

# Theoretical material for the final test of Introduction to Optimization

Tangent and normal cones to sets (definitions)

Proposition 2.21 - with proof

Theorem of Ljusternik (Th. 2.23) - without proof

Proof of the easy part of the Ljusternik theorem (Lemma 2.22)

Farkas lemma (Lemma 2.24) - with proof

The KKT Theorem 2.25 - with the proof in the normal case only

Slater condition and the KKT Theorem 3.8 (without proof)

Saddle points and Theorem 3.14 (with proof)

Dual Problem to a general non-linear one and Theorem 3.18 (with proof)

Formulation of the dual problem to a GLP

Statements (A1) - (A3) of Theorem 4.4 (without proof)

Proof of implication (B4)  $\implies$  (B1) in Theorem 4.4

Definitions of extreme point and basis point of polyhedral sets

Theorem 4.11 (prove that existence of a feasible point implies existence of a basis point)



# Samples of the test on Optimization, WS 2017/18

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**Theoretical questions.** One theoretical question will be given, similar to the following ones.

1. (Part of the Ljusternik theorem.) Let

$$K := \{x \in \mathbf{R}^n : g(x) = 0, h(x) \leq 0\},$$

where  $g : \mathbf{R}^n \mapsto \mathbf{R}^m$ ,  $h : \mathbf{R}^n \mapsto \mathbf{R}^r$  are differentiable at  $x \in K$ . Prove that

$$T_K(x) \subset \{l \in \mathbf{R}^n : \partial g(x)l = 0, \partial h(x)|_{J(x)}l \leq 0\}.$$


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2. (Part of the Farkas lemma.) Let

$$P = \{x \in \mathbf{R}^n : Gx = 0, Hx \leq 0\},$$

where  $G$  is an  $(m \times n)$ -matrix and  $H$  is an  $(r \times n)$ -matrix. Prove that the polar cone  $P^\circ$  satisfies the inclusion

$$P^\circ \subset \{G'\lambda + H'\mu : \lambda \in \mathbf{R}^m, \mu \in \mathbf{R}^r, \mu \geq 0\}.$$


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3. Formulate the KKT theorem for general differentiable (not necessarily convex) functions. Is the normal form of the theorem (with  $\lambda_0 \neq 0$ ) true for the example

$$\min\{x_1\}$$

subject to

$$-(x_1)^3 + x_2 \leq 0$$

$$-x_2 \leq 0.$$

**Solution.**

The problem is

$$\min_{x \in K} f(x), \quad (1)$$

where

$$K = \{x \in \mathbf{R}^n : g(x) = 0, h(x) \leq 0\}, \quad (2)$$

and  $g : \mathbf{R}^n \mapsto \mathbf{R}^m$ ,  $h : \mathbf{R}^n \mapsto \mathbf{R}^r$ .

Define  $J(x) = \{j : h_j(x) = 0\}$ .

**Theorem.** *Let  $x^* \in K$  be a local solution of problem (1), (2). Assume that the functions  $f$ ,  $g$  and  $h$  are continuously differentiable around  $x^*$ .*

*Then*

(i) *there exists a non-zero vector  $(\lambda_0, \lambda, \mu) \in \mathbf{R} \times \mathbf{R}^m \times \mathbf{R}^r$  such that*

$$\lambda_0 \partial f(x^*) + \lambda' \partial g(x^*) + \mu' \partial h(x^*) = 0, \quad (3)$$

$$\mu_j h_j(x^*) = 0, \quad j = 1, \dots, r, \quad (4)$$

$$\mu \geq 0, \quad \lambda_0 \geq 0. \quad (5)$$

(ii) *If, in addition, the matrix*

$$\begin{pmatrix} \partial g(x^*) \\ \partial h(x^*)|_{J(x^*)} \end{pmatrix}$$

*is surjective (that is, its rank is  $m + |J(x^*)|$ , where  $|J(x^*)|$  is the number of the elements of  $J(x^*)$ ), then claim (i) of the theorem is true with  $\lambda_0 = 1$ .*

Answer to the additional question.

The point  $(0, 0)'$  satisfies the constraints. For any other admissible point  $(x_1, x_2)'$  we have  $0 \leq x_2 \leq (x_1)^3$ , hence  $x_1 \geq 0$ . Then  $(0, 0)'$  is a solution of the problem. Equation (3) reads in this case as

$$\lambda_0 \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \mu_1 \begin{pmatrix} 0 \\ 1 \end{pmatrix} + \mu_2 \begin{pmatrix} 0 \\ -1 \end{pmatrix} = 0,$$

which is possible only if  $\lambda_0 = 0$ . Thus the normal form of the KKT does not hold in this case.

4. Formulate the KKT theorem for convex problems of the type

$$\min_{K \cap K_0} f(x) \quad (6)$$

where  $K \subset \mathbf{R}^n$  is defined by the constraints

$$G_i x - b_i = 0, \quad i = 1, \dots, m, \quad (7)$$

$$H_j x - c_j \leq 0, \quad j = 1, \dots, r', \quad (8)$$

$$h_j(x) \leq 0, \quad j = r' + 1, \dots, r, \quad (9)$$

$0 \leq r' \leq r$ , and  $K_0$  is a polyhedral set.

Is the Slater condition fulfilled for the following sets (**the sets may be changed in the test!**)

$$K := \{x \in \mathbf{R}^3 : x_1 - x_3 = 0, x_2 - x_3 \leq -1, (x_1)^2 + (x_2)^2 \leq 1\},$$

$$K_0 := \{x \in \mathbf{R}^3 : x_2 \geq 0\}.$$

**Solution.**

**Theorem.** Assume that the functions  $f$  and  $h_j$  are convex and that there exists  $\tilde{x} \in K_0$  satisfying (7) and (8), and also satisfying (9) as strict inequalities (the Slater condition).

Then  $x^*$  is an optimal solution if and only if  $x^* \in K \cap K_0$  and there exist  $\lambda \in \mathbf{R}^m$  and  $\mu \in \mathbf{R}_+^r$  such that

$$L(x^*, \lambda, \mu) \leq L(x, \lambda, \mu) \quad \forall x \in K_0, \quad \text{and} \quad \mu' h(x^*) = 0, \quad (10)$$

where  $L(x, \lambda, \mu) = f(x) + \lambda' g(x) + \mu' h(x)$ .

The example: If  $x \in K \cap K_0$ , then  $x_3 = x_1$ , hence  $x_2 - x_1 \leq -1$ . Since  $x_2 \geq 0$  we have  $x_1 \geq 1 + x_2 \geq 1$  and from  $(x_1)^2 + (x_2)^2 \leq 1$  we obtain that  $x_1 = 1$  and  $x_2 = 0$ . This is the only point in  $K \cap K_0$  and the non-linear constraint is satisfied as an equality. Thus the Slater condition is not fulfilled. (This conclusion is also evident from geometric representation on the  $(x_1, x_2)$ -plane.)

*Remark.* If the non-linear constraint were  $(x_1)^2 + (x_2)^2 \leq 1.0001$ , the Slater condition would be fulfilled with  $\tilde{x} = (1, 0, 1)$ , for example.

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5. Formulate the dual problem to the following linear optimization problem

$$\min \langle c, x \rangle$$

$$A_i x = b_i, \quad i = 1, \dots, m,$$

$$A_i x \geq b_i, \quad i = m + 1, \dots, m + r,$$

where  $c \in \mathbf{R}^n$ ,  $A_i$  are  $n$ -dimensional row-vectors and  $b_i$  are real numbers.

Formulate the existence claims of the the duality theorem 4.18.

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6. Formulate the dual problem to the following linear optimization problem

$$\min \langle c, x \rangle$$

$$A_i x \geq b_i, \quad i = 1, \dots, r,$$

$$x_k \geq 0, \quad k = 1, \dots, l,$$

where  $c \in \mathbf{R}^n$ ,  $A_i$  are  $n$ -dimensional row-vectors and  $b_i$  are real numbers.

One of the claims of the duality theorem 4.4 applied to this problem is:

(B4)  $x^*$  and  $y^*$  are feasible points for GLP and GLD, respectively, and the complementary slackness conditions hold:

$$y_j^* (A_j x^* - b_j) = 0, \quad j = 1, \dots, r, \quad x_k^* ([A']_k y^* - c_k) = 0, \quad k = 1, \dots, l.$$

Prove that it implies the following one: (B3)  $(x^*, y^*)$  is a saddle point of the Lagrange function in  $K_0 \times Y_0$ ;

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7. Formulate and prove the general duality theorem (Theorem 3.18).

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8. Prove the following statement (a part of Theorem 4.11 in the script):

**Theorem.** *Let  $A$  be a non-zero matrix. If the linear problem*

$$\min_{x \in K} \{c'x\}, \quad K = \{x \in \mathbf{R}^n : Ax = b, x \geq 0\},$$

*has a feasible point (that is,  $K \neq \emptyset$ ), then it has a basis point.*

**Samples of particular problems.** Three problems similar to the following ones will be given in addition to the theoretical task. (The problems will be individual for every student!)

**Problem 1.** Solve the problem

$$\min\{2x_1 + x_2 + 2x_3 + x_4\}$$

subject to

$$\begin{aligned} x_1 + x_2 + 5x_3 + x_4 &= 7 \\ 2x_2 + 7x_3 + x_4 &= 11, \\ x_1, x_2, x_3, x_4 &\geq 0. \end{aligned}$$

by using the simplex method.

*Hint:* Use  $x_1$  and one additional variable  $z_5$  in the second equation as initial basis variables in the auxiliary problem for finding an initial basis for the given problem.

**Solution.** The auxiliary problem for finding an initial basis point is

$$\min\{z_5\}$$

subject to

$$\begin{aligned} x_1 + x_2 + 5x_3 + x_4 &= 7 \\ z_5 + 2x_2 + 7x_3 + x_4 &= 11, \\ x_1, x_2, x_3, x_4, z_5 &\geq 0. \end{aligned}$$

We have (with  $c = (0, 0, 0, 0, 1)'$ )

$$\Delta_2 = c_2 - (c_1 \cdot 1 + c_5 \cdot 2) = -2,$$

$$\Delta_3 = c_3 - (c_1 \cdot 5 + c_5 \cdot 7) = -7,$$

$$\Delta_4 = c_4 - (c_1 \cdot 1 + c_5 \cdot 1) = -1.$$

We can chose any of the non-basis variables as a new basis variable. Let us take  $\nu = 3$  (since  $\Delta_2$  is the minimal). Then according to the rule for choosing which basis variable to be removed we have to choose  $\kappa = 1$ , since

$$\frac{7}{5} < \frac{11}{7}.$$

Then we have

$$\begin{array}{rcccccc} x_3 & & + & \frac{1}{5}x_1 & + & & & + & \frac{1}{5}x_4 & = & \frac{7}{5} \\ z_5 & + & 2x_2 & + & 7\left(\frac{7}{5} - \frac{1}{5}x_1 - \frac{1}{5}x_2 - \frac{1}{5}x_4\right) & + & x_4 & = & 11, \end{array}$$

Hence, the adapted to the basis  $(x, z_5)$  representation of the equality constraints is

$$\begin{array}{rcccccc} x_3 & & + & \frac{1}{5}x_1 & + & \frac{1}{5}x_2 & + & \frac{1}{5}x_4 & = & \frac{7}{5} \\ z_5 & - & \frac{7}{5}x_1 & + & \frac{3}{5}x_2 & - & \frac{2}{5}x_4 & = & \frac{6}{5}, \end{array}$$

Then we calculate (still with  $c = (0, 0, 0, 0, 1)'$ )

$$\Delta_1 = c_1 - \left(c_3 \cdot \frac{1}{5} + c_5 \cdot \frac{-7}{5}\right) = \frac{7}{5},$$

$$\Delta_2 = c_2 - \left(c_3 \cdot \frac{1}{5} + c_5 \cdot \frac{3}{5}\right) = -\frac{3}{5},$$

$$\Delta_4 = c_4 - \left(c_3 \cdot \frac{1}{5} + c_5 \cdot \frac{-2}{5}\right) = \frac{2}{5}.$$

According to the rule of the simplex algorithm we have to choose  $\nu = 2$ , and since

$$\frac{7/5}{1/5} > \frac{6/5}{3/5}$$

we have to choose  $\kappa = 5$ . That is,  $z_5$  leaves the basis and  $x_2$  enters the basis.

We express

$$\begin{array}{rcccccc} x_2 & & - & \frac{7}{3}x_1 & - & & & \frac{2}{3}x_4 & + & \frac{5}{3}z_5 & = & 2 \\ x_3 & + & \frac{1}{5}x_1 & + & \frac{1}{5}\left(2 - \frac{7}{3}x_1 - \frac{2}{3}x_4 - \frac{5}{3}z_5\right) & + & \frac{1}{5}x_4 & = & \frac{7}{5}, \end{array}$$

Since all the auxiliary variables have left the basis (this is only  $z_5$  in our problem and we ignore it further) we have found an initial basis for the original problem,  $((x_2, x_3))$ , and the adapted representation, found from the above equations, is

$$\begin{array}{rcccccc} x_2 & & - & \frac{7}{3}x_1 & - & \frac{2}{3}x_4 & = & 2 \\ x_3 & + & \frac{2}{3}x_1 & + & \frac{1}{3}x_4 & = & 1. \end{array}$$



Now we calculate (this time with  $c = (2, 1, 2, 1)'$ )

$$\Delta_1 = c_1 - \left(c_2 \cdot \frac{-7}{3} + c_3 \cdot \frac{2}{3}\right) = 3,$$

$$\Delta_4 = c_4 - \left(c_2 \cdot \frac{-2}{3} + c_3 \cdot \frac{1}{3}\right) = 1.$$

Since  $\Delta_1$  and  $\Delta_4$  are both positive, we have reached an optimal solution, namely,  $x^* = (0, 2, 1, 0)'$ .

**Problem 2.** Describe analytically the tangent and the normal cone to the set  $K \subset \mathbf{R}^3$  defined by the constraints

$$(x_1)^2 + 2(x_3)^2 \leq 3,$$

$$(x_1)^3 + (x_2)^3 = 9$$

at the point  $x^* = (1, 2, 1)' \in K$ .

*Hint:* Check if the assumptions of the Ljusternik theorem are fulfilled and apply it to find  $T_K(x^*)$ . Then apply the Farkas lemma to find  $N_K(x^*)$ .

**Solution.** Here

$$g(x) = (x_1)^3 + (x_2)^3 - 9, \quad h(x) = (x_1)^2 + 2(x_3)^2 - 3.$$

Then

$$\partial g(x^*) = (3(x_1^*)^2, 3(x_2^*)^2, 0) = (3, 12, 0),$$

$$\partial h(x^*) = (2x_1^*, 0, 4x_3^*) = (2, 0, 4).$$

For  $\bar{l} := (4, -1, -3)'$  (this is just one of many possible choices) we have

$$\partial g(x^*)\bar{l} = (3, 12, 0) \begin{pmatrix} 4 \\ -1 \\ -3 \end{pmatrix} = 0, \quad \partial h(x^*)\bar{l} = (2, 0, 4) \begin{pmatrix} 4 \\ -1 \\ -3 \end{pmatrix} = -4 < 0,$$

thus the assumptions of the Ljusternik theorem are fulfilled.

Then

$$\begin{aligned} T_K(x^*) &= \left\{ l \in \mathbf{R}^3 : (3, 12, 0) \begin{pmatrix} l_1 \\ l_2 \\ l_3 \end{pmatrix} = 0, \quad (2, 0, 4) \begin{pmatrix} l_1 \\ l_2 \\ l_3 \end{pmatrix} \leq 0 \right\} \\ &= \{l \in \mathbf{R}^3 : 3l_1 + 12l_2 = 0, 2l_1 + 4l_3 \leq 0\}. \end{aligned}$$

Then we apply the Farkas lemma with  $G = (3, 12, 0)$  and  $H = (2, 0, 4)$  and obtain

$$\begin{aligned} N_K(x^*) &= \left\{ \begin{pmatrix} 3 \\ 12 \\ 0 \end{pmatrix} \lambda + \begin{pmatrix} 2 \\ 0 \\ 4 \end{pmatrix} \mu : \lambda \in \mathbf{R}, \mu \geq 0 \right\} \\ &= \left\{ \begin{pmatrix} 3\lambda + 2\mu \\ 12\lambda \\ 4\mu \end{pmatrix} : \lambda \in \mathbf{R}, \mu \geq 0 \right\}. \end{aligned}$$

**Problem 3.** Consider the problem

$$\min_{x \in K} \{f(x) := x_1 - 2x_2 + 3x_3 - 4x_4\}$$

with

$$K := \{x \in \mathbf{R}^4 : (x_1)^2 + (x_2)^2 + (x_3)^2 + (x_4)^2 \leq 2\}$$

as a primal problem and find the function  $D(\mu)$  of the corresponding dual problem

$$\max_{\mu \geq 0} D(\mu).$$

Then find the solution  $\mu^*$  of the dual problem and evaluate the corresponding  $x^*$  (resulting from the definition of  $D(\mu^*)$ ). Is  $x^*$  a solution of the primal problem?

**Solution.** Here

$$L(x, \mu) = x_1 - 2x_2 + 3x_3 - 4x_4 + \mu ((x_1)^2 + (x_2)^2 + (x_3)^2 + (x_4)^2 - 2). \quad (11)$$

Since by definition  $D(\mu) = \min_{x \in \mathbf{R}^4} L(x, \mu)$  and  $L$  is convex with respect to  $x$  the minimizing  $x$  is determined by the condition  $\partial_x L(x, \mu) = 0$ :

$$x_1 = -\frac{1}{2\mu}, \quad x_2 = \frac{1}{\mu}, \quad x_3 = -\frac{3}{2\mu}, \quad x_4 = \frac{2}{\mu}. \quad (12)$$

Then substituting in (11) we calculate

$$D(\mu) = -\frac{15}{2\mu} - 2\mu.$$

Since  $D(\mu)$  is concave, its maximum on the set  $\mu \geq 0$  is attained either for  $\mu = 0$  (which gives the “bad” value  $D = -\infty$ , which cannot be maximal), or at a point where  $\partial D(\mu) = 0$ :

$$\frac{15}{2\mu^2} - 2 = 0$$

hence (taking into account that  $\mu \geq 0$ )

$$\mu^* = \frac{\sqrt{15}}{2}.$$

We evaluate  $D(\mu^*) = -2\sqrt{15}$ .

From (12) we calculate a “candidate” for a solution of the primal problem:

$$x_1^* = -\frac{1}{\sqrt{15}}, \quad x_2^* = \frac{2}{\sqrt{15}}, \quad x_3^* = -\frac{3}{\sqrt{15}}, \quad x_4^* = \frac{4}{\sqrt{15}},$$

Then we evaluate

$$f(x^*) = \frac{1}{\sqrt{15}}(-1 - 4 - 9 - 16) = -2\sqrt{15} = D(\mu^*),$$

which means that  $x^*$  is an optimal solution due to the general duality theorem (Theorem 3.12 in the script).

**Problem 4.** Write down the dual problem to the following one:

$$\min \{2x_1 + x_2 + 4x_3\}$$

$$x_1 - x_2 + x_3 \geq 2$$

$$-2x_1 + x_2 + 2x_3 \geq 1$$

$$x_1 \geq 0, \quad x_2 \geq 0, \quad x_3 \geq 0.$$

Solve the dual problem geometrically and use the duality theorem to determine an optimal solution of the primal problem.

**Solution.** The dual problem reads as

$$\max \{2y_1 + y_2\}$$

$$y_1 - 2y_2 \leq 2$$

$$-y_1 + y_2 \leq 1$$

$$y_1 + 2y_2 \leq 4$$

$$y_1 \geq 0, \quad y_2 \geq 0.$$

The graphical solution (I skip it here) gives  $y_1^* = 3$ ,  $y_2^* = \frac{1}{2}$  and the second inequality is non-active. The complementary slackness condition in the duality theorem implies that  $x_2^* = 0$ . Since  $y_1^*$  and  $y_2^*$  are both positive, the complementary slackness condition implies also that the two inequality constraints in the primal problem are active. Taking into account that  $x_2^* = 0$  we have the equations

$$\begin{aligned}x_1 + x_3 &= 2 \\ -2x_1 + 2x_3 &= 1.\end{aligned}$$

Solving them we obtain the following solution of the primal problem:  $x^* = (\frac{3}{4}, 0, \frac{5}{4})'$ .

**Problem 5.** Solve the problem

$$\min \{(x_1)^2 + 2(x_2)^2 + (x_3)^2\}$$

$$\begin{aligned}x_1 + x_2 - \ln x_3 &\geq 1 \\ x_3 &\geq 1.\end{aligned}$$

by using the KKT theorem.

**Solution.** First we reformulate the problem in the form as in the KKT theorem:

$$\min \{(x_1)^2 + 2(x_2)^2 + (x_3)^2\}$$

$$-x_1 - x_2 + \ln x_3 + 1 \leq 0 \tag{13}$$

$$-x_3 + 1 \leq 0. \tag{14}$$

This problem has a solution since the level sets  $K_c := \{x \in \mathbf{R}^3 : f(x) \leq c\}$  of the objective function are compact and, for example,  $K_2 \cap K \neq \emptyset$ . Then the existence theorem from Chapter 1 is applicable. Every optimal solution is a part of a KKT point.

First we search for KKT points with  $\lambda_0 = 1$ . The Lagrange function is

$$L(x, \mu) = (x_1)^2 + 2(x_2)^2 + (x_3)^2 + \mu_1(-x_1 - x_2 + \ln x_3 + 1) + \mu_2(-x_3 + 1)$$

and the KKT conditions consist of the equations

$$\partial_{x_1} L = 2x_1 - \mu_1 = 0 \quad (15)$$

$$\partial_{x_2} L = 4x_2 - \mu_1 = 0 \quad (16)$$

$$\partial_{x_3} L = 2x_3 + \frac{\mu_1}{x_3} - \mu_2 = 0 \quad (17)$$

$$\mu_1(-x_1 - x_2 + \ln x_3 + 1) = 0 \quad (18)$$

$$\mu_2(-x_3 + 1) = 0 \quad (19)$$

and the inequalities (13), (14) and  $\mu_1 \geq 0$ ,  $\mu_2 \geq 0$ .

We consider the following four cases.

(i)  $\mu_1 = \mu_2 = 0$ . Then from (15)–(17)  $x_1 = x_2 = x_3 = 0$ , which is not a feasible point due to (14).

(ii)  $\mu_1 = 0$ ,  $\mu_2 > 0$ . Then from (19)  $x_3 = 1$  and from (15), (16)  $x_1 = x_2 = 0$ . This is not a feasible point due to (13).

(iii)  $\mu_1 > 0$ ,  $\mu_2 = 0$ . Then (15)–(18) give

$$2x_1 - \mu_1 = 0$$

$$4x_2 - \mu_1 = 0$$

$$2x_3 + \frac{\mu_1}{x_3} = 0$$

$$x_1 + x_2 - \ln x_3 = 1.$$

The third equation gives  $\mu_1 = -2(x_3)^2$ , which contradicts  $\mu_1 > 0$ . Thus we do not obtain a KKT point also in this case.

(iv)  $\mu_1 > 0$ ,  $\mu_2 > 0$ . From (19) we get  $x_3 = 1$ . Equations (15), (16) imply  $x_1 = 2x_2$  and (18) implies  $x_1 + x_2 = 1$ . Then  $x_1 = \frac{2}{3}$ ,  $x_2 = \frac{1}{3}$ . So we obtain a KKT point with  $x = (\frac{2}{3}, \frac{1}{3}, 1)'$ .

Now we try to find abnormal KKT points (with  $\lambda_0 = 0$ ). Such may arise only if the surjectivity condition in the KKT theorem is not fulfilled. We have

$$\partial h(x) = \begin{pmatrix} -1 & -1 & \frac{1}{x_3} \\ 0 & 0 & -1 \end{pmatrix},$$

which has rank = 2 for every  $x_3 \geq 1$ . Thus the surjectivity condition is fulfilled and there are no abnormal KKT points.

Since the only KKT point is  $x = (\frac{2}{3}, \frac{1}{3}, 1)'$  and since the problem has a solution, this point is the unique optimal solution.

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## Sample test

### Tasks:

1. Formulate the KKT theorem for convex problems:

Is the Slater condition fulfilled for the following sets

$$K := \{x \in \mathbf{R}^3 : x_1 - x_3 = 0, x_2 - x_3 \leq -1, (x_1)^2 + (x_2)^2 \leq 1\},$$

$$K_0 := \{x \in \mathbf{R}^3 : x_2 \geq 0\}.$$

2. Consider the problem

$$\min_{x \in K} \{f(x) := x_1 - 2x_2 + 3x_3 - 4x_4\}$$

with

$$K := \{x \in \mathbf{R}^4 : (x_1)^2 + (x_2)^2 + (x_3)^2 + (x_4)^2 \leq 9\}$$

as a primal problem and find the function  $D(\mu)$  of the corresponding dual problem

$$\max_{\mu \geq 0} D(\mu).$$

Then find the solution  $\mu^*$  of the dual problem and evaluate the corresponding  $x^*$  (resulting from the definition of  $D(\mu^*)$ ). Is  $x^*$  a solution of the primal problem and why?

3. Solve the problem

$$\min \{3(x_1)^2 + 4(x_2)^2 + (x_3)^2\}$$

$$x_1 + x_2 - \ln x_3 \geq 1$$

$$x_3 \geq 1.$$

by using the KKT theorem.

4. Apply the simplex method to solve the linear problem

$$\min\{x_1 + x_2 - x_4\}$$

subject to

$$x_1 + 5x_3 = 1,$$

$$x_2 + 2x_4 = 2,$$

$$x_1, x_2, x_3, x_4 \geq 0.$$

*Hint:* Use  $(1, 2, 0, 0)'$  as an initial basis point.