

Economics 245

Lecture 1: Differentiability

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Leading Examples

- ▶ Find a firm's supply function:

$$\max_{y \geq 0} py - c(y).$$

- ▶ Find a firm's input demand functions:

$$\max_{x_1, x_2 \geq 0} pf(x_1, x_2) - w_1x_1 - w_2x_2.$$

Derivatives

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The Usual Loose Definition. The “derivative” of a function f at a point a is the slope of the tangent line.

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Definition (Better)

A function $f : \mathbb{R} \rightarrow \mathbb{R}$ is **differentiable** at a iff it is “very well approximated” near a by a linear function of the form

$$g(x) = f(a) + (x - a)b.$$

The number b , the constant slope of this approximating function, is called the **derivative** of f at a . It is denoted as $f'(a)$ or $\frac{df(a)}{dx}$ or $Df(a)$.

What does **very well approximated** mean?

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“Very Well Approximated”

- ▶ Two ways in which f may be approximated at a by $g(x) = f(a) + (x - a)b$:

Way 1. $f(a) = g(a)$.

Way 2. $\lim_{x \rightarrow a} [f(x) - g(x)] = 0$.

- ▶ These ways are not good enough. E.g., whether f is approximated by g in either way **does not depend on the value of b !**

The stronger notion of approximation we need is that as $x \rightarrow a$, the difference $f(x) - g(x)$ goes to zero so fast that

$$\text{Way 3. } \lim_{x \rightarrow a} \frac{f(x) - g(x)}{x - a} = 0.$$

Since $g(x) = f(a) + (x - a)b$, this can be written as

$$\lim_{x \rightarrow a} \frac{f(x) - f(a)}{x - a} = b,$$

the usual beginner's definition of a derivative.

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Definition (Proper)

A function $f : \mathbb{R} \rightarrow \mathbb{R}$ is **differentiable** at a iff the limit,

$$\lim_{x \rightarrow a} \frac{f(x) - f(a)}{x - a},$$

exists (i.e., yields the same number regardless of how x converges to a). When it exists, the limit is called the **derivative** of f at a .

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Example

$$f(x) = x^2.$$

We of course know that $f'(a) = 2a$.

Here's the proof:

$$\begin{aligned}\frac{f(x) - f(a)}{x - a} &= \frac{x^2 - a^2}{x - a} \\ &= \frac{(x - a)(x + a)}{x - a} \\ &= x + a \\ &\rightarrow 2a \text{ as } x \rightarrow a. \blacksquare\end{aligned}$$

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First-Order Expressions

- ▶ Given the differentiable f , the point a , and the approximating function

$$g(x) = f(a) + f'(a)(x - a),$$

we now change the variable from x to $h = x - a$, the “deviation” of x away from a .

- ▶ We also define the **remainder** function, $R(h; a)$, to be the difference between f and g at the point $x = a + h$:

$$\begin{aligned} R(h; a) &= f(a + h) - g(a + h) \\ &= f(a + h) - [f(a) + f'(a)h] . \end{aligned}$$

- ▶ Hence, $f(a+h)$ is the sum of the linear approximation and the remainder:

$$f(a+h) = f(a) + f'(a)h + R(h;a).$$

- ▶ Because f is differentiable, $R(h;a)/h \rightarrow 0$ as $h \rightarrow 0$.
- ▶ This is the meaning of the **first-order expression**,

$$f(a+h) \approx f(a) + f'(a)h.$$

In the texts you will see a first-order approximation written as

$$\Delta f \approx f'(a) \Delta x \quad (\text{or } \Delta y \approx f'(a) \Delta x),$$

which is the same as on the previous slide, with $\Delta f = f(a+h) - f(a)$ being the change in f and $\Delta x = h$ being the change in x .

You will also see this in differential terms:

$$df = f'(a) dx \quad (\text{or } dy = f'(a) dx).$$

This expression refers to movements along the tangent line, i.e., actually to changes in g rather than f .

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Differentiable Monotonicity

Key Fact: If f is differentiable at a and the approximating linear function g is strictly increasing (decreasing), then f is strictly increasing (decreasing) in a neighborhood of a .

This is geometrically, “intuitively,” obvious.

Formal statement, for the strictly increasing case:

Theorem

Suppose $f : \mathbb{R} \rightarrow \mathbb{R}$ is differentiable at a , with $f'(a) > 0$. Then $\delta > 0$ exists such that for any h satisfying $|h| < \delta$,

$$f(a + h) > f(a) \text{ if } h > 0,$$

$$f(a + h) < f(a) \text{ if } h < 0.$$

Proof.

- ▶ By definition of the remainder, for $h \neq 0$ we have

$$\begin{aligned} f(a+h) - f(a) &= f'(a)h + R(h;a) \\ &= \left[f'(a) + \frac{R(h;a)}{h} \right] h. \end{aligned}$$

- ▶ Since f is differentiable at a , $R(h;a)/h \rightarrow 0$ as $h \rightarrow 0$. Thus, since $f'(a) > 0$, there exists $\delta > 0$ such that

$$0 < |h| < \delta \quad \implies \quad f'(a) + \frac{R(h;a)}{h} > 0.$$

- ▶ So, for any h satisfying $|h| < \delta$, $f(a+h) - f(a)$ and h have the same sign. This proves the theorem.



Consequence for Optimization

Corollary

*If f is defined on an open interval, and it is maximized or minimized on this interval at a point a at which f is differentiable, then it satisfies the following necessary **first-order condition**:*

$$\text{(FOC)} \quad f'(a) = 0.$$

So if we want to find the maximizer of a differentiable function f on an interval $[\underline{x}, \bar{x}]$, we can restrict attention to the endpoints, \underline{x} and \bar{x} , and to the solutions of the equation $f'(x) = 0$, the **critical points** of f .

Examples

1. Maximize $f(x) = 6x - x^2$ on $[0, \infty)$.

- ▶ Obviously 0 is not a maximizer. So any maximizer x^* is interior, and so satisfies

$$f'(x^*) = 6 - 2x^* = 0.$$

The only possible maximizer is therefore 3. If we knew from a separate argument that a maximizer exists, we would now know that it is 3.

- ▶ The following is a direct proof that 3 is the maximizer:
 - ▶ $f'(x) > 0$ for any $x < 3$.
 - $\Rightarrow f$ is strictly increasing on $[0, 3)$.
 - $\Rightarrow f(3) > f(x)$ for all $x < 3$.
 - ▶ $f'(x) < 0$ for any $x > 3$.
 - $\Rightarrow f$ is strictly decreasing on $(3, \infty)$.
 - $\Rightarrow f(3) > f(x)$ for all $x > 3$.

Examples

2. Maximize $6x - x^2$ on $[0, 2]$.

▶ Answer: $x^* = 2$, a corner. Note that $f'(x^*) = 2 \neq 0$.

3. Maximize $6x - x^2$ on $[0, 2)$.

▶ Answer: no solution.

4. Minimize $6x - x^2$ on $[0, \infty)$.

▶ Answer: No solution.

Examples

5. Maximize and minimize $f(x) = \sin x$ on $[0, 2\pi]$.

- Answer: maximizer is $\pi/2$, minimizer is $3\pi/2$.

Proof. We know $f'(x) = \cos x$, the zeros of which are $\pi/2$ and $3\pi/2$. So the maximizer(s) lie in the set $\{0, \pi/2, 3\pi/2, 2\pi\}$. Checking each of these points yields $f(x) = 0, 1, -1$, and 0 , respectively.

6. Maximize $f(x) = x^3$ on \mathbb{R} .

- Answer: no solution.

Proof. There are no corners, and only one critical point:

$$f'(x) = 3x^2 = 0 \iff x = 0.$$

But 0 does not maximize x^3 .

Multivariable Differentiability

A function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is **differentiable** at $a \in \mathbb{R}^n$ iff numbers $f_1(a), \dots, f_n(a)$ exist such that

$$\lim_{h \rightarrow 0} \frac{R(h; a)}{\|h\|} \rightarrow 0,$$

where $R(\cdot; a)$ is the **remainder** function defined at any $h \in \mathbb{R}^n$ by

$$f(a + h) = f(a) + \sum_{i=1}^n f_i(a) h_i + R(h; a).$$

So, f is differentiable at a iff it is approximated near a by the linear function

$$g(x) = f(a) + \sum_{i=1}^n f_i(a) (x_i - a_i).$$

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- ▶ The numbers $f_i(a)$ are the **partial derivatives** of f at a , and are sometimes written as $\frac{\partial f}{\partial x_i}(a)$ or $D_i f(a)$.

- ▶ The **gradient** of f at a is the (column) vector of partial derivatives:

$$\nabla f(a) = \begin{pmatrix} f_1(a) \\ \vdots \\ f_n(a) \end{pmatrix}.$$

The gradient at a is sometimes denoted as $Df(a)$.

Unconstrained Multivariable Optimization

Theorem

If $x^* \in \mathbb{R}^n$ is a maximizer or a minimizer of a differentiable function $f : \mathbb{R}^n \rightarrow \mathbb{R}$, then it satisfies the following necessary *first-order condition*:

$$\text{(FOC)} \quad \nabla f(x^*) = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}.$$

(We shall usually write the FOC as $\nabla f(x^*) = 0$, where it is understood that 0 is the n -vector of zeros.)

Proof. Same as for one variable. If $f_i(x^*) \neq 0$, consider deviations h such that $h_j = 0$ for all $j \neq i$.