

# ECON8000: Quantitative Skills for Economics

## Lecture 6: Functions

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February 2020

# Functions: The Environment

We will study functions  $f : S \rightarrow \mathbb{R}$  where  $S \subset \mathbb{R}$  will usually be either open or compact. Of course such a function is *continuous* if  $f^{-1}(U)$  is an open subset of  $S$  whenever  $U \subset \mathbb{R}$  is open.

# The Intermediate Value Theorem

**Theorem:** If  $f : [a, b] \rightarrow \mathbb{R}$  is continuous and  $f(a) \leq f(b)$  ( $f(b) \leq f(a)$ ), then  $[f(a), f(b)] \subset f([a, b])$  ( $[f(b), f(a)] \subset f([a, b])$ ).

# Differentiation

Let  $I \subset \mathbb{R}$  be an open interval, and let  $f : I \rightarrow \mathbb{R}$  be a function. It is *differentiable* at  $x \in I$  if

$$\lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

exists, in which case the limit is called the *derivative* of  $f$  at  $x$ , and is denoted by  $f'(x)$  or  $f^{(1)}(x)$ . We say that  $f$  is *continuously differentiable* if it is differentiable and  $f' : I \rightarrow \mathbb{R}$  is continuous. Inductively, for  $k \geq 2$  we say that  $f$  is  *$k$  times continuously differentiable* if it is  $k - 1$  times continuously differentiable and  $f^{(k-1)}$  is continuously differentiable, in which case its derivative is denoted by  $f^{(k)}$ . If  $f$  is  $k$  times continuously differentiable for every  $k$  it is  $C^\infty$ .

# The Chain Rule

**Theorem:** If  $f : (a, b) \rightarrow (c, d)$  is differentiable at  $t$  and  $g : (c, d) \rightarrow \mathbb{R}$  is differentiable at  $f(t)$ , then

$$(g \circ f)'(t) = g'(f(t)) \cdot f'(t).$$

**Proof.** Clearly  $(g \circ f)'(t)$  and  $f'(t)$  are zero if  $f$  is constant on a neighborhood of  $t$ , and otherwise

$$\begin{aligned} & \lim_{h \rightarrow 0} \frac{g(f(t+h)) - g(f(t))}{h} \\ &= \lim_{h \rightarrow 0} \frac{g(f(t+h)) - g(f(t))}{f(t+h) - f(t)} \cdot \frac{f(t+h) - f(t)}{h} = g'(f(t)) \cdot f'(t). \end{aligned}$$

# The Product Rule

**Theorem:** If  $f, g : (a, b) \rightarrow \mathbb{R}$  are continuous and differentiable at  $t$ , then

$$(fg)'(t) = f'(t)g(t) + f(t)g'(t).$$

**Proof.** We have

$$\begin{aligned}\lim_{h \rightarrow 0} \frac{f(t+h)g(t+h) - f(t)g(t)}{h} &= \lim_{h \rightarrow 0} \frac{f(t+h) - f(t)}{h} \cdot g(t+h) \\ &\quad + \lim_{h \rightarrow 0} f(t) \cdot \frac{g(t+h) - g(t)}{h}.\end{aligned}$$

# The Mean Value Theorem

**Theorem:** If  $f : (a, b) \rightarrow \mathbb{R}$  is differentiable at  $t$  and  $f(t) = \max_s f(s)$  or  $f(t) = \min_s f(s)$ , then  $f'(t) = 0$ .

**Proof.** If  $f(t) = \max_s f(s)$ , then

$$\lim_{h \rightarrow 0^-} \frac{f(t+h) - f(t)}{h} \geq 0 \geq \lim_{h \rightarrow 0^+} \frac{f(t+h) - f(t)}{h}.$$

The proof when  $f(t) = \min_s f(s)$  is similar.

## Rolle's Theorem

**Rolle's Theorem:** If  $f : [a, b] \rightarrow \mathbb{R}$  is continuous and differentiable at every point in  $(a, b)$ , and  $f(a) = f(b)$ , then there is a  $t \in (a, b)$  such that  $f'(t) = 0$ .

**Proof.** Since  $[a, b]$  is compact there are  $t_-, t_+$  such that  $f(t_-) = \min_s f(s)$  and  $f(t_+) = \max_s f(s)$ . The last result gives the desired conclusion when  $t_- \in (a, b)$  or  $t_+ \in (a, b)$ . If  $t_-, t_+ \in \{a, b\}$ , then  $f(t) = f(a)$  for all  $t$  and  $f'(t) = 0$  for all  $t \in (a, b)$ .



In applications the most convenient form of Rolle's theorem is:

**Mean Value Theorem:** If  $f : [a, b] \rightarrow \mathbb{R}$  is continuous and differentiable at every point in  $(a, b)$ , then there is a  $t \in (a, b)$  such that

$$f'(t) = \frac{f(b) - f(a)}{b - a}.$$

**Proof.** Let  $g : [a, b] \rightarrow \mathbb{R}$  be the function

$$g(t) = f(t) - \frac{f(b) - f(a)}{b - a}(t - a).$$

Applying Rolle's theorem to  $g$  gives a  $t$  such that

$$0 = g'(t) = f'(t) - \frac{f(b) - f(a)}{b - a}.$$

# Maximization

If  $f : (a, b) \rightarrow \mathbb{R}$  is a function, a point  $t$  is a *strict local maximum* if there is an  $\delta > 0$  such that  $f(t) > f(s)$  for all  $s \in (t - \delta, t) \cup (t, t + \delta)$ .

**Theorem:** Let  $f : (a, b) \rightarrow \mathbb{R}$  be  $C^2$ , and let  $t$  be a point in  $(a, b)$ . If  $t$  is a local maximum, then  $f'(t) = 0$  and  $f''(t) \leq 0$ . If  $f'(t) = 0$  and  $f''(t) < 0$ , then  $t$  is a local maximum.

# The Lagrangean

The following result may seem a bit silly, but the one dimensional case is our model for higher dimensional problems, so we should be sure to understand it.

**Theorem:** Suppose that  $a < 0 < b$  and  $u : (a, b) \rightarrow \mathbb{R}$  is  $C^1$ .  
Let

$$\mathcal{L}(t; \lambda) = u(t) + \lambda t.$$

If  $t^*$  solves that problem of maximizing  $u(t)$  subject to  $t \geq 0$ , then there is a  $\lambda^*$  such that  $\frac{\partial \mathcal{L}}{\partial t}(t^*; \lambda^*) = 0$ . In addition,  $\lambda^* \geq 0$  and  $\lambda^* t^* = 0$ . (This final condition is called *dual slackness*.)

**Proof.** If  $t^* > 0$ , then  $u'(t^*) = 0$  and we can set  $\lambda^* = 0$ . If  $t^* = 0$ , then  $u'(0) \leq 0$ , and we can set  $\lambda^* = -u'(0)$ .

# Convex and Concave Functions

Let  $I \subset \mathbb{R}$  be an interval. A function  $f : I \rightarrow \mathbb{R}$  is *convex* (*concave*) if, for all  $s, t \in I$  and  $\alpha \in [0, 1]$ ,

$$f((1 - \alpha)s + \alpha t) \leq (\geq) (1 - \alpha)f(s) + \alpha f(t).$$

We will only discuss convex functions, but everything we say applies equally to concave functions, with obvious modifications. We say that  $f$  is *strictly convex* if the inequality above holds strictly whenever  $s \neq t$  and  $\alpha \in (0, 1)$ .

## Theorem: Continuous Convex Function

**Theorem:** If  $I \subset \mathbb{R}$  is an open interval and  $f : I \rightarrow \mathbb{R}$  is convex, then  $f$  is continuous.

**Proof.** Fix  $a, b \in I$  with  $a < b$  and  $\delta > 0$  such that  $a - \delta, b + \delta \in I$ . Let  $M$  and  $m$  be upper and lower bounds for  $f$  on  $[a - \delta, b + \delta]$ . If  $a \leq s < t \leq b$ , then

$$\begin{aligned} f(t) &\leq \frac{\delta}{t-s+\delta} f(s) + \frac{t-s}{t-s+\delta} f(t+\delta) = f(s) + \frac{t-s}{t-s+\delta} (f(t+\delta) - f(s)) \\ &\leq f(s) + \frac{M-m}{\delta} (t-s) \end{aligned}$$

and

$$\begin{aligned} f(s) &\leq \frac{t-s}{t-s+\delta} f(s-\delta) + \frac{\delta}{t-s+\delta} f(t) = f(t) + \frac{t-s}{t-s+\delta} (f(s-\delta) - f(t)) \\ &\leq f(t) + \frac{M-m}{\delta} (t-s) \end{aligned}$$

Therefore

$$|f(t) - f(s)| \leq \frac{M-m}{\delta} (t-s).$$

## Lemma: Convex Function

**Lemma:** If  $f : I \rightarrow \mathbb{R}$  is convex and  $a, b \in I$ , then there are  $m$  and  $M$  such that  $m \leq f(t) \leq M$  for all  $t \in [a, b]$ .

**Proof.** The claim is trivial when  $a = b$ , so without loss of generality assume that  $a < b$ .

- ▶ Fix  $s \in (a, b)$ .
- ▶ For  $t \in (a, s)$  we have  $f(s) \leq \frac{b-s}{b-t}f(t) + \frac{s-t}{b-t}f(b)$  and thus  $f(t) \geq \frac{b-t}{b-s}f(s) - \frac{s-t}{b-s}f(b)$ .
- ▶ Similarly, for  $t \in (s, b)$  we have  $f(t) \geq \frac{t-a}{s-a}f(s) - \frac{t-s}{s-a}f(a)$ .

# Jensen's Inequality

**Theorem:** If  $I \subset \mathbb{R}$  is an interval,  $f : I \rightarrow \mathbb{R}$  is a convex function,  $k \geq 2$  is an integer,  $t_1, \dots, t_k \in I$ ,  $\alpha_1, \dots, \alpha_k > 0$ , and  $\sum_i \alpha_i = 1$ , then

$$f(\alpha_1 t_1 + \dots + \alpha_k t_k) \leq \alpha_1 f(t_1) + \dots + \alpha_k f(t_k).$$

## Jensen's Inequality: Proof

**Proof.** We argue by induction on  $k$ . When  $k = 2$  the asserted inequality is just the definition of convexity. Assuming the inequality has been established with  $k - 1$  in place of  $k$ , we apply the definition of convexity, then the induction hypothesis, to compute that

$$\begin{aligned} f(\alpha_1 t_1 + \cdots + \alpha_k t_k) &= f((1 - \alpha_k)(\frac{\alpha_1}{1 - \alpha_k} t_1 + \cdots + \frac{\alpha_{k-1}}{1 - \alpha_k} t_{k-1}) + \alpha_k t_k) \\ &\leq (1 - \alpha_k) f(\frac{\alpha_1}{1 - \alpha_k} t_1 + \cdots + \frac{\alpha_{k-1}}{1 - \alpha_k} t_{k-1}) + \alpha_k f(t_k) \\ &\leq (1 - \alpha_k)(\frac{\alpha_1}{1 - \alpha_k} f(t_1) + \cdots + \frac{\alpha_{k-1}}{1 - \alpha_k} f(t_{k-1})) + \alpha_k f(t_k) = \sum_i \alpha_i f(t_i). \end{aligned}$$