

**Final Exam: Question 1 (19 January 2022)**

- Maximum marks: **20**
- Time allotted (including uploading on Moodle): **45 minutes**
- Prove the following theorem.

Let  $f, h^1, \dots, h^m : \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}$  be  $C^1$  functions. Consider the problem of maximizing  $f(x_1, \dots, x_n; a)$  subject to the constraints

$$h^1(x_1, \dots, x_n; a) = 0, \dots, h^m(x_1, \dots, x_n; a) = 0.$$

Let  $x^*(a) = (x_1^*(a), \dots, x_n^*(a))$  solves the problem with corresponding Lagrange multipliers  $\mu_1^*(a), \dots, \mu_m^*(a)$ . Suppose further that  $x_1^*, \dots, x_n^*$  and  $\mu_1^*, \dots, \mu_m^*$  are  $C^1$  functions of  $a$  and that the nondegenerate constraint qualification (NDCQ) holds. Then

$$\frac{d}{da} f(x^*(a); a) = \frac{\partial}{\partial a} L(x^*(a), \mu^*(a); a)$$

where  $L(x, \mu; a) = f(x; a) - \mu_1 h^1(x; a) - \dots - \mu_m h^m(x; a)$ , the Lagrangian for this problem.

**Final Exam: Question 2 (19 January 2022)**

- Maximum marks: **30**
- Time allotted (including uploading on Moodle): **60 minutes**
- Consider the following constrained maximization problem:

$$\text{Maximize}_{\{x \in \mathbb{R}^n\}} f(x) = x^T A x \quad \text{subject to} \quad x^T x = 1,$$

where  $A$  is a given *symmetric*  $n \times n$  matrix.

- (a) [5 marks] Mention clearly which route (necessary or sufficient) you are taking to solve this problem and demonstrate carefully that all the required conditions of that route are satisfied.
- Define the Lagrangian for this problem as  $L(x, \lambda) = x^T A x + \lambda(1 - x^T x)$ .
- (b) [15 marks] Suppose  $x^* \in \mathbb{R}^n$  is a solution to the problem. Show that  $x^*$  is a normalized *eigenvector* of  $A$  and  $\lambda = f(x^*)$ .
- (c) [10 marks] Explain how you can obtain  $f(x^*)$ . Illustrate your method when

$$A = \begin{pmatrix} a & b \\ b & c \end{pmatrix},$$

with  $a > 0$  and  $c > 0$ .

**Final Exam: Question 3 (19 January 2022)**

- Maximum marks: **25**
- Time allotted (including uploading on Moodle): **50 minutes**
- Recall the following theorem.

Theorem (Taylor's Expansion upto Second Order):

*Suppose  $A$  is an open convex subset of  $\mathbb{R}^n$ , and  $f : A \rightarrow \mathbb{R}$  is twice continuously differentiable on  $A$ . Suppose  $x^1$  and  $x^2$  are in  $A$ . Then there is  $0 \leq \theta \leq 1$  such that*

$$f(x^2) - f(x^1) = (x^2 - x^1) \cdot \nabla f(x^1) + \frac{1}{2} (x^2 - x^1) \cdot H_f(\theta x^1 + (1 - \theta)x^2) \cdot (x^2 - x^1).$$

- Suppose  $D$  is an open convex set in  $\mathbb{R}^n$ , and  $g : D \rightarrow \mathbb{R}$  is twice continuously differentiable and *quasiconcave* on  $D$ . Suppose there exists  $x^* \in D$  satisfying

- (i)  $\nabla g(x^*) = 0$ , and
- (ii)  $H_g(x^*)$  is *negative definite*.

- (a) [15 marks] Use the above theorem to prove that  $g$  has a *strict* local maximum at  $x^*$ .
- (b) [10 marks] Prove further that  $x^*$  is a point of *global maximum* of  $g$  on  $D$ .

**Final Exam: Question 4 (19 January 2022)**

- Maximum marks: **25**
- Time allotted (including uploading on Moodle): **50 minutes**
- In this question we will prove the following *sufficiency theorem* under quasiconcave programming.

Suppose  $f, g^j$  ( $j = 1, 2, \dots, m$ ) are continuously differentiable quasiconcave functions on  $\mathbb{R}^n$ . Suppose there is a pair  $(\hat{x}, \hat{\lambda}) \in (\mathbb{R}_+^n \times \mathbb{R}_+^m)$  such that  $(\hat{x}, \hat{\lambda})$  satisfies the Kuhn-Tucker conditions. Suppose further that  $\nabla f(\hat{x}) \neq 0$ . Then  $\hat{x}$  solves the following problem:

Maximize  $f(x)$ , subject to  $x \in C \equiv \{x \in \mathbb{R}_+^n : g^j(x) \geq 0, \text{ for } j = 1, 2, \dots, m\}$ .

- We will develop the proof in the following three steps.
- (a) [4 marks] **Step I:** Given the premises of the theorem, in the lectures we have proved that

$$x \in C \text{ implies } (x - \hat{x}) \cdot \nabla f(\hat{x}) \leq 0.$$

– Which premises of the theorem we do *not* need to get this result?

- (b) [5 marks] **Step II:** Prove that if  $x \in C$  and  $(x - \hat{x}) \cdot \nabla f(\hat{x}) < 0$ , then  $\hat{x}$  solves the problem.
- (c) [16 marks: 5 + 6 + 5] **Step III:** In this step we will prove that if  $x \in C$  and  $(x - \hat{x}) \cdot \nabla f(\hat{x}) = 0$ , then  $\hat{x}$  solves the problem.

We will prove this by contradiction. That is, we will show that if  $(x - \hat{x}) \cdot \nabla f(\hat{x}) = 0$  and  $f(x) > f(\hat{x})$ , then, given the premises of the theorem, a contradiction will arise.

- (i) Given the premises of the theorem, choose  $v \neq 0$  appropriately so that if  $(x - \hat{x}) \cdot \nabla f(\hat{x}) = 0$ , then  $(x + tv - \hat{x}) \cdot \nabla f(\hat{x}) < 0$ , for all  $t > 0$ .

- (ii) If  $f(x) > f(\hat{x})$ , then argue that there exists  $t > 0$  small enough so that  $f(x + tv) > f(\hat{x})$ .
- (iii) Explain the contradiction clearly.