Please answer four out of the following five questions, on a separate paper. Note that there are two pages in the exam. No books. *Calculators are allowed*.

A Collection of Formulas will be handed. Kaavakokoelma jaetaan.

1. Consider the optimization problem:

$$\min f(x,y) = x^2 + y^3$$

s.t.

$$x^2 + y^2 \le 1$$

- (a) Is the feasible set convex, and is the function convex? (Prove or disprove)
- (b) Find the KKT-points of the problem. Is one of them a minimum?
- 2. Consider the linear problem:

$$\min z = x_1 + 2x_2 - x_3$$

s.t.

$$x_1 - x_2 - x_3 \ge 2$$
$$2x_1 + x_2 + 3x_3 \le 1$$

$$x_1, x_2, x_3 \ge 0$$

- (a) Transform the problem into the standard format
- (b) Find the dual of the problem (either directly or from the standard form; your choice)
- 3. Assume that we have a linear problem:

$$\min c^T x$$

s.t.

$$Ax = b$$

- (a) Suppose we construct the dual and find that the dual is infeasible. What can we say about the original problem?
- (b) Suppose that we wish to use the Simplex algorithm. To do so, we need one basic feasible solution to the LP. Explain a method for finding a basic feasible solution.
- (c) Suppose we run the dual simplex algorithm and find a point u that is a feasible solution to the dual. What can we say about possible solutions of the original problem?

4. Consider the unconstrained problem

$$\min f(x, y) = x - y + 2xy + 2x^2 + y^2$$

- (a) Suppose we run a minimization algorithm to find a local minimum, starting from the point $x_0 = (1, 5)$. Pick any of the methods discussed during the course to find a minimum. (I suggest the Newton method, but feel free to use any of the others), and compute at least two other points from the iteration (i.e. x_1 and x_2).
- (b) How would you determine, after reaching a result, whether the point you found is a local minimum?
- (c) How would you determine if it is a global minimum?
- 5. Let $f: X \to \mathbb{R}$ be convex. Which of the following claims hold? Prove or disprove (i.e., give an example where the claim is false)
 - (a) The function $f(x)^2$ is convex.
 - (b) The set $f(x) \leq b$ is convex for every $b \in \mathbb{R}$.
 - (c) If $K \subseteq X$ is a closed and bounded convex set and f is differentiable in K, then

$$\min_{x \in K} f(x)$$

has the solution x^* such that $\nabla f(x^*) = 0$.