## Advanced Topics in Control 2020: Large-Scale Convex Optimization

Exercise 7: Operator Splitting Methods

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Please submit your solutions via Moodle as a PDF with filename Ex07\_Surname.pdf, replacing Surname with your surname.

## Problem 1

Compute the proximal mapping  $\mathbf{prox}_f(x)$  of the following functions:

- a) Convex quadratic function:  $f(x) = \frac{1}{2}x^T A x + b^T x + c, A \in \mathbb{S}^n_+, b \in \mathbb{R}^n;$
- b) Indicator function of the Euclidean ball:

$$f(x) = \mathcal{I}_C(x), C = \{x \in \mathbb{R}^n : ||x - c||_2 \le r\}, c \in \mathbb{R}^n, r \in \mathbb{R}_+;$$

Hint: You can derive it geometrically.

c) Indicator function of an affine set:

$$f(x) = \mathcal{I}_C(x), C = \{x \in \mathbb{R}^n : Px = q\}, P \in \mathbb{R}^{m \times n} \text{ has full row rank}, q \in \mathbb{R}^m, m \leq n;$$

- d) One-dimensional:  $f(x) = \lambda |x|, x \in \mathbb{R}, \lambda > 0$ ; *Hint:* You can use the fact that  $\mathbf{prox}_{\lambda f} = (\mathrm{Id} + \lambda \partial f)^{-1}(x)$  and the graphical arguments.
- e)  $\ell_1$ -norm:  $f(x) = ||x||_1 = \sum_{i=1}^n |x_i|$ ; Hint: (it was shown already in the lecture using the Moreau's identity) Now show it using a separability property of the proximal operators.
- f)  $\ell_2$ -norm:  $f(x) = ||x||_2 = \sqrt{\sum_{i=1}^n x_i^2}$ . Hint: Use Moreau's identity to connect it to one of the previous subtasks. You can use the result that you derived in the Exercise 4.b) of the Exercise set 2.
- g) Under which condition on matrix A in the subtask 1.a) is the  $\mathbf{prox}_f$  operator contractive? Derive the contraction factor.

## Problem 2

- a) Let  $f: \mathbb{R}^n \to \mathbb{R} \cup \{+\infty\}$  be proper, closed, and convex. Prove that
  - 1.  $u = \mathbf{prox}_f(x)$  if and only if  $x u \in \partial f(u)$ .
  - 2. x is a minimizer of f if and only if  $x = \mathbf{prox}_f(x)$ .

Hint: Use Fermat's rule. Prove the second statement using the first one.

b) Let f be a proper, closed and convex. Show that the  $\mathbf{prox}_f$  operator is nonexpansive, i.e., that for any  $x, y \in \mathbb{R}^n$ :  $\|\mathbf{prox}_f(x) - \mathbf{prox}_f(y)\| \le \|x - y\|$ .

Hint: Let  $\mathbf{prox}_f(x) = u$ ,  $\mathbf{prox}_f(y) = v$  and apply for them the result of previous subtask 2.a)1. Then, use the subgradient inequalities, sum them together and derive that  $\langle x - y, u - v \rangle \ge \|u - v\|^2$ . Next, considering separately the cases when u = v and  $u \ne v$ , and using the Cauchy-Schwarz inequality, derive the result.

## Problem 3

Consider the problem

$$\min_{x \in \mathbb{R}^n} f(x) + g(x) = \min_{x \in \mathbb{R}^n} ||Ax - b||_2^2 + ||x||_1,$$

where  $A \in \mathbb{R}^{m \times n}$  and  $b \in \mathbb{R}^m$ . For this problem, write down the iteration of

- a) Douglas-Rachford splitting;
- b) Forward-backward splitting.