## Final - Optimization (2023-24) Time: 3 hours.

Attempt all questions. The total marks is 50.

1. Let  $f : \mathbf{R}^n \to \mathbf{R}$  be convex and differentiable. Show that  $\mathbf{x}^*$  solves

$$\min_{\mathbf{x}\in\mathbf{R}^n}f(\mathbf{x})$$

if and only if  $\nabla f(\mathbf{x}^*) = \mathbf{0}$ . [6 marks]

2. Consider the constrained optimization problem

$$\min_{\mathbf{x}\in\mathbf{R}^n} f(\mathbf{x}) \qquad \text{subject to } g_m(\mathbf{x}) \le 0, \quad m = 1, 2, \cdots, M, \tag{0.1}$$

where  $f, g_1, \dots, g_m : \mathbf{R}^n \to \mathbf{R}$ . Corresponding to the above problem, for each choice of real numbers  $\lambda_1, \dots, \lambda_M$ , is an unconstrained problem

$$\min_{\mathbf{x}\in\mathbf{R}^n}\left[f(\mathbf{x})+\sum_{m=1}^M\lambda_m g_m(\mathbf{x}).\right]$$

- (a) Write down the dual problem. [2 marks]
- (b) Let  $\lambda^*$  be the solution of the dual problem, and  $\mathbf{x}^*$  be the solution of the primal problem (0.1). Show that if strong duality holds, then  $\lambda^*$  is exactly what is needed to make  $\mathbf{x}^*$  the solution to the unconstrained problem. [6 marks]
- 3. Let  $f : \mathbf{R}^n \to \mathbf{R}$ . Recall that a *subgradient* of f at  $\mathbf{x}$  is a vector  $\mathbf{g}$  such that

$$f(\mathbf{y}) \ge f(\mathbf{x}) + \langle \mathbf{y} - \mathbf{x}, \mathbf{g} \rangle, \quad \text{for all } \mathbf{y} \in \mathbf{R}^n.$$

- (a) Show that if f is convex then there is at least one subgradient of f. [7 marks]
- (b) Show that a subgradient may not exist if f is not convex. [3 marks]
- 4. Consider a convex three times continuously differentiable function  $f : \mathbf{R} \to \mathbf{R}$ . Let  $\xi$  be the global minimizer of f. Let  $\delta > 0$  and  $I_{\delta} = [\xi \delta, \xi + \delta]$ . Assume further that  $f'''(\xi) \neq 0$ , and there exists A > 0 such that

$$\frac{|f'''(x)|}{|f''(y)|} \le A \quad \text{ for all } x, y \in I_{\delta}.$$

Consider the sequence

$$x_{k+1} = x_k - \frac{f'(x_k)}{f''(x_k)}$$

with  $|x_0 - \xi| \leq \min(\delta, \frac{1}{A})$ . Show that

$$\lim_{k \to \infty} \frac{|x_{k+1} - \xi|}{|x_k - \xi|^2} = \mu,$$

for some  $\mu \in (0, \frac{A}{2}]$ . [8 marks]

- 5. Consider the standard form polyhedron  $\{\mathbf{x} : \mathbf{A}\mathbf{x} = \mathbf{b}, \mathbf{x} \ge \mathbf{0}\}$ , where **b** is an  $m \times 1$  vector, and assume that the *m* rows of the matrix  $\mathbf{A}_{m \times n}$  are linearly independent.
  - (a) Suppose that two different bases lead to the same basic solution. Show that the basic solution is degenerate. [3 marks]
  - (b) Consider a degenerate basic solution. Is it true that it corresponds to two or more distinct bases? Prove or give a counterexample. [3 marks]

6. While solving a standard form problem, we arrive at the following tableau, with  $x_3, x_4$ , and  $x_5$  being the basic variables:

	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$
-10	δ	-2	0	0	0
4	-1	$\eta$	1	0	0
1	$\alpha$	-4	0	1	0
β	$\gamma$	3	0	0	1

The entries  $\alpha$ ,  $\beta$ ,  $\gamma$ ,  $\delta$ ,  $\eta$  in the tableau are unknown parameters. For each one of the following statements, find parameter values that will make the statement true.

- (a) The current solution is optimal. [4 marks]
- (b) The current solution is feasible but not optimal. [3 marks]
- 7. The convex hull  $\operatorname{conv}(X)$  of a set  $X \subset \mathbf{R}^n$  is the intersection of all convex sets containing X.
  - (a) Show that conv(X) is convex. [2 marks]
  - (b) Show that conv(X) need not be closed. [3 marks]