

# EC9A0: Pre-sessional Advanced Mathematics Course

## Constrained Optimisation: Equality Constraints

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September 2023

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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  $(x^*, \lambda^*)$  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  $\lambda^*$  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  $(x^*, y^*, \lambda^*)$  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  $(x^*, \lambda^*)$  solves:

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

- 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  $g : D \rightarrow \mathbb{R}^J \in \mathbb{C}^1$ , 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  $h : B_\varepsilon(x^*) \rightarrow B_\delta(y^*) \in \mathbb{C}^1$  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  $(x^*, \lambda^*)$  solves:

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

# 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  $D^2_{(x,y)}\mathcal{L}(x^*, y^*, \lambda^*)$  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  $\Delta^\top D^2_{(x,y)}\mathcal{L}(x^*, y^*, \lambda^*)\Delta \leq 0$ .

- 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  $\Delta^\top D^2_{xx}\mathcal{L}(x^*, \lambda^*)\Delta \leq 0$  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

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

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

## Counterexample

- $f(x_1, x_2) = -(\frac{1}{2} - x_1)^3$  and  $g(x_1, x_2) = 1 - x_1 - x_2$ .
- $(x_1^*, x_2^*, \lambda^*) = (\frac{1}{2}, \frac{1}{2}, 0)$  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

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

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

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