

SECTION 3.2 TAYLOR'S THEOREM

Homework: 1, 2, 3, 4, 5, 6

A, B, C

REVIEW OF TAYLOR SERIES AND TAYLOR POLYNOMIALS
IN ONE VARIABLE

Suppose $f : \mathbf{R} \rightarrow \mathbf{R}$ and $a \in \mathbf{R}$. Suppose f and its derivatives of all orders are continuous. The *Taylor series* of f at a is given by

$$P(x, a) = \sum_{n=0}^{\infty} \frac{f^{(n)}(a)}{n!} (x - a)^n.$$

The *Taylor polynomial of order k* of f at a is given by

$$P_k(x, a) = \sum_{n=0}^k \frac{f^{(n)}(a)}{n!} (x - a)^n.$$

EXAMPLE 1: If $f(x) = e^x$ and $a = 0$, then for all n we have $f^{(n)}(x) = e^x$, and so $f^{(n)}(0) = 1$ and thus our n^{th} coefficient is $\frac{1}{n!}$, so we have

$$P(x, 0) = \sum_{n=0}^{\infty} \frac{x^n}{n!} = 1 + x + \frac{x^2}{2} + \frac{x^3}{3!} + \cdots + \frac{x^n}{n!} + \cdots$$

and

$$P_k(x, 0) = \sum_{n=0}^k \frac{x^n}{n!} = 1 + x + \frac{x^2}{2} + \frac{x^3}{3!} + \cdots + \frac{x^k}{k!}.$$

There are two basic issues to consider: For a fixed value of x what happens to $P_k(x, a)$ as $k \rightarrow \infty$? For a fixed value of k what happens to $P_k(x, a)$ as $x \rightarrow a$?

What one would hope is that as $k \rightarrow \infty$ one would have that $P_k(x, a) \rightarrow f(x)$. In this case we would say that $P(x, a)$ converges to $f(x)$, so that it would be legitimate to write $f(x) = P(x, a) = \sum_{n=0}^{\infty} \frac{f^{(n)}(a)}{n!} (x - a)^n$. For $f(x) = e^x$ it can be shown that this does indeed happen. However, this need not always be the case.

EXAMPLE 2: Let $f(x) = \ln(x)$ and $a = 1$. Then $f'(x) = x^{-1}$, $f''(x) = -x^{-2}$, $f^{(3)}(x) = 2x^{-3}$, $f^{(4)}(x) = -3 \cdot 2x^{-4}$, etc. In general, for $n \geq 1$ we have $f^{(n)}(x) = (-1)^{n+1}(n-1)!x^{-n}$, and so $\frac{f^{(n)}(1)}{n!} = \frac{(-1)^{n+1}(n-1)!}{n!} = \frac{(-1)^{n+1}}{n}$. For $n = 0$ we have $f^{(0)}(1) = f(1) = \ln(1) = 0$. Thus we have

$$P(x, 1) = \sum_{n=1}^{\infty} \frac{(-1)^{n+1}(x-1)^n}{n} = (x-1) - \frac{(x-1)^2}{2} + \frac{(x-1)^3}{3} - \frac{(x-1)^4}{4} + \cdots$$

It can be shown that if $0 < x \leq 2$, then this series will converge to $\ln(x)$. But look what happens for $x = 3$: We have that the n^{th} term of the series is $\frac{(-1)^{n+1}2^n}{n}$. By l'Hopital's Rule $\frac{2^n}{n} \rightarrow \infty$ as $n \rightarrow \infty$, so the series cannot converge to anything, much less to $\ln(3)$. More generally, the series doesn't converge at all when $x > 2$ even though $\ln(x)$ is defined for these values.

Suppose the Taylor series does converge. Does it converge to $f(x)$?

EXAMPLE 3: Let $f(x) = e^{-1/x^2}$ for $x \neq 0$ and set $f(0) = 0$. It can be shown that $f^{(n)}(0) = 0$ for all n , and hence $P(x, 0) = 0$. So the Taylor series converges to 0 for all x , but $f(x)$ is 0 only at $x = 0$. So the answer to our question is no.

Let's give a name to the difference between $f(x)$ and $P_k(x, a)$; we set $R_k(x, a) = f(x) - P_k(x, a)$. Thus $P_k(x, a)$ converging to $f(x)$ is equivalent to $R_k(x, a) \rightarrow 0$ as $k \rightarrow \infty$. There are several formulas for $R_k(x, a)$. One of them is the *Lagrange Remainder Formula* which says that there is a number t between x and a such that

$$R_k(x, a) = \frac{f^{(k+1)}(t)}{(k+1)!} (x-a)^{k+1}.$$

Note that the formula doesn't say how to find t , so it may appear pretty useless. However one can sometimes use it to estimate $P_k(x, a)$ well enough to show that $P_k(x, a)$ converges to $f(x)$. Consider $f(x) = e^x$ and $a = 0$ from Example 1. We have $R_k(x, 0) = \frac{e^t}{(k+1)!} x^{k+1}$. If $x \leq 0$, then $x \leq t \leq 0$, and so $e^x \leq e^t \leq e^0 = 1$. If $x \geq 0$, then $0 \leq t \leq x$ and so $1 = e^0 \leq e^t \leq e^x$. In either case this part of the formula won't go off to ∞ as $k \rightarrow \infty$. It can be shown that $\frac{x^{k+1}}{(k+1)!} \rightarrow 0$ as $k \rightarrow \infty$, so we have that $R_k(x, 0) \rightarrow 0$ as $k \rightarrow \infty$, and thus we can legitimately say that $e^x = 1 + x + \frac{x^2}{2} + \frac{x^3}{3!} + \dots$ for all x .

We now consider the second question. For a fixed value of k what happens as $x \rightarrow a$? Here the results are much more encouraging. Look at $R_k(x, a)$. Note that as $x \rightarrow a$ the mysterious number t between x and a is forced to go to a as well, so we have that

$$R_k(x, a) = \frac{f^{(k+1)}(t)}{(k+1)!} (x-a)^{k+1} \rightarrow \frac{f^{(k+1)}(a)}{(k+1)!} \cdot 0 = 0.$$

Note that we can actually get a stronger result here; not only does $R_k(x, a)$ go to zero, but it goes to zero faster than $(x-a)^k$ does.

$$\frac{R_k(x, a)}{(x-a)^k} = \frac{f^{(k+1)}(t)}{(k+1)!} (x-a) \rightarrow \frac{f^{(k+1)}(a)}{(k+1)!} \cdot 0 = 0.$$

This is called *Taylor's Theorem*.

Theorem 1 (Taylor's Theorem) Let $P_k(x, a) = \sum_{n=0}^k \frac{f^{(n)}(a)}{n!} (x - a)^n$.

Then $f(x) = P_k(x, a) + R_k(x, a)$, where $\frac{R_k(x, a)}{(x - a)^k} \rightarrow 0$ as $x \rightarrow a$.

For $k = 1$ note that $P_1(x, a) = f(a) + f'(a)(x - a)$, which is just the function giving the tangent line to the graph of f at $(a, f(a))$.

$$\lim_{x \rightarrow a} \frac{R_1(x, a)}{(x - a)} = \lim_{x \rightarrow a} \frac{f(x) - f(a) - f'(a)(x - a)}{x - a} = \lim_{x \rightarrow a} \frac{f(x) - f(a)}{x - a} - f'(a) = f'(a) - f'(a) = 0,$$

so in this case Taylor's Theorem is just the fact that $f'(a)$ exists.

Consider Example 1 again. For $k = 1$ we have $P_1(x, 0) = 1 + x$. Then

$$\lim_{x \rightarrow 0} \frac{R_1(x, 0)}{(x - 0)} = \lim_{x \rightarrow 0} \frac{e^x - 1 - x}{x} = \lim_{x \rightarrow 0} \frac{e^x - 1}{1} = \frac{e^0 - 1}{1} = \frac{0}{1} = 0$$

by l'Hopital's Rule. For $k = 2$ we have $P_2(x, 0) = 1 + x + \frac{x^2}{2}$. Then

$$\lim_{x \rightarrow 0} \frac{R_2(x, 0)}{(x - 0)^2} = \lim_{x \rightarrow 0} \frac{e^x - 1 - x - x/2}{x^2} = \lim_{x \rightarrow 0} \frac{e^x - 1 - x}{2x} = 0,$$

where we have applied l'Hopital's Rule twice.

TAYLOR'S THEOREM FOR SEVERAL VARIABLES

Now consider a function $f : \mathbf{R}^n \rightarrow \mathbf{R}$ and a point $\mathbf{a} \in \mathbf{R}^n$. We assume that f and its partial derivatives of all orders are continuous. The general formulas for Taylor series and Taylor polynomials require a lot of notation, so we will concentrate on the cases $k = 1$ and $k = 2$. In order to further simplify the notation we set $\mathbf{h} = \mathbf{x} - \mathbf{a}$. Note that $\mathbf{x} \rightarrow \mathbf{a}$ is equivalent to $\mathbf{h} \rightarrow \mathbf{0}$.

Consider $k = 1$. Recall that in Chapter 2 we defined a linear approximation $g(\mathbf{x})$ to $f(\mathbf{x})$. This will be $P_1(\mathbf{x}, \mathbf{a})$.

$$P_1(\mathbf{x}, \mathbf{a}) = g(\mathbf{x}) = f(\mathbf{a}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{a})(x_i - a_i) = f(\mathbf{a}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{a}) h_i.$$

Note that by using the gradient $\nabla f(\mathbf{a}) = \left(\frac{\partial f}{\partial x_1}(\mathbf{a}), \dots, \frac{\partial f}{\partial x_n}(\mathbf{a}) \right)$ we can rewrite this as

$$P_1(\mathbf{x}, \mathbf{a}) = f(\mathbf{a}) + \nabla f(\mathbf{a}) \cdot (\mathbf{x} - \mathbf{a}) = f(\mathbf{a}) + \nabla f(\mathbf{a}) \cdot \mathbf{h}.$$

For the case $n = 2$ we can write $x = x_1$, $y = x_2$, $a = a_1$, $b = a_2$, $h = h_1$, and $k = h_2$. Then this is just the function whose graph is the tangent plane to the graph of f at $(a, b, f(a, b))$.

$$P_1(\mathbf{x}, \mathbf{a}) = f(a, b) + \frac{\partial f}{\partial x}(a, b)(x - a) + \frac{\partial f}{\partial y}(a, b)(y - b) = f(a, b) + \frac{\partial f}{\partial x}(a, b)h + \frac{\partial f}{\partial y}(a, b)k.$$

As before we set $R_1(\mathbf{x}, \mathbf{a}) = f(\mathbf{x}) - g(\mathbf{x}) = f(\mathbf{x}) - P_1(\mathbf{x}, \mathbf{a})$. Recall that we defined $f(\mathbf{x})$ to be differentiable at \mathbf{a} if $\lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{f(\mathbf{x}) - g(\mathbf{x})}{\|\mathbf{x} - \mathbf{a}\|} = 0$. With our new notation this translates into saying $\lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{R_1(\mathbf{x}, \mathbf{a})}{\|\mathbf{x} - \mathbf{a}\|} = 0$. This is then the case $k = 1$ of Taylor's Theorem.

(The book, after introducing the notation $R_1(\mathbf{x}, \mathbf{a})$, rewrites it on page 183 as $R_1(\mathbf{h}, \mathbf{a})!$ They do apologize for doing this, but even so, this is way too confusing, so I'm going to stick with $R_1(\mathbf{x}, \mathbf{a})$. Just watch out for conflicting notation in the book.)

Note the following comparison with the one variable case. $f(a)$ is replaced by $f(\mathbf{a})$, and $f'(a)(x - a)$ is replaced by the sum of all the expressions of the form $\frac{\partial f}{\partial x_i}(\mathbf{a})(x_i - a_i)$. This gives us an idea of what the order two contribution should be for the case $k = 2$. Replace $\frac{1}{2}f^{(2)}(a)(x - a)^2$ by $\frac{1}{2}$ times the sum of all the expressions of the form $\frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a})(x_i - a_i)(y_j - a_j)$. This gives the Taylor polynomial $P_2(\mathbf{x}, \mathbf{a})$.

$$P_2(\mathbf{x}, \mathbf{a}) = f(\mathbf{a}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{a})(x_i - a_i) + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a})(x_i - a_i)(y_j - a_j) =$$

$$f(\mathbf{a}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{a})h_i + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a})h_i h_j.$$

There is also a shorthand version of this. Regard \mathbf{x} , \mathbf{a} , and \mathbf{h} as column vectors and $\nabla f(\mathbf{a})$ as a row vector. Then the dot product $\nabla f(\mathbf{a}) \cdot \mathbf{h}$ can be regarded as the result of matrix multiplication $\nabla f(\mathbf{a})\mathbf{h}$. (A $1 \times n$ matrix times an $n \times 1$ matrix is a 1×1 matrix, i.e. just a number.) Now let $H(\mathbf{a})$ be the $n \times n$ matrix whose entry in the i^{th} row and j^{th} column is $\frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a})$. This is called the *Hessian matrix* of f at \mathbf{a} . Let \mathbf{h}^T be the transpose of \mathbf{h} , so it is just a row vector, or $1 \times n$ matrix. Then $\mathbf{h}^T H(\mathbf{a})\mathbf{h}$ is, as you can check, a 1×1 matrix, so is just a number, and its value is $\sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a})h_i h_j$. So we can write

$$P_2(\mathbf{x}, \mathbf{a}) = f(\mathbf{a}) + \nabla f(\mathbf{a})\mathbf{h} + \frac{1}{2}\mathbf{h}^T H(\mathbf{a})\mathbf{h}.$$

The last term in this expression (including the factor of $\frac{1}{2}$) is sometimes called the *Hessian quadratic form of f at \mathbf{a}* .

Consider the case $n = 2$. Then, using the same notation as before, this becomes

$$P_2(\mathbf{x}, \mathbf{a}) = f(a, b) + \frac{\partial f}{\partial x}(a, b)(x - a) + \frac{\partial f}{\partial y}(a, b)(y - b) +$$

$$\frac{1}{2} \left[\frac{\partial^2 f}{\partial x^2}(a, b)(x - a)^2 + \frac{\partial^2 f}{\partial x \partial y}(a, b)(x - a)(y - b) + \frac{\partial^2 f}{\partial y \partial x}(a, b)(y - b)(x - a) + \frac{\partial^2 f}{\partial y^2}(a, b)(y - b)^2 \right] =$$

$$f(a, b) + \frac{\partial f}{\partial x}(a, b)h + \frac{\partial f}{\partial y}(a, b)k + \frac{1}{2} \left[\frac{\partial^2 f}{\partial x^2}(a, b)h^2 + \frac{\partial^2 f}{\partial x \partial y}(a, b)hk + \frac{\partial^2 f}{\partial y \partial x}(a, b)kh + \frac{\partial^2 f}{\partial y^2}(a, b)k^2 \right] =$$

$$f(a, b) + \frac{\partial f}{\partial x}(a, b)h + \frac{\partial f}{\partial y}(a, b)k + \frac{1}{2} \left[\frac{\partial^2 f}{\partial x^2}(a, b)h^2 + 2 \frac{\partial^2 f}{\partial x \partial y}(a, b)hk + \frac{\partial^2 f}{\partial y^2}(a, b)k^2 \right],$$

where in the last step we used the equality of mixed partials.

Warning: The most common mistakes on this material are (1) leaving out the factor of $\frac{1}{2}$ in front of the left bracket and (2) leaving out the factor of 2 inside the brackets.

EXAMPLE 4: Find the order two Taylor polynomial for $f(x, y) = \sqrt{x + 2y + 1}$ at $(0, 0)$.

We first collect the computations we will need.

$$f(x, y) = (x + 2y + 1)^{1/2}, f(0, 0) = 1$$

$$f_x(x, y) = \frac{1}{2}(x + 2y + 1)^{-1/2}, f_x(0, 0) = \frac{1}{2}$$

$$f_y(x, y) = (x + 2y + 1)^{-1/2}, f_y(0, 0) = 1$$

$$f_{xx}(x, y) = -\frac{1}{4}(x + 2y + 1)^{-3/2}, f_{xx}(0, 0) = -\frac{1}{4}$$

$$f_{xy}(x, y) = -\frac{1}{2}(x + 2y + 1)^{-3/2}, f_{xy}(0, 0) = -\frac{1}{2}$$

$$f_{yy}(x, y) = -(x + 2y + 1)^{-3/2}, f_{yy}(0, 0) = -1$$

Plugging these values into our formula we get

$$P_2(\mathbf{x}, \mathbf{0}) = 1 + \frac{1}{2}h + k + \frac{1}{2} \left[-\frac{1}{4}h^2 + 2\left(-\frac{1}{2}\right)hk - k^2 \right] = 1 + \frac{1}{2}h + k - \frac{1}{8}h^2 - \frac{1}{2}hk - \frac{1}{2}k^2$$

The book uses h_1 and h_2 instead of h and k and tacks on R_2 to get $f(\mathbf{x})$ rather than $P_2(\mathbf{x}, \mathbf{0})$, so in this style the answer would be written

$$1 + \frac{1}{2}h_1 + h_2 - \frac{1}{8}h_1^2 - \frac{1}{2}h_1h_2 - \frac{1}{2}h_2^2 + R_2$$

EXAMPLE 5: Find the order two Taylor polynomial for $f(x, y) = \frac{1}{xy}$ at $(1, 2)$.

$$f(x, y) = \frac{1}{xy}, f(1, 2) = \frac{1}{2}$$

$$f_x(x, y) = -\frac{1}{x^2y}, f_x(1, 2) = -\frac{1}{2}$$

$$f_y(x, y) = -\frac{1}{xy^2}, f_y(1, 2) = -\frac{1}{4}$$

$$f_{xx}(x, y) = \frac{2}{x^3y}, f_{xx}(1, 2) = 1$$

$$f_{xy}(x, y) = \frac{1}{x^2y^2}, f_{xy}(1, 2) = \frac{1}{4}$$

$$f_{yy}(x, y) = \frac{2}{xy^3}, f_{yy}(1, 2) = \frac{1}{4}$$

Plugging these values into our formula we get

$$P_2(\mathbf{x}, \mathbf{a}) = \frac{1}{2} - \frac{1}{2}h_1 - \frac{1}{4}h_2 + \frac{1}{2} \left[h_1^2 + 2\left(\frac{1}{4}h_1h_2\right) + \frac{1}{4}h_2^2 \right] = \frac{1}{2} - \frac{1}{2}h_1 - \frac{1}{4}h_2 + \frac{1}{2}h_1^2 + \frac{1}{4}h_1h_2 + \frac{1}{8}h_2^2$$

There are similar (but of course longer) formulas for higher values of k . They all have in common the general version of Taylor's Theorem.

Theorem 2 (Taylor's Theorem) Let $P_k(\mathbf{x}, \mathbf{a})$ be the Taylor polynomial of order k .

Then $f(\mathbf{x}) = P_k(\mathbf{x}, \mathbf{a}) + R_k(\mathbf{x}, \mathbf{a})$, where $\frac{R_k(\mathbf{x}, \mathbf{a})}{\|\mathbf{x} - \mathbf{a}\|^k} \rightarrow 0$ as $\mathbf{x} \rightarrow \mathbf{a}$.