

Mathematics for Economists: Lecture 7

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This lecture covers

1. Existence of solutions to optimization problems
2. Optimization subject to an equality constraint: first-order conditions
3. Examples of equality constrained optimization problems

Existence of optimal choices: why care?

Example (1 is the largest natural number)

Proof.

- ▶ Denote the largest natural number (i.e. strictly positive integer) by x .
- ▶ Since x is a natural number, also x^2 is a natural number.
- ▶ Since x is the largest natural number, we have:

$$x \geq x^2.$$

- ▶ Dividing both sides by x (a positive number since it is a natural number), we get

$$1 \geq x.$$

- ▶ Since all natural numbers are larger than or equal to 1, the claim follows.
- ▶ Where is the mistake?

Existence helps the characterization

- ▶ Suppose that you know that a problem has a solution
- ▶ Suppose you know that a single point satisfies first-order necessary conditions
- ▶ Do you need to worry about second-order conditions at the critical point?
- ▶ Hence it is important to find general enough conditions that guarantee the existence of a solution
- ▶ Weierstrass' theorem is pretty good at this

Existence of optimal choices

- ▶ You need to check conditions
 1. On the domain (feasible set) of the problem
 2. On the objective function
- ▶ Feasible set F bounded
 1. Can you fit the feasible set in a large enough box of finite size?
 2. To do this, for $F \subset \mathbb{R}^n$, show that $F \subset [-M, M]^n$ for some $M < \infty$.
 3. E.g. $F = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{p} \cdot \mathbf{x} \leq w, \mathbf{x} \geq 0\}$ for some strictly positive price vector \mathbf{p} satisfies $x_i \leq \frac{w}{\min_i p_i}$ for all i .
- ▶ Feasible set F closed
 1. If $\mathbf{x}_n \rightarrow \mathbf{x}$ and $\mathbf{x}_n \in F$ for all n , then $\mathbf{x} \in F$.
 2. E.g. $F = \{\mathbf{x} \mid g(\mathbf{x}) \leq 0\}$ for a continuous $g : \mathbb{R}^n \rightarrow \mathbb{R}^k$.
 3. $F = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{p} \cdot \mathbf{x} \leq w, \mathbf{x} \geq 0\}$ satisfies this.

Existence of optimal choices

- ▶ Objective function continuous
 1. Almost all functions you will encounter in economics are continuous
 2. Utility functions, cost functions etc.
 3. Failure of continuity: demand for price setting firms where all buyers buy from the firm with the lowest price.
- ▶ A set $F \subset \mathbb{R}^n$ is called *compact* if it is closed and bounded.
- ▶ Once we recall that all sets in \mathbb{R} have a greatest lower bound (infimum) and a lowest upper bound (supremum), we are ready for the main existence result, Weierstrass theorem

Main existence theorem

Theorem (Weierstrass' Theorem)

Suppose f is a continuous function on a compact set F , and

$$M = \sup_{x \in F} f(x), \quad m = \inf_{x \in F} f(x).$$

Then there exist points $\bar{x}, \underline{x} \in F$ such that $f(\bar{x}) = M$ and $f(\underline{x}) = m$.

Weierstrass' theorem

Remark

To see that F must be closed and bounded and that f has to be continuous, consider the following examples where one property fails in each case:

1. $f(x) = x$ and $F = \mathbb{R}$ (F not bounded).
2. $f(x) = x$ and $F = \{x : 0 < x < 1\}$. (F not closed).
3. $f(x) = x$ for $0 \leq x < 1$, $f(1) = 0$ and $E = \{x : 0 \leq x \leq 1\}$ (f not continuous).

Example: Consumer's problem

- ▶ A consumer chooses how to spend her wealth w on two goods x_1, x_2 whose prices are p_1, p_2 .
- ▶ Continuously differentiable utility, strictly positive marginal utility for both goods.
- ▶ The feasible set (budget set) $\{(x_1, x_2) | x_1 \geq 0, x_2 \geq 0, p_1 x_1 + p_2 x_2 \leq w\}$ is closed (since it is defined by weak inequalities for a continuous constraint function).
- ▶ By Weierstrass theorem, a utility maximizing choice exists. What can we say about it?
- ▶ Budget constraint must bind: if $p_1 \hat{x}_1 + p_2 \hat{x}_2 < w$, then $(\hat{x}_1 + h, \hat{x}_2)$ is feasible and:

$$u(\hat{x}_1 + h, \hat{x}_2) > u(\hat{x}_1, \hat{x}_2).$$

- ▶ Therefore, we must have $p_1 \hat{x}_1 + p_2 \hat{x}_2 = w$.
- ▶ Can the indifference curve through the optimum intersect the budget line?

Optimization with a single equality constraint

- ▶ Local considerations: Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be the objective function to be maximized
- ▶ Suppose the constraints take the form $g(x) = g(x_1, \dots, x_n) = 0$.
- ▶ In other words, $F = \{x : g(x) = 0\}$. We write the maximization problem often as:

$$\begin{aligned} & \max_x f(x) \\ & \text{subject to } g(x) = 0. \end{aligned}$$

Optimization with a single equality constraint

- ▶ A solution to this problem finds point $\hat{\mathbf{x}}$ such that $f(\hat{\mathbf{x}}) \geq f(\mathbf{x})$ for all $\mathbf{x} \in F$.
- ▶ What can we say about such an $\hat{\mathbf{x}}$?
- ▶ At this point, we do not know if it exists. If it exists, and f is differentiable, then for small Δ ,

$$f(\hat{\mathbf{x}} + \Delta(\mathbf{x} - \hat{\mathbf{x}})) - f(\hat{\mathbf{x}}) = Df(\hat{\mathbf{x}})(\mathbf{x} - \hat{\mathbf{x}})\Delta \leq 0$$

for all feasible directions $(\mathbf{x} - \hat{\mathbf{x}})$.

- ▶ But how do we know which directions are feasible?

Optimization with a single equality constraint

- ▶ Assume that the function g defining the constraint is also differentiable.
- ▶ To find the feasible directions, we go back to implicit function theorem.
- ▶ If $\hat{\mathbf{x}} \in F$ and $\frac{\partial g}{\partial x_i}(\hat{\mathbf{x}}) \neq 0$ for some $i \in \{1, \dots, n\}$, then we can find a write $x_i = h(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n) =: h(\mathbf{x}_{-i})$ in a neighborhood of $\hat{\mathbf{x}}_{-i}$ so that

$$g(h(\mathbf{x}_{-i}), \mathbf{x}_{-i}) = 0.$$

- ▶ Notice that it is not possible to use the implicit function theorem at a critical point of the constraint function.
- ▶ Therefore, we must assume that $Dg(\hat{\mathbf{x}}) \neq 0$.
- ▶ We call this the constraint qualification and we will see different versions of this in more complex situations with multiple constraints.

Optimization with a single equality constraint

- ▶ Since the function g is at constant value in the feasible set, we have for all feasible directions $(\mathbf{x} - \hat{\mathbf{x}})$:

$$Dg(\hat{\mathbf{x}})(\mathbf{x} - \hat{\mathbf{x}}) = 0.$$

- ▶ Notice also that if $(\mathbf{x} - \hat{\mathbf{x}})$ is feasible, then also $-(\mathbf{x} - \hat{\mathbf{x}})$ is feasible.
- ▶ Linear approximation for optimum at $\hat{\mathbf{x}}$ implies that for all feasible directions,

$$Df(\hat{\mathbf{x}})(\mathbf{x} - \hat{\mathbf{x}}) = 0.$$

- ▶ But therefore we have shown that at optimum $\hat{\mathbf{x}}$,

$$\nabla f(\hat{\mathbf{x}}) = \mu \nabla g(\hat{\mathbf{x}}).$$

- ▶ We have the following necessary condition for a constrained optimum at $\hat{\mathbf{x}}$:
 1. the gradient of the objective function must be a scalar multiple of the gradient of the constraint function at the optimum
 2. the choice must be feasible, i.e. $g(\hat{\mathbf{x}}) = 0$
 3. we have assumed constraint qualification at optimum

Lagrangian function

- ▶ The previous discussion motivates the following function that incorporates the constraints into an augmented objective function called the Lagrangian function.
- ▶ For a constrained optimization problem, we define the following function of $n + 1$ variables:

$$\mathcal{L}(\mathbf{x}, \mu) = f(\mathbf{x}) - \mu g(\mathbf{x}).$$

- ▶ We call the new variable μ the Lagrange multiplier. We will give it a good economic interpretation next week.
- ▶ We are interested in the critical points of this augmented function. Therefore we look for $(\hat{\mathbf{x}}, \hat{\mu})$ such that

$$\frac{\partial \mathcal{L}}{\partial x_i}(\hat{\mathbf{x}}, \hat{\mu}) = \frac{\partial f}{\partial x_i}(\hat{\mathbf{x}}) - \hat{\mu} \frac{\partial g}{\partial x_i}(\hat{\mathbf{x}}) = 0 \text{ for all } i,$$

$$\frac{\partial \mathcal{L}}{\partial \mu}(\hat{\mathbf{x}}, \hat{\mu}) = g(\hat{\mathbf{x}}) = 0.$$

Figure: Consumer's problem on a budget line

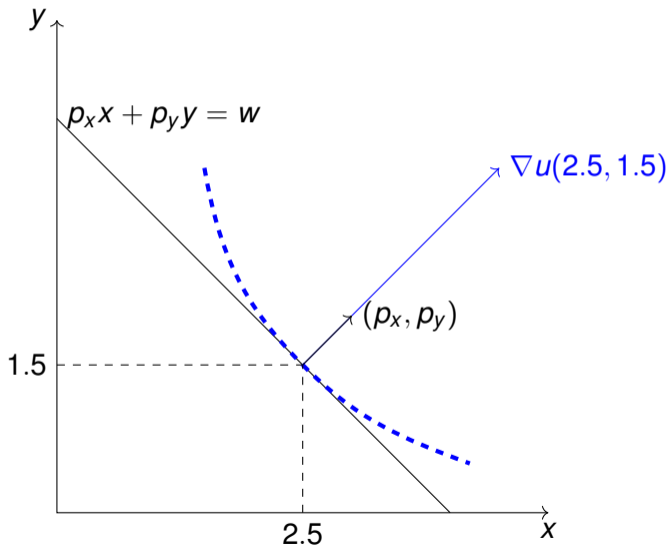
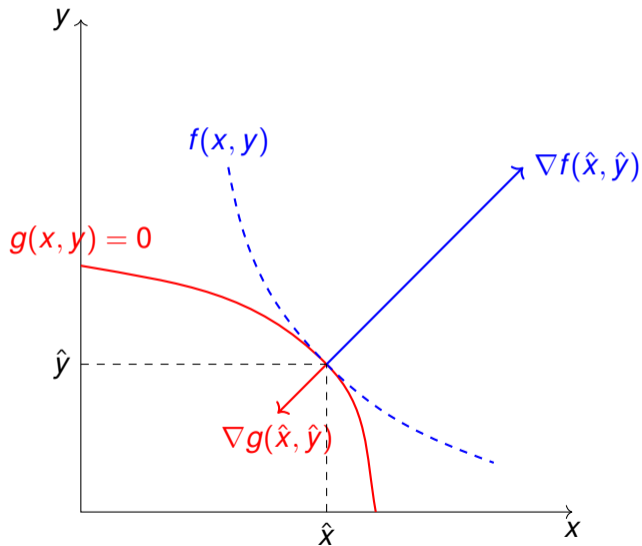


Figure: Single equality constraint



Optimization with a single equality constraint

Find the minima and maxima of $f(x, z) = x + z^2$ subject to constraints

$$x^2 + z^2 = 1. \quad (1)$$

- ▶ Form the Lagrangean:

$$\mathcal{L}(x, z, \mu) = x + z^2 - \mu(x^2 + z^2 - 1). \quad (2)$$

- ▶ Differentiate to get the first-order conditions (FOC):

$$\frac{\partial \mathcal{L}}{\partial x} = 1 - 2\mu x = 0, \quad (3)$$

$$\frac{\partial \mathcal{L}}{\partial y} = 2z - 2\mu z = 0, \quad (4)$$

$$\frac{\partial \mathcal{L}}{\partial \mu} = 1 - x^2 - z^2 = 0. \quad (5)$$

Optimization with a single equality constraint

- ▶ The second FOC gives:

$$z(2 - 2\mu) = 0$$

- ▶ therefore either $z = 0$, or $\mu = 1$. Consider first the possibility that $z = 0$. In that case, (5) implies that $x = \pm 1$. We get two critical points from (3):

$$(x = 1, z = 0, \mu = \frac{1}{2}) \text{ and } (x = -1, z = 0, \mu = -\frac{1}{2})$$

- ▶ If $\mu = 1$, (3) implies that $x = \frac{1}{2}$. By substituting into (5) we get the critical points:

$$(x = \frac{1}{2}, z = \frac{\sqrt{3}}{2}, \mu = 1) \text{ and } (x = \frac{1}{2}, z = -\frac{\sqrt{3}}{2}, \mu = 1)$$

Optimization with a single equality constraint

- ▶ As a result, we have four critical points for the Lagrangean. Draw the constraint set and level curves for the objective function to get a guess of the classification of the critical points.
- ▶ Can you show the existence of a maximum? Which of the local maxima is the global maximum?

Optimization with multiple equality constraints

Consider next the case, where we have k equality constraints $g(\mathbf{x}) = (g_1(\mathbf{x}), \dots, g_k(\mathbf{x})) : \mathbb{R}^n \rightarrow \mathbb{R}^k$. In this case, we have the problem:

$$\max_{\mathbf{x}} f(\mathbf{x})$$

$$\text{subject to } g_1(\mathbf{x}) = 0,$$

$$g_2(\mathbf{x}) = 0,$$

\vdots

$$g_k(\mathbf{x}) = 0.$$

Form the Lagrangean now with k constraints as a function of $n + k$ variables:

$$\mathcal{L}(\mathbf{x}, \mu_1, \dots, \mu_k) = f(\mathbf{x}) - \sum_{j=1}^k \mu_j g_j(\mathbf{x}).$$

Optimization with multiple equality constraints

We can proceed exactly as before to use the linear approximations to characterize the feasible directions from $\hat{\mathbf{x}}$ as $\{(\mathbf{x} - \hat{\mathbf{x}}) : Dg(\hat{\mathbf{x}})(\mathbf{x} - \hat{\mathbf{x}})\} = 0$.

Since the objective function cannot increase at the optimum in any feasible direction, we have that

$$Df(\hat{\mathbf{x}})(\mathbf{x} - \hat{\mathbf{x}}) = 0 \text{ whenever } Dg(\hat{\mathbf{x}})(\mathbf{x} - \hat{\mathbf{x}}) = 0.$$

If $Dg(\hat{\mathbf{x}})$ has full rank, then this is equivalent to requiring that $Df(\hat{\mathbf{x}})$ and $Dg_j(\hat{\mathbf{x}})$ must be linearly dependent. Since we assume that $Dg(\hat{\mathbf{x}})$ has full rank, this means that there must exist (μ_1, \dots, μ_k) such that

$$\nabla f(\hat{\mathbf{x}}) = \sum_{j=1}^k \mu_j \nabla g_j(\hat{\mathbf{x}}).$$

Optimization with multiple equality constraints

Hence we can summarize the three necessary conditions for local maximum:

i) Gradient alignment: $\nabla f(\hat{\mathbf{x}}) = \sum_{j=1}^k \mu_j \nabla g_j(\hat{\mathbf{x}}),$

ii) Constraint holds: $g(\hat{\mathbf{x}}) = 0,$

iii) Constraint qualification: $Dg_1(\hat{\mathbf{x}}), \dots, Dg_k(\hat{\mathbf{x}})$ are linearly independent.

The first two can be achieved by requiring that $(\hat{\mathbf{x}}, \hat{\mu}_1, \dots, \hat{\mu}_k)$ be a critical point of the Lagrangean.

Optimization with multiple equality constraints: an example

- ▶ Consider the problem of maximizing

$$f(x, y, z) = xz + yz$$

subject to:

$$g_1(x, y, z) = y^2 + z^2 - 1$$

$$g_2(x, y, z) = xz - 3$$

1. Find the critical points of f subject to constraints $g_1(x, y, z) = 0$ and $g_2(x, y, z) = 0$.
2. How would you determine which of the critical points are local minima and which are local maxima?
3. What about constraint qualification?

Optimization with multiple equality constraints: an example

1. Find first the critical points of the Lagrangean

$$\mathcal{L}(x, y, z, \mu_1, \mu_2) = xz + yz - \mu_1(y^2 + z^2 - 1) - \mu_2(xz - 3)$$

2. First-order conditions:

$$\frac{\partial \mathcal{L}}{\partial x} = z - \mu_2 z = 0 \quad (6)$$

$$\frac{\partial \mathcal{L}}{\partial y} = z - 2\mu_1 y = 0 \quad (7)$$

$$\frac{\partial \mathcal{L}}{\partial z} = x + y - 2\mu_1 z - \mu_2 x = 0 \quad (8)$$

$$\frac{\partial \mathcal{L}}{\partial \mu_1} = y^2 + z^2 - 1 = 0 \quad (9)$$

$$\frac{\partial \mathcal{L}}{\partial \mu_2} = xz - 3 = 0 \quad (10)$$

Optimization with multiple equality constraints: an example

We need to solve this system of equations to find the critical points. Start with (6), giving

$$z(1 - \mu_2) = 0, \Leftrightarrow z = 0 \text{ or } \mu_2 = 1$$

If $z = 0$, then (10) is not true for any x and as a result, we must have $z \neq 0$. Therefore, we can only have $\mu_2 = 1$ as a candidate solution. The second FOC (7) gives

$$y - 2\mu_1 z = 0, \Leftrightarrow y = \frac{z}{2\mu_1}.$$

Optimization with multiple equality constraints: an example

Plug in the solutions for y and μ_2 into (8) :

$$\frac{z}{2\mu_1} - 2\mu_1 z = 0 \Leftrightarrow z \left(\frac{1}{2\mu_1} - 2\mu_1 \right) = 0.$$

We already know that $z \neq 0$, and therefore

$$\frac{1}{2\mu_1} - 2\mu_1 = 0 \Leftrightarrow 4\mu_1^2 = 1 \Leftrightarrow \mu_1 = \pm \frac{1}{2}$$

Optimization with multiple equality constraints: an example

We have now solved for possible Lagrange multipliers μ_1 ja μ_2 , i.e. we have:

$$\mu_1 = \pm \frac{1}{2} \text{ and } \mu_2 = 1$$

Optimization with multiple equality constraints: an example

To get the values of the choice variables, plug in the values of the multipliers into (8) to get:

$$y = \pm z.$$

Substituting into (9), we get (by squaring):

$$2z^2 - 1 = 0 \Leftrightarrow z = \pm \frac{1}{\sqrt{2}}$$

The fifth FOC (10) gives:

$$x = \frac{3}{z},$$

or $x = 3\sqrt{2}$ if $z = \frac{1}{\sqrt{2}}$ and $x = -3\sqrt{2}$ if $z = -\frac{1}{\sqrt{2}}$. We have now found all that we need for the critical points of f subject to the constraints.

Optimization with multiple equality constraints: an example

If $z = \frac{1}{\sqrt{2}}$, then $x = 3\sqrt{2}$, $y = \pm z$. This yields two critical points (x, y, z) :

$$1 : \left(3\sqrt{2}, \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$

$$2 : \left(3\sqrt{2}, -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$

If $z = -\frac{1}{\sqrt{2}}$, then $x = -3\sqrt{2}$, $y = \pm z$. This gives also two critical points (x, y, z) :

$$3 : \left(-3\sqrt{2}, -\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right)$$

$$4 : \left(-3\sqrt{2}, \frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right)$$

Optimization with multiple equality constraints: an example

We know that for all critical points, $\mu_2 = 1$, and we can check the sign of μ_1 from FOC (7). After this, we have all the critical points of the problem as:

Critical points for the problem (x, y, z, μ_1, μ_2) :

$$1 : \left(3\sqrt{2}, \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, \frac{1}{2}, 1 \right)$$

$$2 : \left(3\sqrt{2}, -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, -\frac{1}{2}, 1 \right)$$

$$3 : \left(-3\sqrt{2}, -\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, \frac{1}{2}, 1 \right)$$

$$4 : \left(-3\sqrt{2}, \frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, -\frac{1}{2}, 1 \right)$$

Next Lecture

- ▶ Optimization with inequality constraints
- ▶ Economic examples of constrained optimization