

### Problem 9.1: ADMM and Scaled Form ADMM

In this exercise, we derive a scaled form for the Alternating Direction Method of Multipliers (ADMM). Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  and  $g : \mathbb{R}^m \rightarrow \mathbb{R}$  be convex functions. Consider the following optimization problem

$$\min_{x,z} f(x) + g(z) \quad (1)$$

$$\text{subject to: } Ax + Bz = c \quad (2)$$

with variables  $x \in \mathbb{R}^n$  and  $z \in \mathbb{R}^m$ . Assume that the problem data is  $A \in \mathbb{R}^{p \times n}$ ,  $B \in \mathbb{R}^{p \times m}$ , and  $c \in \mathbb{R}^p$ . Notice that the objective function has two independent sets of variables  $x$  and  $z$ . Let us define the augmented Lagrangian of (1) – (2) as

$$L_\rho(x, z, y) = f(x) + g(z) + y^\top (Ax + Bz - c) + \frac{\rho}{2} \|Ax + Bz - c\|_2^2. \quad (3)$$

with dual variables  $y \in \mathbb{R}^p$  and penalty parameter  $\rho > 0$ . The augmented Lagrangian (3) can be seen as the (unaugmented) Lagrangian of the problem

$$\min_{x,z} f(x) + g(z) + \frac{\rho}{2} \|Ax + Bz - c\|_2^2 \quad (4)$$

$$\text{subject to: } Ax + Bz = c \quad (5)$$

The problem (4) – (5) is equivalent to the problem (1) – (2): for any feasible solution  $(x, z)$ , the additional term in the objective (4) evaluates to zero. Solving the augmented Lagrangian (3) by ADMM consists of the following iterations

$$x^{k+1} = \operatorname{argmin}_x L_\rho(x, z^k, y^k) \quad (6)$$

$$z^{k+1} = \operatorname{argmin}_z L_\rho(x^{k+1}, z, y^k) \quad (7)$$

$$y^{k+1} = y^k + \rho(Ax^{k+1} + Bz^{k+1} - c) \quad (8)$$

- (a) Motivate a suitable stopping criterion for the ADMM iterations (6) – (8).
- (b) Derive the *scaled form* for the ADMM iterations (6) – (8) by defining the *primal residual*  $r$  and the *scaled dual variables*  $u$  as

$$r = Ax + Bz - c \quad \text{and} \quad u = \frac{y}{\rho} \quad (9)$$

*Hint:* Apply the definitions of  $r$  and  $u$  to (3) and rewrite the ADMM iterations (6) – (8) by replacing the original dual variables  $y$  by their scaled counterparts  $u$ .

### Problem 9.2: ADMM for Quadratic Optimization Problems

Consider the following standard form quadratic optimization problem

$$\min_x \frac{1}{2} x^\top P x + q^\top x \quad (10)$$

$$\text{subject to: } Ax = b \quad (11)$$

$$x \geq 0 \quad (12)$$

with variables  $x \in \mathbb{R}^n$ . Assume that  $P \in \mathbb{R}^{n \times n}$  is a symmetric positive definite matrix,  $q \in \mathbb{R}^n$ ,  $A \in \mathbb{R}^{p \times n}$ , and  $b \in \mathbb{R}^p$ . We can express the problem (10) – (12) in ADMM form as

$$\min_{x,z} f(x) + g(z) \quad (13)$$

$$\text{subject to: } x = z \quad (14)$$

where

$$f(x) = \frac{1}{2} x^\top P x + q^\top x \text{ with } \mathbf{dom} f = \{x \in \mathbb{R}^n : Ax = b\}$$

is the original objective with a restricted domain, and  $g : \mathbb{R}^n \rightarrow \{0, \infty\}$  is the indicator function of the nonnegative orthant  $\mathbb{R}_+^n$  corresponding to the constraint  $x \geq 0$ . Write the augmented Lagrangian for (13) – (14) using the scaled dual variables, and write the corresponding scaled form ADMM iterations using the results of Exercise 9.1.