1 + \lambda \tau}.$$

f) Following the provided template `primal_dual.py`, implement the resolvent of g_λ .

g) Following the provided template `primal_dual.py`, implement the primal-dual algorithm. Then run the code to obtain reconstructions for two noise cases. Feel free to change the parameters to see different results (λ , the number of iterations, noise level). Changing the values of σ and τ is discouraged, as it is related to the convergence properties of this algorithm, topic not discussed in this exercise.