
Efficient Algorithms and Data Structures II

Deadline: None, Tutorial exercises only.

Tutorial Exercise 1

A closed convex set is called line-free (geradenfrei) if it does not contain any (straight) line.

- (a) Let P be the feasible region of an LP in standard form. Show that P is line free.
- (b) Let P be the feasible region of an LP in the form $\max c^T x$ s.t. $Ax \leq b$ (natural form). Show that P might not be line free.
- (c) Show that Theorem 21 does not hold if P is in natural form.

Tutorial Exercise 2

Consider the following LP:

$$\begin{array}{rllll} \max. & 2x_1 & - & 3x_2 & + & 3x_3 \\ \text{s.t.} & x_1 & + & x_2 & - & x_3 & \leq & 7 \\ & - & x_1 & - & x_2 & + & x_3 & \leq & -7 \\ & & x_1 & - & 2x_2 & + & 2x_3 & \leq & 4 \\ & & & & x_1, x_2, x_3 & & & \geq & 0 . \end{array}$$

- (a) Rewrite the LP in standard form and then in matrix form.
 - (b) Give three feasible solutions and their objective value.
- (Extra) Solve the LP using the Simplex algorithm shown in the lecture.

Tutorial Exercise 3

Formulate the minimum spanning tree problem on an edge-weighted graph $G = (V, E, c)$ as a integer linear program with $|E|$ variables, one for each edge.

Tutorial Exercise 4

Consider the data set found under <http://www14.in.tum.de/lehre/2019SS/ea/uebung/loesungen/data/ex1-4-data.csv> of students in some TUM course. Any student that passed the final exam has PASSED=1 (positive instance), any student that failed the exam has FAILED=-1 (negative instance). We want to determine a hyperplane that separates the students that failed from those that passed.

In order to do so, we want to construct a linear function h such that

$$h(p^i) \geq 1$$

for all positive instances p^i and

$$h(q^i) \leq -1$$

for all negative instances q^i . The *hinge loss* incurred by h is the extend to which the above inequalities fail to hold. For a positive instance p^i this is $\max\{1 - h(p^i), 0\}$, for a negative instance q^i this is $\max\{1 + h(q^i), 0\}$.

The linear function h is to be chosen so that the total hinge loss is minimal. In order to do so, we solve the following linear program:

$$\begin{array}{ll} \min. & \sum_i e_i \\ \text{s.t.} & 1 - \left(\sum_{j=1}^d a_j p_j^i + b \right) \leq e_i \quad \text{for every positive instance } p^i \\ & 1 + \left(\sum_{j=1}^d a_j q_j^i + b \right) \leq e_i \quad \text{for every negative instance } q^i \\ & e_i \geq 0 . \end{array}$$

Solve the given linear program using a method of your choice. It is maybe easiest to use Python Pulp <https://pythonhosted.org/PuLP/index.html>.

The success of solving linear programming therefore depends on a number of factors: (1) the power of computers, (2) extremely clever algorithms; but it depends most of all upon (3) a lot of good luck that the matrices of practical problems will be very very sparse and that their bases, after rearrangement, will be nearly triangular.

- G. B. Dantzig