

Second Order Derivatives for a Function of Two Variables

Let $f(x, y)$ be a function from a subset of \mathbb{R}^2 into the set of real numbers. We defined its derivative at a point (x, y) to be the linear map $Df(x, y)$ that approximates $f(x + h, y + k) - f(x, y)$ in such a way that the error term $f(x + h, y + k) - f(x, y) - Df(x, y)(h, k)$ shrinks to 0 much faster than $\|(h, k)\|$. If f has continuous partial derivatives then a formula for $Df(x, y)(h, k)$ is known; it is

$$Df(x, y)(h, k) = f_x(x, y)h + f_y(x, y)k$$

Example 1 Let $f(x, y) = x^3 + 4xy - y^2$. The domain of f is the set of all the points in \mathbb{R}^2 . The derivative of f at an arbitrary point (x, y) is the linear approximator with formula

$$D(x, y)(h, k) = (3x^2 + 4y)h + (4x - 2y)k \quad (1)$$

For instance, if we choose $(x, y) = (2, -1)$ then (1) states that the derivative of f at $(2, -1)$ has formula

$$D(2, -1)(h, k) = 8h + 10k$$

If you choose a different point, you may get a different derivative. For example, the derivative of f at $(1, 3)$ has formula

$$D(1, 3)(h, k) = 15h - 2k$$

This is expected because the shape of the graph of f changes from point to point. To define the second derivative of f at a point (x, y) we ask the question: What linear expression, (linear in (r, s)), approximates

$$D(x + r, y + s)(h, k) - D(x, y)(h, k)$$

in such a way that the error in the approximation shrinks to 0 faster than $\|(r, s)\|$? To answer the question, we simply expand $D(x + r, y + s)(h, k) - D(x, y)(h, k)$ and simplify. The result is

$$D(x + r, y + s)(h, k) - D(x, y)(h, k) = (6xr + 4s + 3r^2)h + (4r - 2s)k \quad (2)$$

It we write (2) as

$$D(x + r, y + s)(h, k) - D(x, y)(h, k) = (6xr + 4s)h + (4r - 2s)k + 3r^2h$$

then the linear term in (r, s) is clear; it is $(6xr + 4s)h + (4r - 2s)k$. The error term is also clear; it is $3r^2h$ and it shrinks to 0 faster than $\|(r, s)\|$. Note that the expression $(6xr + 4s)h + (4r - 2s)k$ is a function of (r, s) and (h, k) . The function which maps $((r, s), (h, k))$ into $(6xr + 4s)h + (4r - 2s)k$ is called the second derivative of f at (x, y) and it is denoted by $D^2f(x, y)$. More precisely,

$$D^2f(x, y)((r, s), (h, k)) = (6xr + 4s)h + (4r - 2s)k$$

Exercise 2 For each given function f , do the following:

(a) Evaluate the expression $Df(x + r, y + s)(h, k) - Df(x, y)(h, k)$.

(b) Determine the linear function that may be used to approximate $Df(x + r, y + s)(h, k) - Df(x, y)(h, k)$.

(c) Verify that the error in the approximation shrinks to 0 faster than $\|(r, s)\|$ and write down a formula for $D^2f(x, y)((r, s), (h, k))$.

$$1. \ f(x, y) = x^2y^2$$

$$2. \ f(x, y) = 2x^3 + 4xy^3 - y$$

$$3. \ f(x, y) = 4xy - 7x^3 + 3y^2$$

Definition 3 Let $f(x, y)$ be a function of two variables with a derivatives $Df(x, y)$ at every point (x, y) in its domain. It has a second derivative at a point (x, y) in its domain if there is a function that is linear in (r, s) and approximates $Df(x + r, y + s)(h, k) - Df(x, y)(h, k)$ in such a way that the error in the approximation shrinks to 0 faster than $\|(r, s)\|$. The linear approximator is then denoted by $D^2(x, y)$ and is called the second derivative of f at (x, y) .

It can be shown that if f has continuous first and second order partial derivatives then a formula for its second derivative is

$$D^2(x, y)((r, s), (h, k)) = \begin{pmatrix} r & s \end{pmatrix} \begin{pmatrix} f_{xx}(x, y) & f_{xy}(x, y) \\ f_{xy}(x, y) & f_{yy}(x, y) \end{pmatrix} \begin{pmatrix} h \\ k \end{pmatrix}$$

In expanded form, $D^2(x, y)((r, s), (h, k)) = f_{xx}(x, y)rh + (rk + sh)f_{xy}(x, y) + skf_{yy}(x, y)$. To verify it, we apply it to the function $f(x, y)$ in Example 1. Since $f_{xx} = 6x$, $f_{xy} = 4$ and $f_{yy} = -2$, the result, (as expected) is

$$\begin{aligned} D^2(x, y)((r, s), (h, k)) &= \begin{pmatrix} r & s \end{pmatrix} \begin{pmatrix} 6x & 4 \\ 4 & -2 \end{pmatrix} \begin{pmatrix} h \\ k \end{pmatrix} \\ &= \begin{pmatrix} 6xr + 4s & 4r - 2s \end{pmatrix} \begin{pmatrix} h \\ k \end{pmatrix} = (6xr + 4s)h + (4r - 2s)k \end{aligned}$$

We can now state, (without proof), a version of Taylor's theorem for a function of two variables and derive the test for the nature of a critical point which we promised earlier.

Theorem 4 (Taylor's Theorem). Let f have continuous first and second order partial derivatives at all points in its domain. Let (x, y) and $(x + h, y + k)$ be in the domain of f . Also assume that the line segment joining (x, y) to $(x + h, y + k)$ is in the domain of f . Then there is a point (α, β) on the line segment such that

$$f(x + h, y + k) = f(x, y) + Df(x, y)(h, k) + D^2f(\alpha, \beta)((h, k), (h, k))$$

We now justify the test for the nature of a critical point which we introduced on page ???. Thus assume that (c, d) is a critical point of f . Then $f_x(c, d)$ is zero and so is $f_y(c, d)$. Take any point $(c + h, d + k)$ near (c, d) . By Taylor's Theorem, there is point (α, β) on the line segment joining (c, d) and $(c + h, d + k)$ such that

$$f(c + h, d + k) = f(c, d) + Df(c, d)(h, k) + D^2f(\alpha, \beta)((h, k), (h, k)) \quad (3)$$

In matrix form,

$$\begin{aligned} f(c + h, d + k) &= f(c, d) + (0)h + (0)k + \begin{pmatrix} h & k \end{pmatrix} \begin{pmatrix} f_{xx}(\alpha, \beta) & f_{xy}(\alpha, \beta) \\ f_{xy}(\alpha, \beta) & f_{yy}(\alpha, \beta) \end{pmatrix} \begin{pmatrix} h \\ k \end{pmatrix} \\ &= f(c, d) + h^2 f_{xx}(\alpha, \beta) + 2hk f_{xy}(\alpha, \beta) + k^2 f_{yy}(\alpha, \beta) \end{aligned}$$

Since the partial derivatives of f are continuous, $f_{xx}(\alpha, \beta) \simeq f_{xx}(c, d)$, $f_{xy}(\alpha, \beta) \simeq f_{xy}(c, d)$ and $f_{yy}(\alpha, \beta) \simeq f_{yy}(c, d)$. Denote $f_{xx}(c, d)$ by A , $f_{xy}(c, d)$ by B , $f_{yy}(c, d)$ by C , $AC - B^2$ by H and move $f(c, d)$ to the left hand side. The result is

$$f(c + h, d + k) - f(c, d) \simeq Ah^2 + 2Bhk + Ck^2 \quad (4)$$

The right hand side of (4) is an example of what is called a quadratic form in two variables, (in this case the variables are h and k).

Suppose $H > 0$, (i.e. suppose $AC - B^2 > 0$).

Then A and C are non-zero and they have the same sign. The main idea is to write $Ah^2 + 2Bhk + Ck^2$ as a linear combination of two squared terms. To this end, add and subtract $\frac{B^2k^2}{A}$ to the right hand side of (4). The result is

$$\begin{aligned} f(c+h, d+k) - f(c, d) &\simeq Ah^2 + 2Bhk + \frac{B^2k^2}{A} + Ck^2 - \frac{B^2k^2}{A} \\ &= A\left(h^2 + \frac{2Bhk}{A} + \frac{B^2k^2}{A^2}\right) + (AC - B^2)\frac{k^2}{A} \\ &= A\left(h + \frac{Bk}{A}\right)^2 + H\frac{k^2}{A} \end{aligned}$$

If A is positive then as long as h and k are close to 0 (so that the approximation (4) is valid),

$$f(c+h, d+k) - f(c, d) \simeq A\left(h + \frac{Bk}{A}\right)^2 + H\frac{k^2}{A} \geq 0$$

therefore $f(x, y) \geq f(c, d)$ for all (x, y) in some neighborhood of (c, d) . This implies that (c, d) is a point of relative minimum. However, if H is positive and A is negative then

$$f(c+h, d+k) - f(c, d) \simeq A\left(h + \frac{Bk}{A}\right)^2 + H\frac{k^2}{A} \leq 0$$

for all sufficiently small values of h and k . This implies that $f(c, d) \geq f(x, y)$ for all (x, y) in some neighborhood of (c, d) , hence (c, d) is a point of relative maximum.

Suppose H is negative.

If A is also negative then $\frac{H}{A}$ is positive. If we choose $h = -\frac{Bk}{A}$ then

$$f(c+h, d+k) - f(c, d) \simeq A\left(h + \frac{Bk}{A}\right)^2 + H\frac{k^2}{A} = H\frac{k^2}{A} \geq 0$$

On the other hand, the choice $k = 0$ and $h \neq 0$ gives

$$f(c+h, d+k) - f(c, d) \simeq A\left(h + \frac{Bk}{A}\right)^2 + H\frac{k^2}{A} = Ah^2 \leq 0$$

This shows that every neighborhood of (c, d) contains points (x, y) such that $f(x, y) > f(c, d)$ and others (u, v) such that $f(u, v) < f(c, d)$. Therefore (c, d) is neither a point of relative maximum nor a point of relative minimum. A similar conclusion is arrived at if A is positive. If $A = 0$ we arrive at the same, but conclusion along a different route. More precisely, when $A = 0$ then (4) implies that

$$f(c+h, d+k) - f(c, d) \simeq 2Bhk + Ck^2 \tag{5}$$

with $B \neq 0$, (else H would not be negative). By fixing k , we can find values of h that make $2Bhk + Ck^2$ positive and others that make it negative. Therefore every neighborhood of (c, d) contains points (x, y) such that $f(x, y) > f(c, d)$ and others (u, v) such that $f(u, v) < f(c, d)$, which implies that (c, d) is neither a point of relative minimum nor a point of relative maximum.

To prove the last part it suffices to give examples. Take

$$f(x, y) = x^2 + y^3, \quad g(x, y) = x^4 + y^4, \quad h(x, y) = -x^4 - y^4 \text{ and the point } (0, 0).$$

It turns out that $(0, 0)$ is a critical point for all the three functions, (verify). Furthermore, $H = 0$ for all three. Note that $g(x, y) = x^4 + y^4 \geq 0$ for all $(x, y) \in \mathbb{R}^2$ and $g(0, 0) = 0$, therefore $(0, 0)$ is a point of relative minimum for g . A similar argument shows that $(0, 0)$ is a point of relative maximum for h . In the case of f , we observe that if $y > 0$ then $f(0, y) > 0 = f(0, 0)$ and if $y < 0$ then $f(0, y) < f(0, 0)$. This shows that every neighborhood of $(0, 0)$ contains points (x, y) such that $f(x, y) > f(0, 0)$ and others (u, v) such that $f(u, v) < f(0, 0)$, which implies that $(0, 0)$ is neither a point of relative minimum nor a point of relative maximum.