Nonlinear and Discrete Optimization—Homework 8

1. Consider the discrete optimization problem

$$\begin{split} \min_{x \in \mathbb{R}^n} & x^T Q x \\ \text{s.t.} & x_i^2 = 1, \quad i = 1, ..., n \end{split}$$

where Q is an $n \times n$ symmetric matrix. To study this problem, we take two approaches:

- Brute-force search: We compute the objective function for all of the 2^n possible feasible points and denote the smallest objective value as p_* .
- *Dual optimization:* We solve the convex dual problem and denote the optimal objective value as d_{*}.

The goal of this example is to study the duality gap and to analyze the computational complexities of the above methods.

- i) Consider n = 15. Generate 150 random symmetric matrices Q such that each diagonal entry is equal to 25 and each off-diagonal entry is uniformly chosen from the interval [0, 1]. Write a code in CVX that finds the duality gap (in percentage) based on the formula $\frac{p_* d_*}{|p_*|} \times 100$ for each random initialization of Q, and report the average of this gap over the 150 trials.
- ii) For n = 1, 2, ..., 28, generate a random $n \times n$ matrix Q (based on the procedure stated in Part (i)) and then find the corresponding numbers p_* and d_* . Let $t_p(n)$ and $t_d(n)$ denote the runtimes of your codes for finding p_* and d_* , respectively. Draw $t_p(n)$ and $t_d(n)$ as a function of n, and based on these curves discuss the benefits of using the dual optimization over the brute-force search.
- 2. Use the branch-and-bound method to solve the mixed-integer program:

$$\min_{x \in \mathbb{R}^2} \quad -3x_1 - x_2$$

s.t.
$$2x_1 - x_2 \le 6$$
$$x_1 + x_2 \le 4$$
$$x_1, x_2 \ge 0$$
$$x_1 : \text{ integer}$$

(note: you can use CVX to solve the convex problems needed to form the tree).

3. Use the branch-and-bound method to solve the discrete optimization problem:

$$\min_{x \in \mathbb{R}^3} \quad -4x_1 - 3x_2 - x_3 \\ \text{s.t.} \quad 3x_1 + 2x_2 + x_3 \le 7 \\ 2x_1 + x_2 + 2x_3 \le 11 \\ x_1, x_2, x_3 \ge 0 \\ x_1, x_2, x_3 : \text{integer}$$

(note: you can use CVX to solve the convex problems needed to form the tree).