

EC9A0: Pre-sessional Advanced Mathematics Course

Constrained Optimisation: Equality Constraints

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Introduction

- Suppose $D \subseteq \mathbb{R}^K$, K finite, is open.
- $f : D \rightarrow \mathbb{R}$
- $g : D \rightarrow \mathbb{R}^J$, with $J \leq K$.
- We would like to solve:

$$\max_{x \in D} f(x) \text{ s.t. } g(x) = 0, \quad (1)$$

- In the previous notation, one wants to find

$$\max_{x \in D'} f(x)$$

where $D' = \{x \in D \mid g(x) = 0\}$.

- We will analyse when the Lagrangean method can be used.
- We will derive necessary and sufficient conditions for a constrained global maximum.

Pseudo-Theorem

- The method that is usually applied consists of the following steps:

- ① Defining the Lagrangean function $\mathcal{L} : D \times \mathbb{R}^J \rightarrow \mathbb{R}$, by

$$\mathcal{L}(x, \lambda) = f(x) + \sum_{j=1}^J \lambda_j g_j(x);$$

- ② Finding $(x^*, \lambda^*) \in D \times \mathbb{R}^J$ such that $D\mathcal{L}(x^*, \lambda^*) = 0$.
- That is, a recipe is applied as though there is a “Theorem” that states:

Let $f : D \rightarrow \mathbb{R}$ and $g : D \rightarrow \mathbb{R}^J$ be differentiable. Then $x^* \in D$ solves Problem (1) if and only if there exists $\lambda^* \in \mathbb{R}^J$ such that (x^*, λ^*) solves:

$$\frac{\partial f(x^*)}{\partial x_i} + \sum_{j=1}^J \lambda_j^* \frac{\partial g_j(x^*)}{\partial x_i} = 0, \text{ for all } i = 1, \dots, K.$$

Counterexample

- $f(x_1, x_2) = x_1 x_2$ and $g(x_1, x_2) = (1 - x_1 - x_2)^3$.

$$x^* \text{ solves } \max_{x \in \mathbb{R}^2} f(x) \text{ s.t. } g(x) = 0 \Leftrightarrow x^* \text{ solves } \max_{x \in \mathbb{R}_+^2} f(x) \text{ s.t. } g(x) = 0.$$

- The second problem has a solution by Weierstrass Theorem.
- The unique maximiser is $(x_1^*, x_2^*) = (\frac{1}{2}, \frac{1}{2})$.
- According to the "Theorem" there is λ^* such that $(x_1^*, x_2^*, \lambda^*)$ solves:

$$(a) \quad \frac{\partial \mathcal{L}}{\partial x_1} = 0 \Leftrightarrow x_2 - 3\lambda(1 - x_1 - x_2)^2 = 0$$

$$(b) \quad \frac{\partial \mathcal{L}}{\partial x_2} = 0 \Leftrightarrow x_1 - 3\lambda(1 - x_1 - x_2)^2 = 0$$

$$(c) \quad \frac{\partial \mathcal{L}}{\partial \lambda} = 0 \Leftrightarrow (1 - x_1 - x_2)^3 = 0$$

- A solution to this system of equations does not exist.
- Equation (c) implies that at any solution it must be the case that $x_1^* + x_2^* = 1$.
- (a) and (b) imply that both x_1^* and x_2^* are zero, a contradiction.

Intuitive Argument

- Suppose $D = \mathbb{R}^2$ and $J = 1$, Given $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ and $g : \mathbb{R}^2 \rightarrow \mathbb{R}$.
- We want to solve

$$\max_{(x,y) \in \mathbb{R}^2} f(x, y) \quad \text{s.t. } g(x, y) = 0. \quad (\text{P})$$

- Suppose:
 - A1 There is $h : \mathbb{R} \rightarrow \mathbb{R}$ such that $g(x, y) = 0$ if and only if $y = h(x)$.
 - A2 The function h is differentiable.
- A "crude" method would be to study the unconstrained problem

$$\max_{x \in \mathbb{R}} F(x), \quad (\text{P}^*)$$

where $F : \mathbb{R} \rightarrow \mathbb{R}$ is defined by $F(x) = f(x, h(x))$.

Intuitive argument

- $g(x, h(x)) = 0 \Rightarrow g'_x(x, h(x)) + g'_y(x, h(x))h'(x) = 0$,
- $h'(x) = -\frac{g'_x(x, h(x))}{g'_y(x, h(x))}$.
- Apply FONC to (P*): x^* solves $\max_{x \in \mathbb{R}} F(x)$ only if $F'(x^*) = 0$.

$$f'_x(x^*, h(x^*)) + f'_y(x^*, h(x^*))h'(x^*) = 0,$$

$$\Downarrow$$

$$f'_x(x^*, h(x^*)) - f'_y(x^*, h(x^*)) \frac{g'_x(x^*, h(x^*))}{g'_y(x^*, h(x^*))} = 0.$$

- Define $y^* = h(x^*)$ and $\lambda^* = -\frac{\partial_y f(x^*, y^*)}{\partial_y g(x^*, y^*)} \in \mathbb{R}$,
- Then, we get that (x^*, y^*, λ^*) solves

$$f'_x(x^*, y^*) + \lambda^* g'_x(x^*, y^*) = 0,$$

$$f'_y(x^*, y^*) + \lambda^* g'_y(x^*, y^*) = 0.$$

Intuitive argument

- The “crude” method has shown that:

Let $f : D \rightarrow \mathbb{R}$ and $g : D \rightarrow \mathbb{R}^J$ be differentiable and (A1)-(A2) hold. If $x^* \in D$ is a local maximiser in (1), there exists $\lambda^* \in \mathbb{R}^J$ such that (x^*, λ^*) solves:

$$\frac{\partial f(x^*)}{\partial x_i} + \sum_{j=1}^J \lambda_j^* \frac{\partial g(x^*)}{\partial x_i} = 0, \text{ for all } i = 1, \dots, K.$$

- Under what conditions (A1) and (A2) hold?
- Under what conditions h exists and is differentiable?

Implicit Function Theorem

- We assumed h exists and
- We assumed $g'_y(x^*, y^*) \neq 0$. Of course, $g_x(x^*, y^*) \neq 0$ would be enough.
- What we actually require is $Dg(x^*, y^*)$ has rank 1, its maximum possible.
- Is this a general result, or does it only work in our simplified case?

Theorem The Implicit Function Theorem

Let $D \subseteq \mathbb{R}^K$ and let $g : D \rightarrow \mathbb{R}^J \in \mathcal{C}^1$, with $J < K$. If $y^* \in \mathbb{R}^J$ and $(x^*, y^*) \in D$ is such that $\text{rank}(D_y g(x^*, y^*)) = J$, then there exist $\varepsilon, \delta > 0$ and $h : B_\varepsilon(x^*) \rightarrow B_\delta(y^*) \in \mathcal{C}^1$ such that:

- 1 $\forall x \in B_\varepsilon(x^*), (x, h(x)) \in D;$
- 2 $\forall x \in B_\varepsilon(x^*), g(x, y) = g(x^*, y^*)$ for $y \in B_\delta(y^*)$ iff $y = h(x);$
- 3 $\forall x \in B_\varepsilon(x^*), Dh(x) = -D_y g(x, h(x))^{-1} D_x g(x, h(x)).$

First Order Necessary Conditions

Theorem

Let $f : D \rightarrow \mathbb{R}$ and $g : D \rightarrow \mathbb{R}^J$ be \mathbb{C}^1 . If $x^* \in D$ is a local maximiser in (1) and $\text{rank}(Dg(x^*)) = J$, there exists $\lambda^* \in \mathbb{R}^J$ such that (x^*, λ^*) solves:

$$\frac{\partial f(x^*)}{\partial x_i} + \sum_{j=1}^J \lambda_j^* \frac{\partial g_j(x^*)}{\partial x_i} = 0, \text{ for all } i = 1, \dots, K.$$
$$g_j(x^*) = 0 \text{ for all } j = 1, \dots, J.$$

Second Order Necessary Conditions

- The SONC for problem (P^*) is that $F''(x^*) \leq 0$. Note that:

$$F''(x) = f'_{xx}(x, h(x)) + [f'_{xy}(x, h(x)) + f'_{yx}(x, h(x))]h'(x) + f'_{yy}(x, h(x))h'(x)^2 + f'_y(x, h(x))h''(x),$$

$$h''(x) = -\frac{\partial}{\partial x} \left(\frac{g_x(x, h(x))}{g_y(x, h(x))} \right) = -\frac{1}{g_y(x, h(x))} \begin{bmatrix} 1 & h'(x) \end{bmatrix} D^2 g(x, h(x)) \begin{bmatrix} 1 \\ h'(x) \end{bmatrix}$$

- Substituting h'' and writing in matrix form, $F'' \leq 0$ becomes

$$\begin{bmatrix} 1 & h'(x) \end{bmatrix} D^2 f(x, h(x)) \begin{bmatrix} 1 \\ h'(x) \end{bmatrix} - \frac{f'_y(x, h(x))}{g'_y(x, h(x))} \begin{bmatrix} 1 & h'(x) \end{bmatrix} D^2 g(x, h(x)) \begin{bmatrix} 1 \\ h'(x) \end{bmatrix} \leq 0$$

$$\Leftrightarrow \begin{pmatrix} 1 & h'(x^*) \end{pmatrix} D^2_{(x,y)} \mathcal{L}(x^*, y^*, \lambda^*) \begin{pmatrix} 1 \\ h'(x^*) \end{pmatrix} \leq 0.$$

Second Order Necessary Conditions

- This condition is satisfied if $D_{(x,y)}^2 \mathcal{L}(x^*, y^*, \lambda^*)$ is negative semi-definite.
- Notice that

$$\begin{pmatrix} 1 & h'(x^*) \end{pmatrix} \cdot Dg(x^*, y^*) = 0,$$

so it suffices that we guarantee that for every $\Delta \in \mathbb{R}^2 \setminus \{0\}$ such that $\Delta \cdot Dg(x^*) = 0$ we have that $\Delta^\top D_{(x,y)}^2 \mathcal{L}(x^*, y^*, \lambda^*) \Delta \leq 0$.

- So, in summary, we have argued that:

Let $f : D \rightarrow \mathbb{R}$ and $g : D \rightarrow \mathbb{R}^J$ be \mathbb{C}^1 . If $x^* \in D$ is a local maximiser in (1) and $\text{rank}(Dg(x^*)) = J$, then $\Delta^\top D_{xx}^2 \mathcal{L}(x^*, \lambda^*) \Delta \leq 0$ for all $\Delta \in \mathbb{R}^2 \setminus \{0\}$ such that $\Delta \cdot Dg(x^*) = 0$.

First and Second Order Necessary Conditions

Theorem Lagrange - FONC and SONC

Let $f : D \rightarrow \mathbb{R}$ and $g : D \rightarrow \mathbb{R}^J$ be \mathbb{C}^2 with $J \leq K$. If $x^* \in D$ is a local maximiser in (1) and $\text{rank}(Dg(x^*)) = J$, then there exists $\lambda^* \in \mathbb{R}^J$ such that

① $D_{(x,\lambda)}\mathcal{L}(x^*, \lambda^*) = 0.$

② $\Delta^\top D_{xx}^2\mathcal{L}(x^*, \lambda^*)\Delta \leq 0$ for all $\Delta \in \mathbb{R}^J \setminus \{0\}$ satisfying $\Delta \cdot Dg(x^*) = 0;$

Necessary Conditions are not Sufficient

- The existence of $(x^*, \lambda^*) \in \mathbb{R}^K \times \mathbb{R}^J$ such that

$$\frac{\partial f(x^*)}{\partial x_i} + \sum_{j=1}^J \lambda_j^* \frac{\partial g_j(x^*)}{\partial x_i} = 0, \text{ for all } i = 1, \dots, K.$$
$$g_j(x^*) = 0 \text{ for all } j = 1, \dots, J.$$

might not be sufficient for x^* to be a local maximiser of Problem 1.

Counterexample

- $f(x_1, x_2) = -\left(\frac{1}{2} - x_1\right)^3$ and $g(x_1, x_2) = 1 - x_1 - x_2$.
- $(x_1^*, x_2^*, \lambda^*) = \left(\frac{1}{2}, \frac{1}{2}, 0\right)$ satisfies the constraint qualification, it solves

$$(a) \quad \frac{\partial \mathcal{L}}{\partial x_1} = 0 \iff 3\left(\frac{1}{2} - x_1\right)^2 - \lambda = 0$$

$$(b) \quad \frac{\partial \mathcal{L}}{\partial x_2} = 0 \iff -\lambda = 0$$

$$(c) \quad \frac{\partial \mathcal{L}}{\partial \lambda} = 0 \iff 1 - x_1 - x_2 = 0$$

and satisfies the (necessary) second order condition since

$$\frac{\partial \mathcal{L}}{\partial x_i, x_j}(x_1^*, x_2^*, \lambda^*) = 0, \text{ for } i = 1, 2$$

$$\frac{\partial \mathcal{L}}{\partial x_i, x_j}(x_1^*, x_2^*, \lambda^*) = 0, \text{ for } i \neq j.$$

- However, $(x_1^*, x_2^*) = \left(\frac{1}{2}, \frac{1}{2}\right)$ is not a local maximiser since $f\left(\frac{1}{2}, \frac{1}{2}\right) = 0$ but $\left(\frac{1}{2} + \varepsilon, \frac{1}{2} - \varepsilon\right)$ is also in the constrained set and $f\left(\frac{1}{2} + \varepsilon, \frac{1}{2} - \varepsilon\right) > 0$ for any $\varepsilon > 0$.

First and Second Order Sufficient Conditions

Theorem Lagrange - FOSC and SOSC

Let $f : D \rightarrow \mathbb{R}$ and $g : D \rightarrow \mathbb{R}^J$ be \mathbb{C}^2 , with $J \leq K$. If $(x^*, \lambda^*) \in \mathbb{R}^K \times \mathbb{R}^J$ satisfy:

① $D_{(x,\lambda)} \mathcal{L}(x^*, \lambda^*) = 0$ and

② $\Delta^\top D_{xx}^2 \mathcal{L}(x^*, \lambda^*) \Delta < 0$ for all $\Delta \in \{\mathbb{R}^J \setminus \{0\} : \Delta \cdot Dg(x^*) = 0\}$.

Then, x^* is a local maximiser in Problem (1).
