

**Practise Problems-7**  
**MA 321 (Optimization)**

Notations:  $S = \text{Fea}(P) = \{x \in \mathbb{R}^n : g_i(x) \leq 0, \text{ for } i = 1, 2, \dots, m\}$ ,

$D_{x^*} = \{d \in \mathbb{R}^n : g_i(x^* + td) \leq 0, \text{ for all } i = 1, 2, \dots, m, \text{ and for all } 0 \leq t \leq c_d, c_d > 0\}$ ,

$F_0 = \{d \in \mathbb{R}^n : \nabla f(x^*)d < 0\}$ ,  $I = \{i \in \{1, \dots, m\} : g_i(x^*) = 0\}$ ,  $G_0 = \{d \in \mathbb{R}^n : \nabla g_i(x^*)d < 0 \text{ for all } i \in I\}$ .

1. Minimize  $(x_1 - 2)^2 + (x_2 - 3)^2$

subject to

$$x_1^2 + x_2^2 \leq 5.$$

$$2x_1 + x_2 \leq 4.$$

$$-x_1 \leq 0$$

$$-x_2 \leq 0.$$

(a) Check whether all the  $g_i$ 's and  $f$  are convex functions.

(b) Check that  $x^* = [1, 2]^T$  is a KKT point (done in class).

(c) Is  $x^*$  a global minimum of  $f$  over this feasible region?

(d) Does there exist an  $x^* \in \text{Fea}(P)$  such that  $G_0 = D_{x^*}$ ?

(e) Does there exist an  $x^* \in \text{Fea}(P)$  such that  $G_0 \neq D_{x^*}$ ?

2. Consider the problem of minimizing

$$f(x_1, x_2) = -x_1x_2 + x_1^2 + 2x_2^2 - 2x_1 + e^{x_1+x_2} \text{ over } \mathbb{R}^2.$$

(a) Write the first order necessary optimality condition for this problem. Is this condition also sufficient for optimality?

(b) Is  $(0, 0)$  a local minimum point for this problem? If not find a direction  $d$  along which the function  $f$  will decrease.

3. Consider the following problem:

$$\text{Minimize } 4x_1^2 - x_2^2 + 8x_1x_2$$

subject to

$$2x_1 + x_1^2 - x_2 \geq 0$$

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

(a) Does  $(0, 0)$  satisfy the first and the second order necessary conditions for a local minimum?

(b) Find all points  $x^*$  in the feasible region at which  $G_0 = \phi$ .

4. Consider the following problem:

$$\text{Minimize } -x_1^2 - 4x_1x_2 - x_2^2$$

subject to

$$x_2^2 + x_1^2 = 1.$$

(a) If possible find a FJ point which is not a KKT point.

(b) If possible find a KKT point which is not an optimal point.

- (c) Are the first and the second order necessary conditions (that is for any feasible direction  $d$  at  $x^*$ ,  $\nabla f(x^*)d \geq 0$ , and if  $\nabla f(x^*)d = 0$ , then  $d^T \nabla^2 f(x^*)d \geq 0$ ) for a local minimum satisfied at the KKT point/s?
- (d) If the objective function is changed to  $2x_1^2 - x_1x_2 + x_2^2 - x_2$  and if  $S = \text{Fea}(P) = \{(x_1, x_2) : x_2^2 + x_1^2 \leq 1\}$  then find all optimum solutions to this problem.

5. For a nonlinear programming problem (P) of the form,

$$\begin{aligned} &\text{Minimize } f(\mathbf{x}) \\ &\text{subject to } g_i(\mathbf{x}) \leq 0, \text{ for } i = 1, \dots, m, \quad \mathbf{x} \in \mathbb{R}^n, \end{aligned}$$

where all the  $g_i$ 's and  $f$  are continuously differentiable throughout  $\mathbb{R}^n$ , check the correctness of the following statements with proper justification.

- (a) If  $x^* \in \text{Fea}(P)$  is a KKT point of the above problem then  $-\nabla f(x^*)$  lies in the cone generated by  $\nabla g_i(x^*)$ ,  $i \in I$ , where  $I$  gives the indices of the binding constraints (given by  $g_i$ 's) at  $x^*$ .
  - (b) If  $g_2 = -g_3$  then all points  $x \in \text{Fea}(P)$  are FJ (Fritz John) points.
  - (c) If  $\nabla f(x^*) = 0$  for some  $x^* \in \text{Fea}(P)$  then  $x^*$  is a KKT point.
  - (d) If  $x^*$  is an FJ point and there is a solution to the FJ conditions at  $x^*$  with  $u_0 = 0$  ( $u_0$  is the coefficient of  $\nabla f(x^*)$  in the FJ conditions), then  $x^*$  is not a KKT point.
  - (e) If  $x^*$  is a KKT point then it is also an FJ point.
  - (f) If  $x^*$  is an FJ point and  $\nabla g_i(x^*)$ 's are LD, then  $G_0 = \phi$  (or in other words there exists a solution to the FJ conditions with  $u_0 = 0$ ).
  - (g) If  $\nabla g_i(x^*)$ 's are LI, then  $G_0 \neq \phi$ .
  - (h) If  $G_0 = \phi$  then  $\nabla g_i(x^*)$ 's are LD.
  - (i) If  $x^*$  is not an interior point of the feasible region  $S$  of (P) and  $F_0 = \phi$  then  $x^*$  is a local minimizer.
  - (j) If all the  $g_i$ 's and  $f$  are convex functions and  $x^*$  is such that  $F_0 \cap G_0 = \phi$ , then  $x^*$  is a global minimum of  $f$  in  $S$ .
  - (k) If  $x^*$  is a local minima of (P) with  $F_0 \neq \phi$  but  $G_0 = \phi$  then  $x^*$  is not a KKT point.
  - (l) For all  $x^* \in S$ ,  $G_0 = D_{x^*}$ .
  - (m) There exists a (P) such that  $G_0 \neq D_{x^*}$ , for all  $x^* \in S$ .
  - (n) There exists a (P) with  $\text{Fea}(P) \neq \phi$  such that  $G_0 = \phi$  for all  $x^* \in S$ .
  - (o) There exists a (P) with  $\text{Fea}(P) \neq \phi$  such that  $D_{x^*} = \phi$  for all  $x^* \in S$ .
6. Give examples of nonconstant functions on  $\mathbb{R}^n$  which are both convex and concave and those which are neither convex nor concave.
7. For a linear programming problem (P) of the form, Minimize  $\mathbf{c}^T \mathbf{x}$  subject to  $A_{m \times n} \mathbf{x} \leq \mathbf{b}$ ,  $\mathbf{x} \geq \mathbf{0}$ , find the KKT conditions.
8. Consider the linear programming problem.  
Minimize  $2x_1 - 3x_2$

subject to  $x_1 + 2x_2 \leq 3$

$2x_1 + 3x_2 \leq 5$

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

Find the KKT conditions for this problem at a local minimum point of this problem.

Solve the KKT conditions for  $u_i$ 's. Hence find an optimal solution of the dual.