

## What Are the Mean Value and Taylor Theorems Saying?

We have studied two propositions about the derivative of a function that sound vaguely alike.

- (1) On the one hand, the **mean value theorem** (Week 13, Stewart 3.2) says that

$$f(x) = f(a) + f'(c)(x - a) \quad (\text{exactly!})$$

for some  $c$  between  $a$  and  $x$ .

For example,  $e^x = 1 + e^c x$ .

- (2) On the other hand, the **best linear approximation** (Week 16, Stewart 2.9) says that

$$f(x) = f(a) + f'(a)(x - a) + \text{something small}$$

if  $x$  is close to  $a$ .

For example,  $e^x \approx 1 + x$  (that is,  $e^x = 1 + x + \text{a small error}$ ).

**Quick exercise:** What are the **two** crucial differences between (1) and (2)? Read closely.

An immediately interesting question about statement (2) is: **How** small? The word “small” by itself is too vague to be useful in technical work.

We can cast some light on the relation between (1) and (2) by stepping back a bit to the **definition of continuity**, which can be paraphrased this way:

- (0) If  $f$  is continuous, then

$$f(x) = f(a) + \text{something small}$$

if  $x$  is close to  $a$ .

For example,  $e^x \rightarrow e^0 = 1$  as  $x \rightarrow 0$ .

Again one should ask, “**How** small?” In this case the answer is given by the detailed definition of continuity (through the definition of a limit): For every  $\epsilon$  there is a  $\delta$  such that if  $|x - a| < \delta$ , then  $|f(x) - f(a)| < \epsilon$ . We will talk about statements like this in more depth in Week 26; for now the only thing to understand is that (0) merely says that as  $x - a$  goes to 0,

$f(x) - f(a)$  also goes to 0 — it says nothing about **how fast**  $f(x) - f(a)$  approaches 0.

But now look again at the mean value theorem (alias “Taylor’s theorem with  $N = 0$ ”): It says that **if  $f$  is differentiable**, then

$$f(x) = f(a) + f'(c)(x - a)$$

and so, if  $|x - a| < \delta$ , then\*

$$|f(x) - f(a)| < \delta \max_{c \in (a, x)} |f'(c)|.$$

The important thing about this inequality is that it says that the difference between  $f(x)$  and  $f(a)$  is (at worst) **linear** in  $\delta$ , the distance between  $x$  and  $a$ . Thus (1) is a sharpened form of (0), valid if  $f$  is differentiable, not just continuous.

But differentiability also gives us statement (2), and we want to know what “small” means in it. One type of answer is provided by Taylor’s theorem with  $N = 1$ :

(3) If the **second** derivative of  $f$  exists, then

$$f(x) = f(a) + f'(a)(x - a) + \frac{1}{2}f''(c)(x - a)^2$$

for some  $c$ .

For example,  $e^x = 1 + x + \frac{1}{2}e^c x^2$ .

Therefore, if  $|x - a| < \delta$ , then

$$|f(x) - [f(a) + f'(a)(x - a)]| < \frac{1}{2}\delta^2 \max |f''(c)|.$$

That is, the error is **quadratically** small in its dependence on  $\delta$ .

If  $f'(a)$  exists but  $f''(a)$  doesn’t, then it turns out that the error in the linear approximation (2) typically goes to zero faster than  $\delta$  but not as fast as  $\delta^2$ . (Contemplate, for example,  $f(x) = x^{4/3}$  around  $x = 0$ .) We shall not discuss this sort of situation further (leaving it to advanced courses in mathematics).

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\*  $c \in (a, x)$  means  $a < c < x$ , which is the right formulation if  $a < x$ ; if  $x < a$ , of course, we would have  $c \in (x, a)$ .

Of course, when  $f''$  exists we expect to form a **best quadratic approximation**:

(4) If  $x$  is close to  $a$ , then

$$f(x) \approx f(a) + f'(a)(x - a) + \frac{1}{2}f''(a)(x - a)^2.$$

For example,  $e^x \approx 1 + x + \frac{1}{2}x^2$ .

We can ask the usual question: Exactly **how** small is the error in the approximation (4)? By now you can probably guess the answer:

(5) If the **third** derivative exists, then (Taylor's theorem with  $N = 2$ )

$$f(x) = f(a) + f'(a)(x - a) + \frac{1}{2}f''(a)(x - a)^2 + \frac{1}{6}f^{(3)}(c)(x - a)^3$$

for some  $c$ , and therefore the error in the quadratic approximation is **cubic** in  $\delta$ .

Obviously we can continue this game forever; successfully completing each step invites us to attempt the next step. If  $f^{(N)}(a)$  exists, then we can construct the  $N$ th Taylor polynomial

$$T_N(x) = \sum_{j=0}^N \frac{f^{(j)}(a)}{j!} (x - a)^j$$

and expect it to be the **best approximation (near  $a$ ) to  $f(x)$  by a polynomial of degree  $N$** . (This is proposition number  $(2N)$ .) We ask how good this approximation really is. Then Taylor's theorem says that if  $f^{(N+1)}(a)$  exists, then

$$f(x) = T_N(x) + R_N(x),$$

with

$$R_N(x) = \frac{f^{(N+1)}(c)}{(N+1)!} (x - a)^{N+1},$$

and therefore the error in the  $N$ th-degree approximation vanishes as fast as  $\delta^{N+1}$  as  $\delta = |x - a|$  approaches 0. (This is proposition number  $(2N + 1)$ .) This formula tempts us to consider the even better approximation  $T_{N+1}$ , but we have to stop somewhere.