

Mathematics for Economists

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Optimization Problems with Inequality Constraints

Optimization Problem with Inequality Constraints

Proposition (Necessary and sufficient conditions for concave problems)

Let f, g_1, \dots, g_k be C^1 functions defined over \mathbb{R}^n , and let b_1, \dots, b_k be real numbers. Consider the problem of maximizing f on the constraint set defined by the inequalities

$$g_1(\mathbf{x}) \leq b_1, g_2(\mathbf{x}) \leq b_2, \dots, g_k(\mathbf{x}) \leq b_k.$$

Suppose that:

- (1) f is concave
- (2) **either** each g_i is linear **or** each g_i is convex and there exists $\mathbf{x} \in \mathbb{R}^n$ such that $g_i(\mathbf{x}) < b_i$ for $i = 1, \dots, k$.

Form the Lagrangian $L(\mathbf{x}, \mu_1, \dots, \mu_k) = f(\mathbf{x}) - \sum_{i=1}^k \mu_i [g_i(\mathbf{x}) - b_i]$.

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Concave Problems

Proposition (Necessary and sufficient conditions for concave problems)

(Cont'd)

Then $\mathbf{x}^* \in \mathbb{R}^n$ solves the constrained maximization problem under consideration **if and only if** there exist multipliers μ_1^*, \dots, μ_k^* such that

1. $\frac{\partial L}{\partial x_1}(\mathbf{x}^*, \boldsymbol{\mu}^*) = 0, \dots, \frac{\partial L}{\partial x_n}(\mathbf{x}^*, \boldsymbol{\mu}^*) = 0$
2. $\mu_1^* [g_1(\mathbf{x}^*) - b_1] = 0, \dots, \mu_k^* [g_k(\mathbf{x}^*) - b_k] = 0$
3. $\mu_1^* \geq 0, \dots, \mu_k^* \geq 0$
4. $g_1(\mathbf{x}^*) \leq b_1, \dots, g_k(\mathbf{x}^*) \leq b_k.$

Note: The NDCQ is replaced by:

- (2) **either** each g_i is linear **or** each g_i is convex and there exists $\mathbf{x} \in \mathbb{R}^n$ such that $g_i(\mathbf{x}) < b_i$ for $i = 1, \dots, k.$

Concave Problems

- ▶ **Example.** Consider the constrained maximization problem:

$$\max_{x,y,z} f(x, y, z) = x + y - 2z$$

$$\text{s.t. } g_1(x, y, z) = x^2 + y^2 - z \leq 0$$

$$g_2(x, y, z) = -x \leq 0$$

$$g_3(x, y, z) = -y \leq 0$$

$$g_4(x, y, z) = -z \leq 0$$

- ▶ The objective function f is concave
- ▶ Each g_i is convex and there exists a point, e.g. $\mathbf{x} = (1, 1, 3)$, such that $g_i(\mathbf{x}) < 0$ for $i = 1, \dots, 4$
- ▶ Thus a solution to this problem is fully identified by first order conditions

Concave Problems

- ▶ **Example (cont'd).** The Lagrangian is

$$L = x + y - 2z - \lambda_1(x^2 + y^2 - z) + \lambda_2x + \lambda_3y + \lambda_4z$$

- ▶ The first order conditions are

$$2x\lambda_1 = 1 + \lambda_2 \quad (1)$$

$$2y\lambda_1 = 1 + \lambda_3 \quad (2)$$

$$\lambda_1 + \lambda_4 = 2 \quad (3)$$

$$\lambda_1(x^2 + y^2 - z) = 0 \quad (4)$$

$$\lambda_2x = 0 \quad (5)$$

$$\lambda_3y = 0 \quad (6)$$

$$\lambda_4z = 0 \quad (7)$$

$$\lambda_1 \geq 0, \lambda_2 \geq 0, \lambda_3 \geq 0, \lambda_4 \geq 0 \quad (8)$$

$$x^2 + y^2 - z \leq 0, x \geq 0, y \geq 0, z \geq 0 \quad (9)$$

Concave Problems

- ▶ **Example (cont'd).** If $\lambda_1 = 0$ or $x = 0$, then $\lambda_2 = -1$ by (1), so contradicting (8). Thus we must have $\lambda_1 > 0$ and $x > 0$
- ▶ By the same token, we can use (2) to conclude that $y > 0$
- ▶ $x > 0$ and $y > 0$ imply $\lambda_2 = \lambda_3 = 0$ via (5) and (6)
- ▶ Since $\lambda_1 > 0$, we get $x = y = \frac{1}{2\lambda_1}$ from (1) and (2). Consequently, $z = \frac{1}{2\lambda_1^2} > 0$, which in turn implies $\lambda_4 = 0$ via (7)
- ▶ Finally, we get $\lambda_1 = 2$ from (3)
- ▶ Thus the unique solution is

$$x = y = \frac{1}{4}, \quad z = \frac{1}{8}$$

with multipliers

$$\lambda_1 = 2, \quad \lambda_2 = \lambda_3 = \lambda_4 = 0.$$

Concave Problems

- ▶ **Exercise.** Consider the constrained maximization problem:

$$\max_{x,y,z} f(x,y,z) = 3\ln(z+1) - z - 2x - y$$

$$\text{s.t. } g_1(x,y,z) = z^2 - x - y \leq 0$$

$$g_2(x,y,z) = -x \leq 0$$

$$g_3(x,y,z) = -y \leq 0$$

$$g_4(x,y,z) = -z \leq 0$$

- ▶ Can you apply the Proposition at pp. 2-3? Why or why not?
- ▶ Show that the unique solution to this problem is

$$(x,y,z) = \left(0, \frac{1}{4}, \frac{1}{2}\right)$$

Concave Problems

- ▶ **Exercise.** Consider the constrained maximization problem:

$$\begin{aligned} \max_{x,y} \quad & f(x, y) = x + ay \\ \text{s.t.} \quad & g_1(x, y, z) = x^2 + y^2 \leq 1 \\ & g_2(x, y, z) = -x - y \leq 0, \end{aligned}$$

where $a \in \mathbb{R}$ is a parameter

- ▶ Can you apply the Proposition at pp. 2-3? Why or why not?
- ▶ Show that:
 - ▶ when $a \geq -1$, the unique solution is

$$(x, y) = \left(\frac{1}{\sqrt{1+a^2}}, \frac{a}{\sqrt{1+a^2}} \right);$$

- ▶ when $a < -1$, the unique solution is

$$(x, y) = \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right).$$

Mixed Constraints

- ▶ Suppose we have to solve the following constrained maximization problem:

$$\begin{aligned} \max_{x,y} \quad & 3xy - x^3 \\ \text{s.t.} \quad & 2x - y = -5 \\ & -5x - 2y \leq -37 \\ & x \geq 0 \\ & y \geq 0 \end{aligned}$$

- ▶ This is a problem with **mixed** constraints: one *equality* and three *inequality* constraints

Mixed Constraints

- ▶ We can rewrite the problem as one with inequality constraints only and then solve it. That is,

$$\begin{aligned} \max_{x,y} \quad & 3xy - x^3 \\ \text{s.t.} \quad & \mathbf{2x - y \leq -5} \\ & \mathbf{-2x + y \leq 5} \\ & -5x - 2y \leq -37 \\ & x \geq 0 \\ & y \geq 0 \end{aligned}$$

- ▶ *Alternatively*, we can combine results from previous lectures and formulate a general proposition that will enable us to solve a problem like this without doing any rewriting/transformation

Mixed Constraints

- ▶ The general formulation of a constrained maximization problem with n variables and *mixed constraints* (k inequality and m equality constraints) is to
 - ▶ **maximize** the objective function $f(x_1, \dots, x_n)$ with respect to (x_1, \dots, x_n)
 - ▶ subject to the constraints:

$$g_1(x_1, \dots, x_n) \leq b_1$$

$$g_2(x_1, \dots, x_n) \leq b_2$$

... ..

$$g_k(x_1, \dots, x_n) \leq b_k$$

$$h_1(x_1, \dots, x_n) = c_1$$

$$h_2(x_1, \dots, x_n) = c_2$$

... ..

$$h_m(x_1, \dots, x_n) = c_m$$

Mixed Constraints

- ▶ The non-degenerate constraint qualification (NDCQ) at a given point $\mathbf{x} = (x_1, \dots, x_n)$ is formulated as follows:
 - ▶ Without loss of generality, suppose that the first k_0 inequality constraints ($k_0 \leq k$) are binding at \mathbf{x} , and the last $k - k_0$ are inactive at \mathbf{x}
 - ▶ The Jacobian of the *equality* constraints and the *binding inequality* constraints is

$$D\mathbf{g}(\mathbf{x}) = \begin{pmatrix} \frac{\partial g_1}{\partial x_1}(\mathbf{x}) & \cdots & \frac{\partial g_1}{\partial x_n}(\mathbf{x}) \\ \vdots & \ddots & \vdots \\ \frac{\partial g_{k_0}}{\partial x_1}(\mathbf{x}) & \cdots & \frac{\partial g_{k_0}}{\partial x_n}(\mathbf{x}) \\ \frac{\partial h_1}{\partial x_1}(\mathbf{x}) & \cdots & \frac{\partial h_1}{\partial x_n}(\mathbf{x}) \\ \vdots & \ddots & \vdots \\ \frac{\partial h_m}{\partial x_1}(\mathbf{x}) & \cdots & \frac{\partial h_m}{\partial x_n}(\mathbf{x}) \end{pmatrix}$$

- ▶ We say that the NDCQ is satisfied at \mathbf{x} if the rank of $D\mathbf{g}(\mathbf{x})$ is as large as it can be

Mixed Constraints

Proposition (First order necessary conditions)

Let $f, g_1, \dots, g_k, h_1, \dots, h_m$ be C^1 functions defined on \mathbb{R}^n . Suppose that:

1. \mathbf{x}^* is a **local maximizer** of f on the constraint set defined by

$$g_1(\mathbf{x}) \leq b_1, \dots, g_k(\mathbf{x}) \leq b_k, h_1(\mathbf{x}) = c_1, \dots, h_m(\mathbf{x}) = c_m$$

2. the NDCQ is satisfied at \mathbf{x}^* .

Form the Lagrangian $L(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\lambda}) = f(\mathbf{x}) - \sum_{i=1}^k \mu_i [g_i(\mathbf{x}) - b_i] - \sum_{i=1}^m \lambda_i [h_i(\mathbf{x}) - c_i]$.

Then, there exist multipliers $\mu_1^*, \dots, \mu_k^*, \lambda_1^*, \dots, \lambda_m^*$ such that:

1. $\frac{\partial L}{\partial x_1}(\mathbf{x}^*, \boldsymbol{\mu}^*, \boldsymbol{\lambda}^*) = 0, \dots, \frac{\partial L}{\partial x_n}(\mathbf{x}^*, \boldsymbol{\mu}^*, \boldsymbol{\lambda}^*) = 0$
2. $\mu_1^* [g_1(\mathbf{x}^*) - b_1] = 0, \dots, \mu_k^* [g_k(\mathbf{x}^*) - b_k] = 0$
3. $h_1(\mathbf{x}^*) = c_1, \dots, h_m(\mathbf{x}^*) = c_m$
4. $\mu_1^* \geq 0, \dots, \mu_k^* \geq 0$
5. $g_1(\mathbf{x}^*) \leq b_1, \dots, g_k(\mathbf{x}^*) \leq b_k$.

Mixed Constraints

- ▶ Back to the maximization problem:

$$\begin{aligned} \max_{x,y} \quad & 3xy - x^3 \\ \text{s.t.} \quad & 2x - y = -5 \\ & -5x - 2y \leq -37 \\ & x \geq 0 \\ & y \geq 0 \end{aligned}$$

- ▶ The Lagrangian is

$$L = 3xy - x^3 - \lambda(2x - y + 5) - \mu_1(-5x - 2y + 37) + \mu_2x + \mu_3y$$

Mixed Constraints

- ▶ The first order conditions are:

$$\frac{\partial L}{\partial x} = 0 \iff 3y - 3x^2 - 2\lambda + 5\mu_1 + \mu_2 = 0$$

$$\frac{\partial L}{\partial y} = 0 \iff 3x + \lambda + 2\mu_1 + \mu_3 = 0$$

$$\mu_1(-5x - 2y + 37) = 0$$

$$\mu_2 x = 0$$

$$\mu_3 y = 0$$

$$\mu_1, \mu_2, \mu_3 \geq 0$$

$$2x - y + 5 = 0$$

$$-5x - 2y + 37 \leq 0, \quad x \geq 0, \quad y \geq 0$$

- ▶ **Exercise:** Show that the only point that satisfies the first order conditions is such that $x = 5$, $y = 15$, $\lambda = -15$, $\mu_1 = \mu_2 = \mu_3 = 0$
- ▶ **Exercise:** Show that the NDCQ is always satisfied

Verifying the Optimality

- ▶ Assume \mathbf{x}^* is a candidate for an optimal point (satisfies FOCs), is it optimal (locally or globally)?
 1. Is the problem concave (or convex)?
 - ▶ in maximization f should be concave and the feasible set convex
 - ▶ note 1: inequality constraints are $g_i(\mathbf{x}) \leq 0$, $i = 1, \dots, m$ and g_i are convex functions, and inequality constraints are linear, the feasible set is convex
 - ▶ \mathbf{x}^* is a global maximizer
 - ▶ note 2: sometimes equality constraints can be turned into inequalities without affecting the optimality, which may help
 2. Can the problem be transformed into a concave problem?
 - ▶ for example Cobb-Douglas functions are log-concave
 - ▶ note: with log-transformation variables need to be > 0

Verifying the Optimality

3. Is the feasible set compact and objective function continuous? Are all the critical points known?
 - ▶ If yes, and NDCQ does not fail in the feasible set, evaluate the objective function at critical points and find the global maximizer
4. Try the second order conditions
 - ▶ If the Hessian of the Lagrangian is neg. def. you have a local maximizer
 - ▶ If you cant directly say anything about the definiteness of the Hessian of L , try the Bordered Hessian