

Please submit exercises 3 and 4.

## Exercise 1 (Convexity of a sum)

Let  $\alpha \geq 0$  and  $E, F: \mathcal{U} \to \mathbb{R}_{\infty}$  be two convex functions. Prove the following two statements.

- a) The sum of the two functions  $E + \alpha F \colon \mathcal{U} \to \mathbb{R}_{\infty}$  is convex
- b) If, in addition  $\alpha > 0$  and F is strictly convex, then  $E + \alpha F$  is strictly convex.

# Exercise 2 (Convexity of data term)

Let  $\mathcal{U}, \mathcal{V}$  be normed spaces. Furthermore, let  $K \in \mathcal{L}(\mathcal{U}, \mathcal{V}), f \in \mathcal{V}$  and  $D: \mathcal{V} \to \mathbb{R}_{\infty}$  be defined as  $D(u) := \frac{1}{2} \|Ku - f\|_{\mathcal{V}}^2$ . Prove the following two statements.

- a) Show that D is convex.
- b) Show that D may not be strictly convex in general, even if K is injective.
- c) Let  $\mathcal{V}$  be an inner product space and  $\lambda \in (0, 1), u, v \in \mathcal{U}$ . Show that

$$D(\lambda u + (1-\lambda)v) = \lambda D(u) + (1-\lambda)D(v) - \frac{\lambda(1-\lambda)}{2} \|K(u-v)\|_{\mathcal{V}}^2.$$

### Exercise 3 (Convex conjugate)

Let  $\mathcal{U}$  be a Banach space and let  $E: \mathcal{U} \to \mathbb{R}_{\infty}$  be proper, lower semi-continuous and convex. Then the *Fenchel conjugate* or *convex conjugate* of E is defined to be the mapping  $E^*: \mathcal{U}^* \to \mathbb{R}_{\infty}$  with

$$E^*(v) := \sup_{u \in \mathcal{U}} \left\{ \langle v, u \rangle - E(u) \right\}.$$

- a) Compute the convex conjugates of the following functionals.
  - i.  $E \colon \mathbb{R} \to \mathbb{R}, E(u) = \frac{1}{p} |u|^p$  and  $1 < p, q < \infty$  with  $\frac{1}{p} + \frac{1}{q} = 1$ .
  - ii.  $E(u) = \frac{1}{2} ||u||^2$  for a Hilbert space  $\mathcal{U}$ .
  - iii.  $F(u) = E(\alpha u a) + \langle b, x \rangle + \beta$  for  $\alpha \neq 0, \beta \in \mathbb{R}, a \in \mathcal{U}, b \in \mathcal{U}^*$ .
- b) Let  $\mathcal{U}$  be a Hilbert space and  $E: \mathcal{U} \to \mathbb{R}_{\infty}$  a proper, lower semi-continuous and convex functional. Show that

$$p \in \partial E(u) \quad \Leftrightarrow \quad u \in \partial E^*(p)$$

for all  $u, p \in \mathcal{U}$ .

**Hint:** You may exploit the fact that under the stated assumptions  $E = E^{**}$  holds true.

### Exercise 4 (Differentiation)

Let  $\mathcal{U}$  be a Banach space and  $E: \mathcal{U} \to \mathbb{R}$  be a convex functional that is Fréchet-differentiable in  $u \in \mathcal{U}$ . Then

$$\partial E(u) = \{E'(u)\}.$$

#### Exercise 5 (Bregman iteration)

Let  $f \in \mathcal{V}, K \in \mathcal{L}(\mathcal{U}, \mathcal{V})$  and  $J : \mathcal{U} \to \mathbb{R}_{\infty}$  be a functional. The iteration

$$u^{k+1} \in \underset{u \in \mathcal{U}}{\arg\min} \left\{ \frac{1}{2} \| Ku - f \|_{\mathcal{V}}^{2} + D_{J}^{p^{k}}(u, u^{k}) \right\}$$

$$p^{k+1} = p^{k} + K^{*}(f - Ku^{k+1})$$
(1)

with  $u^0 = p^0 = 0$  and  $p^k \in \partial J(u^k)$  for all  $k \in \mathbb{N}$  is known as *Bregman iteration*. In this exercise we will analyse properties of the Bregman iteration which can be seen as an iterative regularisation method.

- a) Show that the Bregman iterates monotonically decrease the data fidelity, i.e. they satisfy  $\|Ku^{k+1} f\|_{\mathcal{V}} \leq \|Ku^k f\|_{\mathcal{V}}$ .
- b) If  $\mathcal{V}$  is a Hilbert space, show that the Bregman iteration can also be written as

$$u^{k+1} \in \underset{u \in \mathcal{U}}{\arg\min} \left\{ \frac{1}{2} \| Ku - f^k \|_{\mathcal{V}}^2 + J(u) \right\}$$
$$f^{k+1} = f^k + f - Ku^{k+1}$$

for  $u^0 = 0, f^0 = f$ .

- c) Show that if there exists a  $k_* \in \mathbb{N}$  such that  $u^{k_*}$  satisfies  $Ku^{k_*} = f$ ,  $u^k = u^{k_*}$  for all  $k \ge k_*$  and  $u^{k_*}$  is a *J*-minimising solution of all elements in the set  $\{u \in \mathcal{U} \mid Ku = f\}$ .
- d) Consider now the noisy case where  $u^k$  are the iterates for noisy data, i.e.  $f^{\delta}$  replaces f in the definition of the iteration and  $||f f^{\delta}|| \leq \delta$  for some  $f \in \mathcal{R}(K)$ . Let  $u^* \in \mathcal{U}$  such that  $Ku^* = f$ . Show that  $u^k$  progressively approximates  $u^*$  in a Bregman sense, i.e.

$$D_J^{p^{k+1}}(u^{\dagger}, u^{k+1}) \le D_J^{p^k}(u^{\dagger}, u^k),$$

as long as  $||Ku^{k+1} - f^{\delta}||_{\mathcal{V}} \ge \delta$ . Thus the Bregman iteration with Morozov's discrepancy principle as a stopping criterion is a useful strategy to find *J*-minimising solutions.

Note that this also implies  $D_J^{p^{k+1}}(u^{\dagger}, u^{k+1}) \leq D_J^{p^k}(u^{\dagger}, u^k)$  for all  $k \in \mathbb{N}$  in case of  $\delta = 0$ .

e) Consider again the noise-free case, i.e.  $\delta = 0$ . Show the estimate

$$D_J^{p^k}(u^{\dagger}, u^k) \le \frac{\|w\|_{\mathcal{V}}^2}{2k}$$

for  $k \geq 1$  on the speed of the convergence under a source condition  $p^{\dagger} := K^* w \in \partial J(u^{\dagger})$  for some  $u^{\dagger} \in \mathcal{U}$  with  $Ku^{\dagger} = f$ .