

Derivatives

Mark Dean

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I would guess that it would be almost impossible for you to get this far in your economics education without having a good intuitive (and probably quite good technical) understanding of how derivatives work, so this section will be very quick. First, a reminder of what a derivative is, and what we mean by a differentiable function on some open interval of \mathbb{R}

Definition 1 Let $f : (a, b) \rightarrow \mathbb{R}$. Define the quotient function $\phi(t)$ as

$$\phi(t) = \frac{f(t) - f(x)}{t - x}$$

We say that f is differentiable at $x \in (a, b)$, if $\lim_{t \rightarrow x} \phi(t)$ exists - in other words there exists some y such that, for every $\varepsilon > 0$ there exists some $\delta > 0$ such that $|t - x| < \delta$ implies that $|y - \phi(t)| < \varepsilon$. If this is the case, we define the derivative as $f'(x) = \lim_{t \rightarrow x} \phi(t)$

We say that a function is differentiable on (a, b) if it is differentiable at any $x \in (a, b)$. We say it is continuously differentiable if the function $f' : (a, b) \rightarrow \mathbb{R}$ is continuous. Such functions are belong to the class \mathcal{C}^1 . A function is twice (continuously) differentiable if $f' : (a, b) \rightarrow \mathbb{R}$ is (continuously) differentiable. Such functions belong to the class \mathcal{C}^2

If $f : X \rightarrow \mathbb{R}$ where X is an open cube in \mathbb{R}^n , then, at any $x = (x_1, x_2, \dots, x_n)$ we define the partial derivative with respect to x_i as

$$\frac{\partial f(x)}{\partial x_i} = \lim_{\varepsilon \rightarrow 0} \frac{f(x) - f(x + \varepsilon e_i)}{\varepsilon}$$

if such a limit exists, and define

$$\nabla f(x) = \begin{pmatrix} \frac{\partial f(x)}{\partial x_1} \\ \vdots \\ \frac{\partial f(x)}{\partial x_n} \end{pmatrix}$$

Obviously, the reason that we are here interested in derivatives is they tell us something about the slope of a function. In particular, we are going to be interested in how the derivative of a function can help us find local maxima and minima. For brevity we will deal with local maxima here, but local minima can be treated analogously.

Definition 2 Let $f : X \rightarrow \mathbb{R}$ where $X \subset \mathbb{R}^n$. x^* is a local maximizer of f if there exists $\varepsilon > 0$ such that $f(x^*) \geq f(x) \forall x \in B(x^*, \varepsilon) \cap X$. It is a strict local maximizer if $f(x^*) > f(x) \forall x \in B(x^*, \varepsilon) \cap X$ s.t. $x \neq x^*$

Intuitively, we know that if x^* is a local maximizer in the interior of the domain of a function, then it must have a derivative of 0.

Lemma 1 Let $f : (a, b) \rightarrow \mathbb{R}$ and $x^* \in (a, b)$ be a local maximizer, then $f'(x^*) = 0$

Proof. Note that, for any $\varepsilon > 0$,

$$\begin{aligned} \phi(x + \varepsilon) &= \frac{f(x + \varepsilon) - f(x)}{\varepsilon} \\ \Rightarrow f(x + \varepsilon) &= \varepsilon \phi(x + \varepsilon) + f(x) \\ &= f(x) + \varepsilon f'(x) + \varepsilon (\phi(x + \varepsilon) - f'(x)) \end{aligned}$$

As $f(x^*)$ is a local maximizer, $f(x + \varepsilon) \leq f(x^*)$ for ε small enough and so

$$\begin{aligned} f(x^*) + \varepsilon f'(x^*) + \varepsilon (\phi(x^* + \varepsilon) - f'(x^*)) &\leq f(x^*) \\ \Rightarrow f'(x^*) &\leq -(\phi(x^* + \varepsilon) - f'(x^*)) \end{aligned}$$

but as $-(\phi(x^* + \varepsilon) - f'(x^*)) \rightarrow 0$ as $\varepsilon \rightarrow 0$, this implies $f'(x^*) \leq 0$

A similar argument gives

$$\begin{aligned} f(x - \varepsilon) &= -\varepsilon \phi(x - \varepsilon) + f(x) \\ &= f(x) - \varepsilon f'(x) + \varepsilon (f'(x) - \phi(x - \varepsilon)) \end{aligned}$$

and so

$$\begin{aligned} f(x^*) - \varepsilon f'(x^*) + \varepsilon (f'(x^*) - \phi(x^* - \varepsilon)) &\leq f(x^*) \\ \Rightarrow f'(x^*) &\geq (f'(x^*) - \phi(x^* - \varepsilon)) \end{aligned}$$

giving $f'(x^*) \geq 0$ and so $f'(x^*) = 0$ ■

We can use this result to derive Rolle's theorem.

Theorem 1 (Rolle) Let $f : [a, b] \rightarrow \mathbb{R}$ be differentiable, and say $f(a) = f(b)$. Then $f'(c) = 0$ for some $c \in (a, b)$

Proof. As f is differentiable, it is continuous (see homework). This implies (by Weierstrass theorem) that there exists an $a \leq x, y \leq b$ such that $f(x) \leq f(t) \leq f(y) \forall t \in [a, b]$. If $\{x, y\} = \{a, b\}$ then f must be constant, and so $f'(t) = 0 \forall t \in [a, b]$. Otherwise, either $x \in (a, b)$ or $y \in (a, b)$, and by lemma 1 $f'(x) = 0$ or $f'(y) = 0$ ■

Another useful thing we can do with derivatives is use them to approximate function: as the derivative gives us the slope of a function at a particular point x , then we can approximate $f(x + \varepsilon)$ by $f(x) + \varepsilon f'(x)$. This is a Taylor series approximation. To make this precise, we are going to formally define the idea of an error being small:

Definition 3 $h : \mathbb{R}^n \rightarrow \mathbb{R}$ is 'little oh' of order k , which we denote as $h(x) = o(\|x\|^k)$ if

$$\lim_{x \rightarrow 0} \frac{h(x)}{\|x\|^k} = 0$$

Thus, if a function $h(x)$ is $o(\|x\|^k)$, then $h(x)$ gets small 'quickly', in the sense that it does so quicker than $\frac{1}{\|x\|^k}$ gets big.

Theorem 2 Let $f : [a, b] \rightarrow \mathbb{R}$ be \mathcal{C}^2 . Then for any $x, x + \varepsilon \in (a, b)$

$$f(x + \varepsilon) = f(x) + f'(x)\varepsilon + o(\varepsilon)$$

Proof. Note that this can be proved relatively easily from the definition of the derivative. Rearranging the above expression gives

$$\frac{f(x + \varepsilon) - f(x)}{\varepsilon} - f'(x) = \frac{o(\varepsilon)}{\varepsilon}$$

The limit of the left hand side equals zero, and therefore so does the limit of the right hand side.

A more long winded proof, which is useful as it can be generalized to prove higher order approximations, is as follows:

Let

$$g(t) = f(t) - [f(x) + f'(x)(t - x)] - M(t - x)^2$$

Where

$$M = \frac{1}{\varepsilon^2} [f(x + \varepsilon) - f(x) - f'(x)\varepsilon]$$

Note that

$$g(x + \varepsilon) = 0$$

$$g(x) = 0$$

Also note that

$$g'(t) = f'(t) - f'(x) - 2M(t - x)$$

and so $g'(x) = 0$

Applying Rolle's theorem tells us that there exists a $c_1 \in (x, x + \varepsilon)$ such that $g'(c_1) = 0$. Applying Rolle's theorem again tells us that there is a $c \in (x, c_1)$ such that $g''(c) = 0$,

As

$$g''(t) = f''(t) - 2M$$

This tells us that $f''(c) = 2M$ and so

$$\begin{aligned} f''(c) &= \frac{2}{\varepsilon^2} [f(x + \varepsilon) - f(x) - f'(x)\varepsilon] \\ \Rightarrow f(x + \varepsilon) &= f(x) + f'(x)\varepsilon + \frac{\varepsilon^2}{2} f''(c) \end{aligned}$$

Note here that c here is really a function of the ε we initially chose. But, as $f''(c)$ is bounded (as it is continuous on $[a, b]$), then

$$\frac{\frac{\varepsilon^2}{2} f''(c(\varepsilon))}{\varepsilon} = \frac{\varepsilon}{2} f''(c(\varepsilon))$$

tends to zero, giving the necessary result ■

An extension, which we will state but not prove, tells us that we can get an even better approximation if we also use a second derivative.

Theorem 3 *Let $f : [a, b] \rightarrow \mathbb{R}$ be \mathcal{C}^3 . Then for any $x, x + \varepsilon \in (a, b)$*

$$f(x + \varepsilon) = f(x) + f'(x)\varepsilon + \frac{1}{2}f''(x)\varepsilon^2 + o(\varepsilon^2)$$