Model Predictive Control Prof. Manfred Morari

ESE6190, Spring 2023 Due: Feb 13

Exercise sheet 2 Optimization

Instructions: You are not allowed to use a calculator / computer unless specified.

Exercise 1 Properties of Sets and Functions

Let $x_1, x_2 \in \mathbb{R}^n$ be in the feasible set X of an optimization problem, i.e.

$$x_1, x_2 \in X = \{x \in \mathbb{R}^n \mid g_i(x) \leq 0, i = 1, \dots, m, h_i(x) = 0, j = 1, \dots, q\}$$

- 1. Can you find simple conditions on functions $g_i(x)$, $i=1,\ldots m$, and $h_j(x)$, $j=1,\ldots q$, such that the feasible set X is convex? [2 pts]
- 2. Let $z = \theta x_1 + (1 \theta)x_2$, $\theta \in [0, 1]$, be any point on the line segment between points x_1 and x_2 . Can you find conditions on function f(x) such that 'z is better than the worst of x_1 and x_2 ', i.e.

$$f(z) \le \max\{f(x_1), f(x_2)\}$$
 ?

If yes, can it happen that point z is better than both of them? Try to sketch such a situation. [5 pts]

3. Which property of the feasible set X would ensure that point $y = x_1 + x_2$ is feasible, given $x_1, x_2 \in X$? [3 pts]

Exercise 2 Checking Convexity of Sets

Which of the following sets are convex? Give reasons for your answers.

- 1. A slab, i.e. the set $\{x \in \mathbb{R}^n \mid \alpha \leq a^T x \leq \beta\}$ where $a \in \mathbb{R}^n$ and $\alpha, \beta \in \mathbb{R}$? Hint: Try to sketch this set for n = 2! [3 pts]
- 2. Let $s : \mathbb{R}^n \to \mathbb{R}$ be any function with domain dom $s \subseteq \mathbb{R}^n$. Let set S be any subset of dom s, though not necessarily a convex one. Is the set M defined as

$$M := \{x \mid ||x - y|| \le s(y) \text{ for all } y \in S\}$$

a convex set? Note that $\|.\|$ denotes any norm on \mathbb{R}^n . [4 pts]

3. Consider s and S as above, again $\|.\|$ denotes any norm on \mathbb{R}^n . We consider the set \tilde{M} defined as

$$\tilde{M} := \{ x \mid \exists y \in S \text{ s.t. } ||x - y|| \le s(y) \}$$

Is \tilde{M} a convex set, assuming that s is a convex function? [4 pts]

4. The set of points closer to a given point x_0 than to a given set $Q \subseteq \mathbb{R}^n$, i.e.

$$\{x \mid ||x - x_0||_2 < ||x - y||_2 \text{ for all } y \in Q\}$$
,

where $\|.\|_2$ denotes the standard 2-norm (also called *Euclidean* norm)? [4 pts]

Exercise 3 1-Norm, ∞ -Norm

Instead of the standard 2-norm, the 1-norm or the ∞ -norm are sometimes used in the MPC cost function. In this exercise you should show that both a minimization problem with a 1-norm objective and an ∞ -norm objective

$$\min_{x} \|Ax\|_{p}, \qquad p \in \{1, \infty\}$$

can be recast as a linear program (LP)

$$\min_{y} b^{T} y$$

s.t. $Fy \leq g$,

with vectors b, g and matrix F defined appropriately. [10 pts]

Note that we assume matrix $A \in \mathbb{R}^{N \times n}$ where its N row vectors are denoted as a_i^T , i = 1, ..., N, i.e.

$$A = \begin{bmatrix} a_1^T \\ \vdots \\ a_N^T \end{bmatrix}.$$

Exercise 4 Linear Regression with 1- and $\infty-$ Norms

We now use the ideas from Exercise 3 to solve a linear regression problem: Use the command linprog from MATLAB to solve the LP's stemming from a reformulation of the linear regression problems

$$\min_{y} \|Ax - b\|_{p}, \qquad p \in \{1, \infty\} ,$$

where A is a random matrix and b is a random vector, e.g. created in MATLAB with the commands A=rand(10,5) and b=rand(10,1). [10 pts]

Note: What we are actually doing here is to solve the overdetermined linear equation system Ax = b. The solution for the case p = 2 is called the linear least square solution and is given directly in MATLAB by $A \setminus b$.

Optional: Try to install e.g. Yalmip (http://users.isy.liu.se/johanl/yalmip/) which helps you to model optimization problems in MATLAB in a more convenient way than before. Also, have a look at different LP solvers (under 'Solvers' on the latter website) that are more powerful than MATLAB's linprog.

Exercise 5 Quadratic Program

Consider the optimization problem:

min
$$\frac{1}{2}(x_1^2 + x_2^2 + 0.1x_3^2) + 0.55 x_3$$
 subject to
$$x_1 + x_2 + x_3 = 1$$

$$x_1 \ge 0$$

$$x_2 \ge 0$$

$$x_3 \ge 0$$

- 1. Show that $x^* = (0.5, 0.5, 0)$ is a local minimum. [5 pts]
- 2. Is x^* also a global minimum? Explain why, or why not. [5 pts]