

### Problem 7.1: KKT Conditions for Equality Constrained Problems

Let  $X \subset \mathbb{R}^n$  be a nonempty open set, and let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable. Moreover, let  $g_i : \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable for all  $i = 1, \dots, m$ , and let  $h_i : \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable for all  $i = 1, \dots, l$ . Consider the following optimization problem  $P$ :

$$\begin{aligned} (P) : \quad & \min. f(x) \\ & \text{subject to: } g_i(x) \leq 0, & i = 1, \dots, m \\ & h_i(x) = 0, & i = 1, \dots, l \\ & x \in X \end{aligned}$$

Let  $\bar{x}$  be a feasible solution to  $P$ , and let  $I = \{i : g_i(\bar{x}) = 0\}$  be the index set of *active* inequality constraints. Also, let  $\nabla g_i(\bar{x})$  for  $i \in I$  and  $\nabla h_i(\bar{x})$  for  $i = 1, \dots, l$  be linearly independent (to enforce constraint qualification). Derive KKT conditions for the problem  $P$ .

*Hint:* Notice that  $h_i(x) = 0$  can be equivalently replaced by the two inequalities

$$h_i(x) \leq 0 \text{ and } -h_i(x) \leq 0.$$

### Problem 7.2: KKT Transformation of a Bilevel Optimization Problem

Consider the following *bilevel* optimization problem:

$$\begin{aligned} & \min_x c_1^\top x + c_2^\top y & (1) \\ & \text{subject to: } Ax + By \leq \alpha & (2) \\ & y \in \operatorname{argmin}_y c_3^\top y & (3) \\ & \text{subject to: } Dx + Ey \leq \beta & (4) \end{aligned}$$

In problem (1) – (4), we seek an optimal value of  $x$  knowing that  $y$ , which minimizes another optimization problem, depends on the value of  $x$ . This is a way of modeling hierarchical decision problems such as [Stackelberg competition](#).

Reformulate the problem (1) – (4) by replacing the constraints (3) – (4) that form the inner optimization problem:

$$\begin{aligned} & y \in \operatorname{argmin}_y c_3^\top y \\ & \text{subject to: } Dx + Ey \leq \beta \end{aligned}$$

with the KKT optimality conditions of this problem. You can assume that  $\beta \in \mathbb{R}^m$ , which implies that (4) has  $i = 1, \dots, m$  inequality constraints. Is the resulting problem convex? Justify your answer.

*Hint:* You can write the constraint (4) as

$$d_i x + e_i y \leq \beta_i, \quad i = 1, \dots, m$$

where  $d_i$  and  $e_i$  correspond to the  $i$ th rows of the matrices  $D$  and  $E$ , respectively, and  $\beta_i$  is the  $i$ th element of the vector  $\beta \in \mathbb{R}^m$ .

### Problem 7.3: Example of a Bilevel Transformation

Consider the following bilevel optimization problem:

$$\min_x x - 4y \tag{5}$$

$$\text{subject to: } x \geq 0 \tag{6}$$

$$y \in \underset{y}{\operatorname{argmin}} y \tag{7}$$

$$\text{subject to: } -x - y \leq -3 \tag{8}$$

$$-2x + y \leq 0 \tag{9}$$

$$2x + y \leq 12 \tag{10}$$

$$-3x + 2y \leq -4 \tag{11}$$

$$y \geq 0 \tag{12}$$

Reformulate the problem (5) – (12) by replacing the constraints (7) – (12) that form the inner optimization problem:

$$y \in \underset{y}{\operatorname{argmin}} y$$

$$\text{subject to: } -x - y \leq -3$$

$$-2x + y \leq 0$$

$$2x + y \leq 12$$

$$-3x + 2y \leq -4$$

$$y \geq 0$$

with the KKT conditions of this problem. Try to model and solve the reformulated problem with **Julia** using **JuMP**. One locally optimal solution for the problem is  $(x, y) = (2, 1)$  with objective value  $f(x, y) = x - 4y = -2$ . Can you find this local optimum by trying different initial (starting) values for the different variables? Is the reformulated problem convex? Justify your answer.