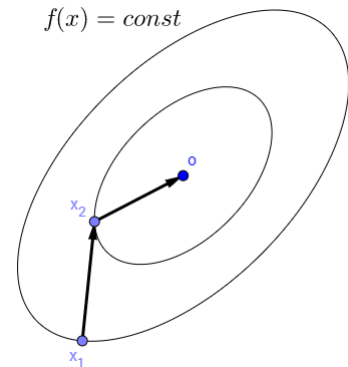


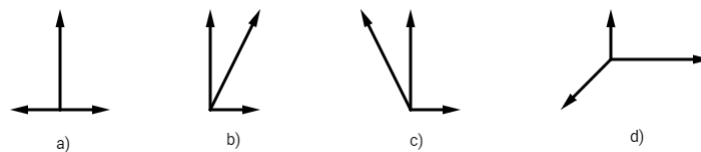
Students may use the formulae sheet and a pocket calculator. All solutions must be properly justified.

1. a) Two consecutive iterations of a minimization algorithm are shown in the picture together with some level sets of a positive definite quadratic form  $f(x) = x^T Hx$ ,  $x \in \mathbb{R}^2$ . Explain for each method below why it could or could not be the one used (with exact line search): (0.6)



- Steepest descent method,
- Conjugate directions method,
- Newton's method,
- Quasi-Newton method.

- b) Assume that the vectors below are gradients of active inequality constraints at a point in the plane. In what case(s) is the point a CQ point? (0.4)



2. Consider the function  $f(x, y) = x^3 - 3xy + y^3$  on the set  $S = \{(x, y) : x \geq 0, xy \geq 1\}$ .
- a) Show that  $f$  is convex on  $S$ . (0.3)
- b) Show that the minimum of  $f$  on  $S$  exists. (0.3)
- c) Construct a differentiable penalty function for the problem  $\min_{x \in S} f(x)$  and find the first Newton direction starting at  $(1, 0)$  with  $\mu = 3$ . (0.4)

3. a) Consider the following LP problem

$$\min (9x_1 + 6x_2 + 8x_3 - 2x_4) \quad \text{subject to} \quad \begin{cases} 3x_1 + x_2 - x_3 - 2x_4 \geq 1, \\ x_1 - x_2 - 2x_3 + x_4 \leq 1, \\ \text{all } x_k \geq 0. \end{cases}$$

Construct the dual problem, solve it graphically and use the solution to solve the primal problem. (0.6)

Please, turn over

- b) Let vectors  $b$ ,  $c$  and a matrix  $A$  (of suitable dimensions) be given and consider two LP problems

$$P_1 : \min c^T x \text{ subject to } \begin{cases} Ax \geq b \\ x \geq 0 \end{cases} \quad \text{and} \quad P_2 : \min c^T x \text{ subject to } \begin{cases} \begin{bmatrix} A \\ I \end{bmatrix} x \geq \begin{bmatrix} b \\ 0 \end{bmatrix} \\ x \text{ is free.} \end{cases}$$

They are the same problem, but in  $P_2$  the positivity constraints are considered to be a part of inequalities within the larger matrix  $\begin{bmatrix} A \\ I \end{bmatrix}$ . Formulate the dual problems to  $P_1$  and  $P_2$  and show that the dual problems are equivalent. (0.4)

4. a) Let  $A$  be a  $m \times n$  matrix and  $b \in \mathbb{R}^m$ . Prove that  $h(x) = Ax + b$  is convex. (0.2)

- b) Is the set

$$\Omega = \{(x, y, z) : \frac{x^2}{y} + \frac{y^2}{z} + \frac{z^2}{x} \leq 1, x > 0, y > 0, z > 0\}$$

convex? (0.4)

- c) Let  $g: \mathbb{R}^n \rightarrow \mathbb{R}$  be a convex function. Show that the penalty function

$$\alpha(x) = \max \{g(x), 0\}^2$$

is convex too. (0.4)

5. Find the (minimal) distance (if it exists) from the origin to the set

$$S = \left\{ (x, y, z) : x > 0, y > 0, z > 0, \frac{1}{xy} + \frac{1}{yz} \leq 1 \right\}.$$

6. a) Solve the optimization problem

$$\min(x^2 + 2xy + 2y^2) \text{ subject to } 2y + x^2 \geq 1, x \geq 0, y \geq 0$$

by the duality method with  $X = \{(x, y) : x \geq 0, y \geq 0\}$ . (0.5)

(Hint: do minimization in  $x$  first.)

- b) Consider a primal problem

$$\min f(x) \text{ subject to } g(x) \leq 0, h(x) = 0$$

where  $g$  and  $h$  are vector valued functions. Prove that if there exist a primal feasible  $\bar{x}$  and a dual feasible  $(\bar{u}, \bar{v})$  such that

$$f(\bar{x}) = \Theta(\bar{u}, \bar{v})$$

then  $\bar{x}$  is the global minimum for the primal problem. (0.5)

**GOOD LUCK!**