

Taylor Polynomials continued.

Given a smooth (all derivatives exist and are continuous) function $f: \mathbb{R} \rightarrow \mathbb{R}$,

the Taylor polynomial $P_n(x)$ of degree n at $x=a$ is the ^{degree n} polynomial that fits f best near $x=a$, and

$$\begin{aligned} P_n(x) &= f(a) + f'(a)(x-a) + \frac{f''(a)}{2!}(x-a)^2 + \\ &\quad \dots + \frac{f^{(n)}(a)}{n!}(x-a)^n \\ &= \sum_{j=0}^n \frac{f^{(j)}(a)}{j!} (x-a)^j \end{aligned}$$

Examples (If not mentioned, assume $a=0$ is the point in question)

$$e^x = 1 + x + \frac{x^2}{2!} + \dots + \frac{x^n}{n!}$$

$$\sin(x) = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \dots + \frac{(-1)^k x^{2k+1}}{(2k+1)!}$$

$$\cos(x) = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \dots + \frac{(-1)^k x^{2k}}{(2k)!}$$

$$(1+x)^p = 1 + px + \frac{p(p-1)}{2} x^2 + \frac{p(p-1)(p-2)}{3!} x^3 + \dots + \frac{p(p-1)(p-2)\dots(p-n+1)}{n!} x^n$$

Interesting tricks

① $f(x) = \frac{1}{2}(x-1)^2 + x^7 \leftarrow \begin{matrix} P_{23}(x) = \\ \text{degree 23 Taylor} \\ \text{polynomial at} \\ x = 17.4. \end{matrix}$

$$P_{23}(x) = \frac{1}{2}(x-1)^2 + x^7$$

② $f(x) = x^5 e^{x^7} \quad P_{25}(x) \leftarrow \begin{matrix} \text{degree 25 Taylor} \\ \text{poly at } x=0. \end{matrix}$

$$e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{6} + \dots$$

$$P_{25}(x) = x^5 \left(1 + (x^7) + \frac{(x^7)^2}{2} + \frac{(x^7)^3}{6} + \dots \right)$$

$$= x^5 \left(1 + x^7 + \frac{x^{14}}{2} + \frac{x^{21}}{6} + \frac{x^{28}}{24} + \dots \right)$$

$$= \boxed{x^5 + x^{12} + \frac{x^{19}}{2}}$$

Taylor-Lagrange Remainder formula

Idea

$$f(x) \approx P_n(x) = f(a) + f'(a)(x-a) + \dots + \frac{f^{(n)}(a)}{n!}(x-a)^n$$

exact approximation

What's the error?

If $f: \mathbb{R} \rightarrow \mathbb{R}$ is smooth near $x=a$,
then

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

where c is between a & x

(a specific # somewhere between a & x).

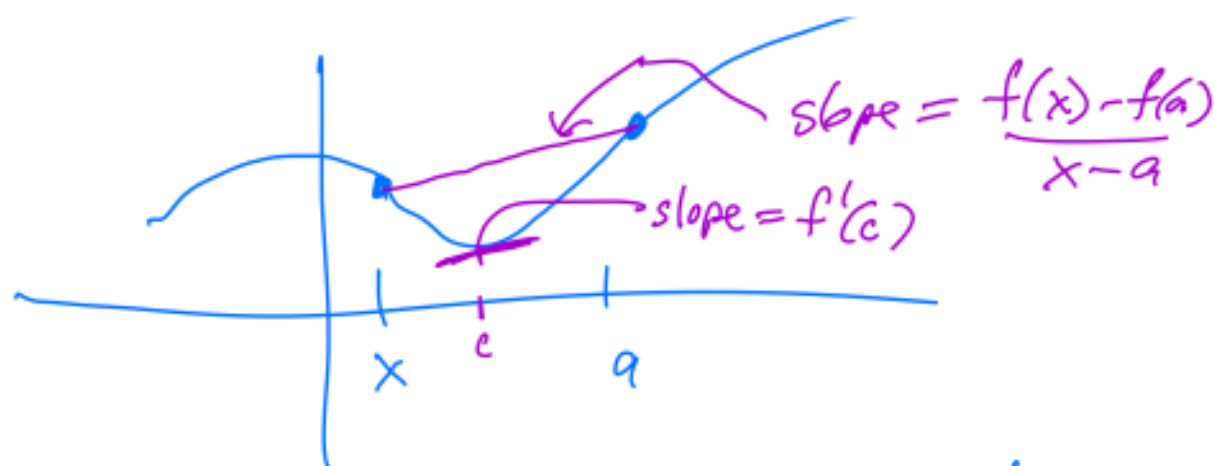
When $n=0$, this says

$$f(x) = f(a) + f^{(n+1)}(c)(x-a) \text{ for some } c$$

between a & x .

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

$$\Rightarrow f'(c) = \frac{f(x) - f(a)}{x-a} \quad \text{Mean Value Theorem.}$$



The Taylor-Lagrange Remainder is a higher degree version of the mean value theorem.

The theorem is useful to us, because it helps us estimate errors.

Example: Estimate $\cos(0.03)$ with the 4th degree Taylor polynomial. Estimate the error in this approximation.

Taylor for $\cos(x)$:

$$P_4(x) = 1 - \frac{x^2}{2} + \frac{x^4}{4!} = 1 - \frac{x^2}{2} + \frac{x^4}{24}$$

approx:

$$P_4(0.03) = 1 - \frac{(0.03)^2}{2} + \frac{(0.03)^4}{24}$$

$$x = 0.03 \quad a = 0 \quad = 1 - \frac{.0009}{2} + \frac{.00000081}{24}$$

$$= 1 - .00045 + .00000003375$$

$$\frac{27}{8} = 3\frac{3}{8}$$

$$= .99955 + .00000003375$$

$$P_4(0.03) = .99955003375 = \bar{x}$$

$$x = \cos(.03)$$

Error term

$$f(x) - P_4(x) = \frac{f^{(5)}(c)}{5!} (.03)^5$$

$$\text{where } 0 < c < 0.03$$

$$f(x) = \cos(x)$$

$$f'(x) = -\sin(x)$$

$$f''(x) = -\cos(x)$$

$$f'''(x) = \sin(x)$$

$$f^{(4)}(x) = \cos(x)$$

$$f^{(5)}(x) = -\sin(x)$$

$$|e_x| = |f(x) - P_4(x)| = \left| \frac{f^{(5)}(c)}{120} (.03)^5 \right|$$

$$= \left| \frac{-\sin(c) (.03)^5}{120} \right| \leq \frac{1(.03)^5}{120}$$

$$= \frac{.0000000081}{120 \cdot 40}$$

$$\frac{81}{40} = 2\frac{1}{6}$$

$$|e_x| \leq .000000002025$$

$$2.025$$

° Our estimate for $\cos(0.03)$ must be accurate at least to 9 decimal places

$$\cos(0.03) = \underline{.99955003375}$$

Application: If f is a smooth function, and we estimate x by $\bar{x}$, we can determine how accurate the approximation $f(\bar{x})$ is to $f(x)$.

$$e_{f(x)} = ?$$

$$e_{f(x)} = f(x) - f(\bar{x})$$

(Taylor-h formula)

$$\Rightarrow e_{f(x)} = f(\bar{x}) + f'(\bar{x})(x - \bar{x}) + \frac{f''(c)}{2}(x - \bar{x})^2 - f(\bar{x})$$

$$\Rightarrow e_{f(x)} = f'(\bar{x})(x - \bar{x}) + \frac{f''(c)}{2}(x - \bar{x})^2$$

$$\Rightarrow \boxed{e_{f(x)} = f'(\bar{x})e_x + \frac{f''(c)}{2}e_x^2}$$

Taylor @ $x = \bar{x}$

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

for some c between x & $\bar{x}$.

$$\Rightarrow |e_{f(x)}| = \left| f'(\bar{x})e_x + \frac{f''(c)}{2}e_x^2 \right|$$

$$\leq |f'(\bar{x})e_x| + \left| \frac{f''(c)}{2}e_x^2 \right|$$

$$= |f'(\bar{x})| |e_x| + \frac{|f''(c)|}{2} |e_x|^2$$

$$\Rightarrow |e_{f(x)}| \leq |f'(\bar{x})| |e_x| + \frac{|f''(c)|}{2} |e_x|^2$$

$$|e_{f(x)}| \approx |f'(\bar{x})| |e_x|$$

typically smaller
than $|e_x|$.

If we have a bound $|f''(c)| \leq K$, then
we can definitely say

$$|e_{f(x)}| \leq |f'(\bar{x})| |e_x| + \frac{K}{2} |e_x|^2$$

Relative Error

$$e_{f(x)} = f(\bar{x}) \mathcal{E}_{f(x)}$$

$$e_x = \bar{x} \mathcal{E}_x$$

$$e_{f(x)} = f'(\bar{x})e_x + \frac{f''(c)}{2}e_x^2$$

$$\Rightarrow f(\bar{x}) \mathcal{E}_{f(x)} = f'(\bar{x}) \bar{x} \mathcal{E}_x + \frac{f''(c)}{2} \bar{x}^2 \mathcal{E}_x^2$$

$$\text{assume } f(\bar{x}) \neq 0 \Rightarrow \varepsilon_{f(x)} = \frac{f'(\bar{x})}{f(\bar{x})} \bar{x} \varepsilon_x + \frac{f''(c)}{2f(\bar{x})} \bar{x}^2 \varepsilon_x^2$$

$$|\varepsilon_{f(x)}| = \left| \frac{f'(\bar{x})}{f(\bar{x})} \bar{x} \varepsilon_x + \frac{f''(c)}{2f(\bar{x})} \bar{x}^2 \varepsilon_x^2 \right|$$

$$\leq \left| \frac{f'(\bar{x})}{f(\bar{x})} \bar{x} \varepsilon_x \right| + \left| \frac{f''(c)}{2f(\bar{x})} \bar{x}^2 \varepsilon_x^2 \right|$$

$$= \left| \frac{f'(\bar{x})}{f(\bar{x})} \right| |\bar{x}| |\varepsilon_x| + \frac{|f''(c)|}{2|f(\bar{x})|} |\bar{x}|^2 |\varepsilon_x|^2$$

$$(\log(f(\bar{x})))' = \frac{f'(\bar{x})}{f(\bar{x})}$$

$$\Rightarrow |\varepsilon_{f(x)}| \leq \left(|(\log f)'(\bar{x})| |\bar{x}| \right) |\varepsilon_x| + \left(\frac{|f''(c)|}{2|f(\bar{x})|} |\bar{x}|^2 \right) |\varepsilon_x|^2$$

$$|\varepsilon_x|^2 < |\varepsilon_x|$$

$$\Rightarrow |\varepsilon_{f(x)}| \lesssim |(\log f)'(\bar{x})| |\bar{x}| |\varepsilon_x|$$

good bound.

$$\begin{aligned} (\log f)'(\bar{x}) &= \frac{f'(\bar{x})}{f(\bar{x})} \\ (\ln f)'(\bar{x}) &= [\ln(f(\bar{x}))]' \end{aligned}$$

This gives a general formula for the bound on $\varepsilon_{f(x)}$.

In text: the general formula for the error in $f(x, y, \dots)$ is derived.

Example: If $x = 2.581 \pm .002$, give the value of e^x along with the error bounds.

Solution: $\bar{x} = 2.581$
 $|e_x| \leq .002$ } given

from our error bounds,

$$|e_{f(x)}| \leq ?$$

$$f(x) = e^x$$

$$f'(x) = e^x$$

$$f''(x) = e^x$$

$$|e_{f(x)}| \leq |e^{\bar{x}}| \cdot .002 + \frac{|e^{\bar{x}}|}{2} (.002)^2$$

$$x \approx \bar{x} = 2.581 \pm .002 \leq 2.583$$

$$e^x \text{ increasing} \Rightarrow e^{\bar{x}} \leq e^{2.583}, e^c \leq e^{2.583}$$

If we have a bound $|f''(c)| \leq K$, then we can definitely say

$$|e_{f(x)}| \leq |f'(\bar{x})| |e_x| + \frac{K}{2} |e_x|^2$$

$$e^{2.583} = 13.2368 \dots$$

$$|e_{f(x)}| \leq (13.2368)(.002) + \frac{13.2368}{2} (.002)^2$$

$$|e_{f(x)}| \leq .0265 + .0000265$$

$$|e_{f(x)}| \leq .0265265$$

$$e^{2.581} = 13.2103$$

Thus

$$e^x = 13.2103 \pm .0266$$

In this example, we can more easily find the error difference

$$2.579 \leq x \leq 2.583$$

$$\Rightarrow e^{2.579} \leq e^x \leq e^{2.583}$$

(e^x is increasing $\Rightarrow$ it preserves inequalities)

$$13.1839 \leq e^x \leq 13.2368$$