14 Partial Differentiation

14.1 Functions of Several Variables

In single-variable calculus we were concerned with functions that map the real numbers \( R \) to \( R \), sometimes called “real functions of one variable”, meaning the “input” is a single real number and the “output” is likewise a single real number. In the last chapter we considered functions taking a real number to a vector, which may also be viewed as functions \( f: R \to R^3 \), that is, for each input value we get a position in space. Now we turn to functions of several variables, meaning several input variables, functions \( f: R^n \to R \). We will deal primarily with \( n = 2 \) and to a lesser extent \( n = 3 \), in fact many of the techniques we discuss can be applied to larger values of \( n \) as well.

A function \( f: R^2 \to R \) maps a pair of values \((x, y)\) to a single real number. The three-dimensional coordinate system we have already used is a convenient way to visualize such functions: above each point \((x, y)\) in the \( x-y \) plane we graph the point \((x, y, z)\), where of course \( z = f(x, y) \).

**EXAMPLE 14.1.1** Consider \( f(x, y) = 3x + 4y - 5 \). Writing this as \( z = 3x + 4y - 5 \) and then \( 3x + 4y - z = 5 \) we recognize the equation of a plane. In the form \( f(x, y) = 3x + 4y - 5 \) the emphasis has shifted: we now think of \( x \) and \( y \) as independent variables and \( z \) as a variable dependent on them, but the geometry is unchanged.

**EXAMPLE 14.1.2** We have seen that \( x^2 + y^2 + z^2 = 4 \) represents a sphere of radius 2. We cannot write this in the form \( f(x, y) \), since for each \( x \) and \( y \) in the disk \( x^2 + y^2 < 4 \) there are two corresponding points on the sphere. As with the equation of a circle, we can resolve this equation into two functions, \( f(x, y) = \sqrt{4-x^2-y^2} \) and \( f(x, y) = -\sqrt{4-x^2-y^2} \), representing the upper and lower hemispheres. Each of these is an example of a function with a restricted domain: only certain values of \( x \) and \( y \) make sense (namely, those for which \( x^2 + y^2 \leq 4 \)) and the graphs of these functions are limited to a small region of the plane.

**EXAMPLE 14.1.3** Consider \( f = x + y \). This function is defined only when both \( x \) and \( y \) are non-negative. When \( y = 0 \) we get \( f(x, y) = x \), the familiar square root function in the \( x-y \) plane, and when \( x = 0 \) we get the same curve in the \( y-z \) plane. Generally speaking, we see that starting from \( f(0, 0) = 0 \) this function gets larger in every direction in roughly the same way that the square root function gets larger. For example, if we restrict attention to the line \( x = y \), we get \( f(x, y) = 2x \) and along the line \( y = 2x \) we have \( f(x, y) = \sqrt{5x^2} \) along the line \( y = 2x \) where the horizontal axis should be. In order to pretend that this line is the horizontal axis, we need to write the function in terms of the distance from the origin, which is \( \sqrt{x^2 + y^2} \). Now \( f(x, y) = x^2 + 2xy = (x + y)^2 \). So the cross-section is the “same” parabola as in the \( x-y \) plane, but the function behaves much like functions of two variables; we will on occasion discuss functions of three variables. The principal difficulty with such functions is visualizing them, as they do not “fit” in the three dimensions we are familiar with. For three variables there are various ways to interpret functions that make them easier to understand. For example, \( f(x, y, z) \) could represent the temperature at the point \((x, y, z)\), the pressure, or the strength of a magnetic field. It remains useful to consider those points at which \( f(x, y, z) = k \), where \( k \) is some constant value. If \( f(x, y, z) \) is temperature, the set of points \((x, y, z)\) such that \( f(x, y, z) = k \) is the collection of points in space with temperature equal to \( k \); in general this is called a level set; for three variables, a level set is typically a surface, called a level surface.

**EXAMPLE 14.1.4** Suppose the temperature at \((x, y, z)\) is \( T(x, y, z) = e^{-(x^2+y^2+z^2)} \). This function has a maximum value of 1 at the origin, and tends to 0 in all directions. If \( k \) is positive and at most 1, the set of points for which \( T(x, y, z) = k \) is those points satisfying \( x^2 + y^2 + z^2 = \ln k \), a sphere centered at the origin. The level surfaces are the concentric spheres centered at the origin.

### Exercises 14.1.

1. Let \( f(x, y) = (x-y)^2 \). Determine the equations and shapes of the cross-sections when \( z = 0 \), \( y = 0 \), \( x = y \), and describe the level curves. Use a three-dimensional graphing tool to graph the surface. ⇧
2. Let \( f(x, y) = |x| + |y| \). Determine the equations and shapes of the cross-sections when \( x = 0 \), \( y = 0 \), \( x = y \), and describe the level curves. Use a three-dimensional graphing tool to graph the surface. ⇧
3. Let \( f(x, y) = e^{-(x^2+y^2)} \). Determine the equations and shapes of the cross-sections when \( x = 0 \), \( y = 0 \), \( x = y \), and describe the level curves. Use a three-dimensional graphing tool to graph the surface. ⇧
4. Let \( f(x, y) = \sin(x - y) \). Determine the equations and shapes of the cross-sections when \( x = 0 \), \( y = 0 \), \( x = y \), and describe the level curves. Use a three-dimensional graphing tool to graph the surface. ⇧
5. Let \( f(x, y) = (x^2 - y^2)^2 \). Determine the equations and shapes of the cross-sections when \( x = 0 \), \( y = 0 \), \( x = y \), and describe the level curves. Use a three-dimensional graphing tool to graph the surface. ⇧
6. Find the domain of each of the following functions of two variables:
   a. \( \sqrt{3-x^2} + \sqrt{3-y^2} \)
   b. \( \arcsin(x^2 + y^2 - 2) \)
   c. \( \sqrt{1 - x^2 - 4y^2} \)

7. Below are two sets of level curves. One is for a cone, one is for a paraboloid. Which is which? Explain.
14.2 LIMITS AND CONTINUITY

To develop calculus for functions of one variable, we needed to make sense of the concept of a limit, which we needed to understand continuous functions and to define the derivative. Limits involving functions of two variables can be considerably more difficult to deal with; fortunately, most of the functions we encounter are fairly easy to understand.

The potential difficulty is largely due to the fact that there are many ways to "approach" a point in the $x$-$y$ plane. If we want to say that \( \lim_{(x,y) \to (a,b)} f(x,y) = L \), we need to capture the idea that as \((x,y)\) gets close to \((a,b)\) then \(f(x,y)\) gets close to \(L\). For functions of one variable, \(f(x)\), there are only two ways that \(x\) can approach \(a\): from the left or right. But there are an infinite number of ways to approach \((a,b)\): along any one of an infinite number of lines, or an infinite number of parabolas, or an infinite number of sine curves, and so on. We might hope that it’s really not so bad—suppose, for example, that along every possible line through \((a,b)\) the value of \(f(x,y)\) gets close to \(L\); surely this means that \(f(x,y)\) approaches \(L\) as \((x,y)\) approaches \((a,b)\). Sadly, no.

**Example 14.2.1** Consider \(f(x,y) = x^2/y^2\) (for \(y \neq 0\)). When \(x = 0\) or \(y = 0\), \(f(x,y)\) is 0, so the limit of \(f(x,y)\) approaching the origin along either the \(x\) or \(y\) axis is 0. Moreover, along the line \(y = mx\), \(f(x,y) = mx^2/(x^2 + m^2x^2)\). As \(x\) approaches 0 this expression approaches 0 as well. So along every line through the origin \(f(x,y)\) approaches 0. Now suppose we approach the origin along \(x = y^2\). Then

\[
\lim_{x \to 0} \frac{x^2}{y^2} = \begin{cases} 
1 & \text{if } y > 0 \\
0 & \text{if } y < 0
\end{cases}
\]

so the limit is 1/2. Looking at figure 14.2.1, it is apparent that there is a ridge above \(x = y^2\). Approaching the origin along a straight line, we go over the ridge and then drop down toward 0, but approaching along the ridge the height is a constant 1/2. Thus, there is no limit at \((0,0)\).

Fortunately, we can define the concept of limit without needing to specify how a particular point is approached—indeed, in definition 2.3.2, we didn’t need the concept of "approach." Roughly, that definition says that when \(x\) is close to \(a\) then \(f(x)\) is close to \(L\); there is no mention of "how" we get close to \(x\) or \(L\). We can adapt that definition to two variables quite easily:

**Definition 14.2.2 Limit** Suppose \(f(x,y)\) is a function. We say that

\[
\lim_{(x,y) \to (a,b)} f(x,y) = L
\]

if for every \(\epsilon > 0\) there is a \(\delta > 0\) so that whenever \(0 < \sqrt{(x-a)^2 + (y-b)^2} < \delta\), \(|f(x,y) - L| < \epsilon\).

We want to force this to be less than \(\epsilon\) by picking \(\delta\) "small enough." If we choose \(\delta = \epsilon/3\) then

\[
\frac{3\sqrt{x^2 + y^2}}{\sqrt{x^2 + y^2}} = \frac{3\epsilon}{3} = \epsilon.
\]

Recall that a function \(f(x)\) is continuous at \(x = a\) if \(\lim_{x \to a} f(x) = f(a)\); roughly this says that there is no "hole" or "jump" at \(x = a\). We can say exactly the same thing about a function of two variables.

**Definition 14.2.4** \(f(x,y)\) is continuous at \((a,b)\) if \(\lim_{(x,y) \to (a,b)} f(x,y) = f(a,b)\).

**Example 14.2.5** The function \(f(x,y) = 3x^2y/(x^2 + y^2)\) is not continuous at \((0,0)\), because \(f(0,0)\) is not defined. However, we know that \(\lim_{(x,y) \to (0,0)} f(x,y) = 0\), so we can easily "fix" the problem, by extending the definition of \(f\) so that \(f(0,0) = 0\). This surface is shown in figure 14.2.2.

![Figure 14.2.2](image)

**Figure 14.2.2** \(f(x,y) = \frac{3x^2y}{x^2 + y^2}\) (AP)

**Exercise 14.2**

Determine whether each limit exists. If it does, find the limit and prove that it is the limit; if it does not, explain how you know.

1. \(\lim_{(x,y) \to (0,0)} \frac{x^2}{y^2} \)
2. \(\lim_{(x,y) \to (0,0)} \frac{x}{x^2 + y^2} \)
3. \(\lim_{(x,y) \to (0,0)} \frac{x}{x^2 + y^2} \)
4. \(\lim_{(x,y) \to (0,0)} \frac{x^2}{x^2 + y^2} \)
5. \(\lim_{(x,y) \to (0,0)} \frac{\sin(x^2 + y^2)}{x^2 + y^2} \)
6. \(\lim_{(x,y) \to (0,0)} \frac{\sqrt{x^2 + y^2}}{x^2 + y^2} \)
7. \(\lim_{(x,y) \to (0,0)} \frac{x^2 + y^2}{x^2 + y^2} \)
8. \(\lim_{(x,y) \to (0,0)} \frac{x^2 + y^2}{x^2 + y^2} \)
9. \(\lim_{(x,y) \to (0,0)} \frac{x^2 + y^2}{x^2 + y^2} \)
10. \(\lim_{(x,y) \to (0,0)} (x^2 - y^2) \ln x \)
11. \(\lim_{(x,y) \to (0,0)} (x + y) \)
12. \(\lim_{(x,y) \to (0,0)} \frac{x^2 + y^2}{x^2 + y^2} \)
13. Does the function \( f(x, y) = \frac{x - y}{1 + x^2 + y^2} \) have any discontinuities? What about \( f(x, y) = \frac{x - y}{1 + x^2 + y^2} \)?

### 14.3 Partial Differentiation

When we first considered what the derivative of a vector function might mean, there was really not much difficulty in understanding either how such a thing might be computed or what it might measure. In the case of functions of two variables, things are a bit harder to understand. If we think of a function of two variables in terms of its graph, a surface, there is a more-or-less obvious derivative-like question we might ask: namely, how “steep” is the surface. But it’s not clear that this has a simple answer, nor how we might proceed.

We will start with what seem to be very small steps toward the goal; surprisingly, it turns out that these simple ideas hold the keys to a more general understanding.

Imagine a particular point on a surface; what might we be able to say about how steep it is? We can limit the question to make it more familiar: how steep is the surface in a particular direction? What does this even mean? Here’s one way to think of it: Suppose we’re interested in the point \((a, b, c)\). Pick a straight line in the \(x\)-\(y\) plane through the point \((a, b, 0)\), then extend the line vertically into a plane. Look at the intersection of the plane with the surface. If we pay attention to just the plane, we see the chosen straight line where the \(x\)-axis would normally be, and the intersection with the surface shows up as a curve in the plane. Figure 14.3.1 shows the parabolic surface from figure 14.1.2, exposing its cross-section above the line \(x + y = 1\).

In principle, this is a problem we know how to solve: find the slope of a curve in a plane. Let’s start by looking at some particularly easy lines: those parallel to the \(x\) or \(y\) axis. Suppose we are interested in the cross-section of \(f(x, y)\) above the line \(y = b\). If we substitute \(b\) for \(y\) in \(f(x, y)\), we get a function in one variable, describing the height of the cross-section as a function of \(x\). Because \(y = b\) is parallel to the \(z\)-axis, if we view it from a vantage point on the negative \(y\)-axis, we will see what appears to be simply an ordinary curve in the \(x\)-\(z\) plane.

Consider again the parabolic surface \(f(x, y) = x^2 + y^2\). The cross-section above the line \(y = 1\) consists of all points \((x, 2, x^2 + 4)\). Looking at this cross-section from somewhere on the negative \(y\) axis, we see what appears to be just the curve \(f(x) = x^2 + 4\). At any point on the cross-section, \((x, 2, x^2 + 4)\), the steepness of the surface in the direction of the line \(y = 2\) is simply the slope of the curve \(f(x) = x^2 + 4\), namely \(2x\). Figure 14.3.2 shows the same paraboloid as before, but now cut by the plane \(y = 2\). The left graph shows the cut-off surface, the right shows just the cross-section, looking up from the negative \(y\)-axis toward the origin.

The partial derivative with respect to \(x\) of \(x^2 + y^2\) is \(2x\). To emphasize that we are only temporarily assuming \(y\) is constant, we use a slightly different notation: \(\frac{df}{dx}(x^2 + y^2) = 2x\); the ‘\(\partial\)’ reminds us that there are more variables than \(x\), but that only \(x\) is being treated as a variable. We read the equation as “the partial derivative of \((x^2 + y^2)\) with respect to \(x\) is \(2x\)”. A convenient alternate notation for the partial derivative of \(f(x, y)\) with respect to \(x\) is \(f_x(x, y)\).

#### EXAMPLE 14.3.1

The partial derivative with respect to \(x\) of \(x^2 + 3xy\) is \(2x + 3y\).

Note that the partial derivative includes the variable \(y\), unlike the example \(x^2 + y^2\). It is somewhat unusual for the partial derivative to depend on a single variable; this example is more typical.

Of course, we can do the same sort of calculation for lines parallel to the \(y\)-axis. We temporarily hold \(x\) constant, which gives us the equation of the cross-section above a line \(x = k\). We can then compute the derivative with respect to \(y\); this will measure the steepness of the curve in the \(y\) direction.

#### EXAMPLE 14.3.2

The partial derivative with respect to \(y\) of \(f(x, y) = \sin(xy) + 3xy\) is \(f_y(x, y) = \cos(xy) + 3x\).

So far, using no new techniques, we have succeeded in measuring the slope of a surface in two quite special directions. For functions of one variable, the derivative is closely linked to the notion of tangent line. For surfaces, the analogous idea is the tangent plane—a plane that just touches a surface at a point, and has the same “steepness” as the surface in all directions. Even though we haven’t yet figured out how to compute the slope in all directions, we have enough information to find tangent planes. Suppose we want the plane tangent to a surface at a particular point \((a, b, c)\). If we compute the two partial derivatives of the function for that point, we get enough information to determine two lines tangent to the surface, both through \((a, b, c)\) and both tangent to the surface in their respective directions. These two lines determine a plane, that is, there is exactly one plane containing the two lines: the tangent plane. Figure 14.3.3 shows (part of) two tangent lines at a point, and the tangent plane containing them.

How can we discover an equation for this tangent plane? We know a point on the plane, \((a, b, c)\); we need a vector normal to the plane. If we can find two vectors, one parallel to each of the tangent lines we know how to find, then the cross product of these vectors will give the desired normal vector.

How can we find vectors parallel to the tangent lines? Consider first the line tangent to the surface above the line \(y = b\). A vector \((u, v, w)\) parallel to this tangent line must have \(y\) component \(v = 0\), and we may as well take the \(x\) component to be \(u = 1\). The ratio
right side does the same, because \( z = x \) and \( z = y \) approach \((x_0, y_0)\), then \( f \) is differentiable there.

\[ \frac{df}{dx} = \frac{dy}{dx} = \frac{dz}{dx} \]

\[ \frac{df}{dy} = \frac{dx}{dy} = \frac{dz}{dy} \]

Consider the surface \( z = x^2 + y^2 \), and suppose that \( x = 2 + t \) and \( y = 1 - t^3 \). We can think of the latter two equations as describing how \( x \) and \( y \) change relative to, say, time. Then

\[ x = x' + 2t \quad y = 2y' + t^3 \quad z = z' + t^6 \]

This tells us explicitly how the \( z \)-coordinate of the point on the surface depends on \( t \). If we want to know \( dx \) and \( dy \), we can compute it more or less directly—it's actually a bit simpler to use the chain rule:

\[ \frac{dz}{dt} = 2x'y' + 2xy'' + zy'' \]

\[ d z = (2x'y')' + (2xy'')' + (zy'')' \]

If we evaluate carefully, at the middle step, we get \( dx/dt = (2x'y')' \). This shows that \( dx + dy + dz \) is differentiable. This turns out to be true in general, and gives us a new chain rule:

**Theorem 14.4.1** Suppose that \( z = f(x, y) \) is differentiable, \( x = g(t) \), and \( y = h(t) \).

Assuming that the relevant derivatives exist.

\[ \frac{dz}{dt} = \frac{dx}{dt} \frac{df}{dx} + \frac{dy}{dt} \frac{df}{dy} \]

**Proof.** If \( f \) is differentiable, then

\[ \Delta z = f(x_0 + \Delta x, y_0 + \Delta y) - f(x_0, y_0) \]

where \( \Delta x \) and \( \Delta y \) approach 0 as \((x, y)\) approaches \((x_0, y_0)\). Then

\[ \frac{\Delta z}{\Delta t} = \frac{f(x_0 + \Delta x, y_0 + \Delta y) - f(x_0, y_0)}{\Delta t} \]

As \( \Delta t \) approaches 0, \((x, y)\) approaches \((x_0, y_0)\) and so

\[ \lim_{\Delta t \to 0} \frac{\Delta z}{\Delta t} = \frac{dx}{dt} \frac{df}{dx} + \frac{dy}{dt} \frac{df}{dy} \]

and we get the desired formula.
We can write the chain rule in a way that is somewhat closer to the single variable chain rule:
\[ \frac{df}{dx} = \left( f_x, f_y \right) \cdot (x', y'), \]
or (roughly) the derivatives of the outside function "times" the derivatives of the inside functions. Not surprisingly, essentially the same chain rule works for functions of more than two variables, for example, given a function of three variables \( f(x,y,z) \), where each of \( x, y \) and \( z \) is a function of \( t \),
\[ \frac{df}{dt} = \left( f_x, f_y, f_z \right) \cdot (x', y', z'). \]

We can even extend the idea further. Suppose that \( f(x,y) \) is a function and \( x = g(x,t) \) and \( y = h(x,t) \) are functions of two variables \( s \) and \( t \). Then \( f \) is "really" a function of \( s \) and \( t \) as well, and
\[ \frac{\partial f}{\partial s} = f_x \frac{\partial x}{\partial s} + f_y \frac{\partial y}{\partial s}, \]
\[ \frac{\partial f}{\partial t} = f_x \frac{\partial x}{\partial t} + f_y \frac{\partial y}{\partial t}. \]

The natural extension of this to \( f(x,y,z) \) works as well.

Recall that we used the ordinary chain rule to do implicit differentiation. We can do the same with the new chain rule.

**Example 14.4.2** \( x^2 + y^2 + z^2 = 4 \) defines a sphere, which is not a function of \( x \) and \( y \), though it can be thought of as two functions, the top and bottom hemispheres. We can think of \( x \) as one of those two functions, so really \( x = x(s,t) \), and \( x \) can think of \( x \) and \( y \) as particularly simple functions of \( x \) and \( y \), and let \( f(x,y,z) = x^2 + y^2 + z^2. \) Since \( f(x,y,z) = 4, \) \( \frac{df}{dx} = 0 \), but using the new chain rule:
\[ 0 = \frac{\partial f}{\partial x} = 2x = 2x \]
\[ \frac{\partial^2 f}{\partial x^2} = -2. \]

In a similar manner we can compute \( \partial y/\partial y. \)

### 14.5 Directional Derivatives

So we need to somehow "mark off" units on the line, and we need a convenient way to refer to the line in calculations. It turns out that we can accomplish both by using the vector form of a line. Suppose that \( \mathbf{u} \) is a unit vector \( (u_1, u_2) \) in the direction of interest. A vector equation for the line through \((x_0, y_0)\) in this direction is \( \mathbf{r}(t) = (x_0, y_0, z_0) + t (u_1, u_2, u_3) \). The height of the surface above the point \((x_0 + u_1 t, y_0 + u_2 t, z_0 + u_3 t)\) is \( g(t) = f(x_0 + u_1 t, y_0 + u_2 t, z_0 + u_3 t) \). Because \( \mathbf{u} \) is a unit vector, the value of \( t \) is precisely the distance along the line from \((x_0, y_0)\) to \((x + u_1 t, y + u_2 t, z + u_3 t)\). This means that the line is effectively a \( t \) axis, with origin at the point \((x_0, y_0)\), so the slope we seek is
\[ y'(0) = \left( f_x(x_0, y_0), f_y(x_0, y_0) \right) \cdot (u_1, u_2) \]
\[ = (f_x, f_y) \cdot \mathbf{u} \]
\[ = \nabla f \cdot \mathbf{u}. \]

Here we have used the chain rule and the derivatives \( \frac{d}{dt}(x_0 + u_1 t) = u_1 \) and \( \frac{d}{dt}(y_0 + u_2 t) = u_2 \). The vector \( (f_x, f_y) \) is very useful, so it has its own symbol, \( \nabla f \); pronounced "nabla f"; it is also called the gradient of \( f \).

**Example 14.5.1** Find the slope of the line \( z = x^2 + y^2 \) at \((1, 1)\) in the direction of the vector \( (3, 4) \).

We first compute the gradient \( \nabla f = (2x, 2y) \), which is \((2, 4)\) at \((1, 1)\). A unit vector in the desired direction is \((3/5, 4/5, 0)\), and the desired slope is then \((2, 4) \cdot (3/5, 4/5, 0) = 6/5 + 16/5 = 22/5\).

**Example 14.5.2** Find a tangent vector to \( z = x^2 + y^2 \) at \((1, 1)\) in the direction of the vector \((3, 4)\), and show that it is parallel to the tangent plane at that point.

Since \((3/5, 4/5)\) is a unit vector in the desired direction, we can easily expand it to a tangent vector simply by adding the third coordinate computed in the previous example: \((3/5, 4/5, 22/5)\). To see that this vector is parallel to the tangent plane, we can compute its dot product with a normal to the plane. We know that a normal to the tangent plane is
\[ (f_x, f_y, -1) = (2, 4, -1), \]
and the dot product is \((2, 4, -1) \cdot (3/5, 4/5, 22/5) = 6/5 + 16/5 - 22/5 = 0\), so the two vectors are perpendicular. (Note that the vector normal to the surface, namely \( (f_x, f_y, -1) \), is simply the gradient with \(-1\) tacked on as the third component.)

The slope of a surface given by \( z = f(x,y) \) in the direction of a (two-dimensional) vector \( \mathbf{u} \) is called the directional derivative of \( f \), written \( D_\mathbf{u} f \). The directional derivative immediately provides us with some additional information. We know that
\[ D_\mathbf{u} f = \mathbf{u} \cdot \nabla f = \left\| \nabla f \right\| \cos \theta = \left\| \nabla f \right\| \cos \theta \]
if \( \mathbf{u} \) is a unit vector; \( \theta \) is the angle between \( \nabla f \) and \( \mathbf{u} \). This tells us immediately that the largest value of \( D_\mathbf{u} f \) occurs when \( \cos \theta = 1 \), namely, when \( \theta = 0 \), so \( \nabla f \) is parallel to \( \mathbf{u} \). In other words, the gradient \( \nabla f \) points in the direction of steepest ascent of the surface, and \(-\nabla f \) is the slope in that direction. Likewise, the smallest value of \( D_\mathbf{u} f \) occurs when \( \cos \theta = -1 \), namely, when \( \theta = \pi \), so \( \nabla f \) is anti-parallel to \( \mathbf{u} \). In other words, \(-\nabla f \) points in the direction of steepest descent of the surface, and \(-\nabla f \) is the slope in that direction.

**Example 14.5.3** Investigate the direction of steepest ascent and descent for \( z = x^2 + y^2 \).

The gradient is \((2x, 2y) = 2(z, y)\); this is a vector parallel to the vector \((x, y)\), so the direction of steepest ascent is directly away from the origin, starting at the point \((x,y)\). The direction of steepest descent is thus directly toward the origin from \((x, y)\). Note that at \((0,0)\) the gradient vector is \((0,0)\), which has no direction, and it is clear from the plot of this surface that there is a minimum point at the origin, and tangent vectors in all directions are parallel to the \(x\)-y plane.

If \( f \) is perpendicular to \( \mathbf{u} \), \( D_{\mathbf{u}} f = \left\| \nabla f \right\| \cos(\theta/2) = 0 \), since \(\cos(\theta/2) = 0\). This means that in either of the two directions perpendicular to \( \nabla f \), the slope of the surface is 0; this implies that a vector in either of these directions is tangent to the level curve at that point. Starting with \( f = f_x f_y \) or \(-f_x f_y \) will work. Thus, \( f = f_x f_y \) is easy to find a vector perpendicular to it; either \( f_x f_y \) or \(-f_x f_y \) will work.

If \( f(x, y, z) \) is a function of three variables, all the calculations proceed in essentially the same way. The rate at which \( f \) changes in a particular direction is \( \nabla f \cdot \mathbf{u} \). To see that \( \nabla f \) points in the direction of maximum rate of increase, \( \nabla f \) points in the direction of maximum rate of decrease, and any vector perpendicular to \( \nabla f \) is tangent to the level surface \( f(x, y, z) = k \) at the point in question. Of course there are no longer just two such vectors, the vectors perpendicular to \( \nabla f \) describe the tangent plane to the level surface, or in other words \( \nabla f \) is a normal to the tangent plane.

**Example 14.5.4** Suppose the temperature at a point in space is given by \( T(x, y, z) = T_0(1 - x^2 + y^2 + z^2) \), at the origin the temperature in Kelvin is \( T_0 \), and it decreases in every direction from there. It might be, for example, that there is a source of heat at the...
origin, and as we get farther from the source, the temperature decreases. The gradient is
\[\nabla T = \left(\frac{-2T_x x}{(1 + x^2 + y^2 + z^2)^2}, \frac{-2T_y y}{(1 + x^2 + y^2 + z^2)^2}, \frac{-2T_z z}{(1 + x^2 + y^2 + z^2)^2}\right)\]

so \(z = \frac{k}{\sqrt{3}}\). The desired points are \(\left(\frac{\sqrt{2}}{\sqrt{3}}, \frac{\sqrt{2}}{\sqrt{3}}, \frac{\sqrt{2}}{\sqrt{3}}\right)\) and \(\left(-\frac{\sqrt{2}}{\sqrt{3}}, -\frac{\sqrt{2}}{\sqrt{3}}, -\frac{\sqrt{2}}{\sqrt{3}}\right)\). The ellipsoid and the three planes are shown in figure 14.5.1.

### 14.6 Higher order derivatives

10. Suppose the temperature at \((x, y, z)\) is given by \(T = x^2 y^2\). In what direction can you go from the point \((1, 1, 1)\) to maintain the same temperature? 

11. Find an equation for the plane tangent to \(x^2 + y^2 + z^2 = 7\) at \((1, 1, 3)\). 

12. Find an equation for the plane tangent to \(xyz = 6\) at \((2, 1, 3)\). 

13. Find a vector function for the line normal to \(x^2 + 2y^2 + 3z^2 = 26\) at \((2, -3, 1)\). 

14. Find a vector function for the line normal to \(x^2 + y^2 + z^2 = 56\) at \((4, 2, -2)\). 

15. Find a vector function for the line normal to \(x^2 + y^2 - z^2 = 0\) at \((4, 2, 6)\). 

16. Find the directions in which the directional derivative of \(f(x, y) = x + \sin(xy)\) at the point \((0, 0)\) has the value 1. 

17. Show that the curve \(r(t) = (\ln 2t, \ln t, 0)\) is tangent to the surface \(x^2 + y^2 + z^2 = 1\) at the point \((0, 0, 1)\). 

18. A bug is crawling on a surface of a hot plate, the temperature of which at the point \(x, y\) is \(T(x, y) = 100 - x^2 - 3y^2\). 

a. If the bug is at the point \((2, 1)\), in what direction should it move to cool off the fastest? 

b. If the bug is at the point \((1, 3)\), in what direction should it move in order to maintain its current temperature? 

19. The elevation on a portion of a hill is given by \(f(x, y) = 100 - 4x^2 - 2y^2\). From the location above \((2, 1)\), in which direction will water run off? 

20. Suppose that \(g(x, y) = y - x^2\). Find the gradient at the point \((1, 3)\). Sketch the level curve of the graph of \(g(x, y) = 2\), and plot both the tangent line and the gradient vector at the point \((1, 3)\). (Make your sketch large.) What do you notice, geometrically? 

21. The gradient \(\nabla f\) is a vector valued function of two variables. Prove the following gradient rules. Assume \(f(x, y)\) and \(g(x, y)\) are differentiable functions.

a. \(\nabla (f + g) = \nabla f + \nabla g\) 

b. \(\nabla (cf) = c\nabla f\) 

c. \(\nabla (f(x, y)) = \frac{\partial f}{\partial x}(x, y) + \frac{\partial f}{\partial y}(x, y)\) 

### 14.6 Higher order derivatives

In single variable calculus we saw that the second derivative is often useful: in appropriate circumstances it measures acceleration; it can be used to identify maximum and minimum points; it tells us something about how sharply curved a graph is. Not surprisingly, second derivatives are also useful in the multi-variable case, but again not surprisingly, things are a bit more complicated.

It’s easy to see where some complication is going to come from: with two variables there are four possible second derivatives. To take a “derivative,” we must take a partial derivative with respect to \(x\) or \(y\), and there are four ways to do it: \(x\) then \(x\), then \(y\), then \(y\).

#### Example 14.6.1
Compute all four second derivatives of \(f(x, y) = x^2 y^2\).

Using an obvious notation, we get:

- \(f_{xx} = 2y^2\) 
- \(f_{xy} = 4xy\) 
- \(f_{yx} = 4xy\) 
- \(f_{yy} = 2x^2\)

You will have noticed that two of these are the same, the “mixed partials” computed by taking partial derivatives with respect to both variables in the two possible orders. This is not an accident—as long as the function is reasonably nice, this will always be true.

#### Theorem 14.6.2 Clairaut’s Theorem
If the mixed partial derivatives are continuous, they are equal.

#### Example 14.6.3
Compute the mixed partials of \(f(x, y) = \sin(x^2 + y^2)\).

\[f_x = 2x \sin(x^2 + y^2)\]
\[f_y = 2y \sin(x^2 + y^2)\]

We leave \(f_{xy}\) as an exercise. 

#### Exercises 14.6

1. Find all first and second partial derivatives of \(f(x, y) = xy(2x^2 + y^2)\). 

2. Find all first and second partial derivatives of \(x^2 y^2 + y^2\). 

3. Find all first and second partial derivatives of \(x^3 + y^2 + 10\). 

4. Find all first and second partial derivatives of \(x^3 y^4\). 

5. Find all first and second partial derivatives of \(\sin(x^2 + y^2)\). 

6. Find all first and second partial derivatives of \(\sin^2(x^2 + y^2)\). 

7. Find all first and second partial derivatives of \(\ln(x^2 + y^2)\). 

8. Find all first and second partial derivatives of \(x^3 + y^2 + 16x^2 - 64 = 0\). 

9. Find all first and second partial derivatives of \(xy + x + y + 2x = 1\). 

10. Let \(a\) and \(b\) be constants. Prove that the function \(f(x, t) = e^{-at}\sin(\alpha t)\) is a solution to the heat equation \(u_t = u_{xx}\).

11. Let \(a\) be a constant. Prove that \(a = \sin(x - at) + \ln(x + at)\) is a solution to the wave equation \(u_{tt} = u_{xx}\).

12. How many third-order derivatives does a function of 2 variables have? How many of these are distinct?
14.7 Maxima and minima

Suppose a surface given by \( f(x, y) \) has a local maximum at \((x_0, y_0, z_0)\); geometrically, this point on the surface looks like the top of a hill. If we look at the cross-section in the plane \( y = y_0 \), we will see a local maximum on the curve at \((x_0, y_0)\), and we know from single-variable calculus that \( \frac{dy}{dx} = 0 \) at this point. Likewise, in the plane \( x = x_0 \), \( \frac{dy}{dx} = 0 \) at this point too. So if there is a local maximum at \((x_0, y_0, z_0)\), both partial derivatives at the point must be zero, and likewise for a local minimum. Thus, to find local maximum and minimum points, we need only consider those points at which both partial derivatives are 0. As in the single-variable case, it is possible for the derivatives to be 0 at a point that is neither a maximum or a minimum, so we need to test these points further.

You will recall that in the single variable case, we examined three methods to identify maximum and minimum points: the most useful is the second derivative test, though it doesn’t always work.

**THEOREM 14.7.1** Suppose that the second partial derivatives of \( f(x, y) \) are continuous near \((x_0, y_0)\) and \( f_{xx}(x_0, y_0) = f_{yy}(x_0, y_0) = 0 \). We denote by \( D \) the discriminant \( D(x_0, y_0) = f_{xx}(x_0, y_0) f_{yy}(x_0, y_0) - (f_{xy}(x_0, y_0))^2 \). If \( D > 0 \) and \( f_{xx}(x_0, y_0) < 0 \) there is a local maximum at \((x_0, y_0)\); if \( D > 0 \) and \( f_{xx}(x_0, y_0) > 0 \) there is a local minimum at \((x_0, y_0)\); if \( D < 0 \) there is neither a maximum nor a minimum at \((x_0, y_0)\); if \( D = 0 \) the test fails.

**EXAMPLE 14.7.2** Verify that \( f(x, y) = x^2 + y^2 \) has a minimum at \((0, 0)\).

First, we compute all the needed derivatives: \( f_x = 2x \), \( f_y = 2y \), \( f_{xx} = 2 \), \( f_{yy} = 2 \), \( f_{xy} = 0 \).

The derivatives \( f_x \) and \( f_y \) are zero only at \((0, 0)\). Applying the second derivative test there:

\[
D(0, 0) = f_{xx}(0, 0) f_{yy}(0, 0) - (f_{xy}(0, 0))^2 = 2 - 0 = 4 > 0
\]

and

\[
f_{xx}(0, 0) = 2 > 0,
\]

so there is a local minimum at \((0, 0)\), and there are no other possibilities.

**EXAMPLE 14.7.3** Find all local maxima and minima for \( f(x, y) = x^2 - y^2 \).

**EXAMPLE 14.7.4** Find all local maximum and minimum points of \( f(x, y) = x^3 + y^3 \).

The derivatives:

\[
f_x = 3x^2, \quad f_y = 3y^2, \quad f_{xx} = 6x, \quad f_{yy} = 6y, \quad f_{xy} = 0.
\]

Again there is a single critical point, at \((0, 0)\), and

\[
D(0, 0) = f_{xx}(0, 0) f_{yy}(0, 0) - (f_{xy}(0, 0))^2 = 0 - 0 = 0,
\]

so we get no information. In this case, a little thought shows there is neither a maximum nor a minimum at \((0, 0)\) when \(x\) and \(y\) both differ from 0, \(f(x, y) > 0\), and when \(x\) and \(y\) both are negative, \(f(x, y) < 0\), and there are points of both kinds arbitrarily close to \((0, 0)\). Alternatively, if we look at the cross-section when \(y = 0\), we get \(f(x, 0) = x^2\), which does not have either a maximum or minimum at \(x = 0\).

**EXAMPLE 14.7.5** Find all local maxima and minima for \( f(x, y) = x^3 - y^2 \).

The derivatives:

\[
f_x = 3x^2, \quad f_y = -2y, \quad f_{xx} = 6x, \quad f_{yy} = 2, \quad f_{xy} = 0.
\]

Again there is a single critical point, at \((0, 0)\), and

\[
D(0, 0) = f_{xx}(0, 0) f_{yy}(0, 0) - (f_{xy}(0, 0))^2 = 2 - 2 = 0 = 0,
\]

so there is neither a maximum nor minimum there, and so there are no local maxima or minima. The surface is shown in figure 14.7.1.

**EXAMPLE 14.7.6** Suppose a box with no top is to hold a certain volume \(V\). Find the length of the diagonal of a box is to be 1 meter; find the maximum possible volume. The box is placed with one corner at the origin, and sides along the axes, the length of the diagonal is \( \sqrt{x^2 + y^2 + z^2} \), and the volume is

\[
V = xyz = \sqrt{1 - x^2 - y^2}.
\]

Clearly, \(x^2 + y^2 \leq 1\), so the domain we are interested in is the quarter of the unit disk in the first quadrant.

**Exercises 14.7.**

1. Find all local maximum and minimum points of \( f = x^2 + 4y^2 - 2x + 8y - 1 \).
2. Find all local maximum and minimum points of \( f = x^2 - y^2 + 6x - 10y + 2 \).
3. Find all local maximum and minimum points of \( f = x^2 + y^2 \).
4. Find all local maximum and minimum points of \( f = 3x - y + 3y^3 \).

Figure 14.7.1 A saddle point, neither a maximum nor a minimum. (AP)
14.7 Maxima and minima

5. Find all local maximum and minimum points of \( f = x^2 + 4xy + y^2 - 6y + 1 \).

6. Find all local maximum and minimum points of \( f = x^2 - 9y + 2y^2 - 5x + 6y - 9 \).

7. Find the absolute maximum and minimum points of \( f = x^2 + 3y - 3xy \) over the region bounded by \( y = x \), \( y = 0 \), and \( x = 2 \).

8. A six-sided rectangular box is to hold 1/2 cubic meter; what shape should the box be to minimize surface area?

9. The post office will accept packages whose combined length and girth is at most 130 inches. What is the largest volume that can be sent in a rectangular box?

10. The bottom of a rectangular box costs twice as much per unit area as the sides and top. Find the shape for the given volume that will minimize cost.

11. Using the methods of this section, find the shortest distance from the origin to the plane \( x + y + z = 10 \).

12. Using the methods of this section, find the shortest distance from the point \((x_0, y_0, z_0)\) to the plane \( ax + by + cz = d \). You may assume that \( c \neq 0 \); use of Sage or similar software is recommended.

13. A trough is to be formed by bending up two sides of a long metal rectangle so that the cross-section of the trough is an isosceles trapezoid, as in Figure 6.2.6. If the width of the metal sheet is 2 meters, how should it be bent to maximize the volume of the trough?

14.8 Lagrange Multipliers

Many applied max/min problems take the form of the last two examples; we want to find an extreme value of a function, like \( V = xyz \), subject to a constraint, like \( f = \sqrt{x^2 + y^2} \). Often this can be done, as we have, by explicitly combining the equations and then finding critical points. There is another approach that is often convenient, the method of Lagrange multipliers.

It is somewhat easier to understand two variable problems, so we begin with one as an example. Suppose the perimeter of a rectangle is to be 100 units. Find the rectangle with largest area. This is a fairly straightforward problem from single-variable calculus. We write down the two equations: \( A = xy \), \( P = 2x + 2y \), solve the second of these for \( y \) or \( x \), substitute into the first, and end up with a one-variable maximization problem. Let’s now think of it differently: the equation \( A = xy \) defines a surface, and the equation \( 100 = 2x + 2y \) defines a curve (a line, in this case) in the \( x-y \) plane. If we graph both of these in the three-dimensional coordinate system, we can phrase the problem like this: what is the highest point on the surface above the line? The solution we already understand effectively produces the equation of the cross-section of the surface above the line and then treats it as a single-variable problem. Instead, imagine that we draw the level curves (the contour lines) for the surface in the \( x-y \) plane, along with the line. Imagine that the line represents a hiking trail and the contour lines are, as on a topographic map, the lines of constant altitude. How could you estimate, based on the graph, the high (or low) points on the path? As the path crosses contour lines, you know the path must be increasing or decreasing in elevation. At some point you will see the path just touch a contour line (tangent to it), and then begin to cross contours in the opposite direction—that point of tangency must be a maximum or minimum point. If we can identify all such points, we can then check them to see which gives the maximum and which gives the minimum value. As usual, we also need to check boundary points; in this problem, we know that \( x \) and \( y \) are positive, so we are interested in just the portion of the line in the first quadrant, as shown. The endpoints of the path, the two points on the axes, are not points of tangency, but they are the two places that the function \( f \) is a minimum in the first quadrant.

How can we actually make use of this? At the points of tangency that we seek, the constraint curve (in this case the line) and the level curve have the same slope—their tangent lines are parallel. This also means that the constraint curve is perpendicular to the gradient vector of the function; going a bit further, if we can express the constraint curve itself as a level curve, then we seek the points at which the two level curves have parallel gradients. The curve 100 = 2x + 2y can be thought of as a level curve of the function 2x + 2y; figure 14.8.2 shows both sets of level curves on a single graph. We are interested in those points where two level curves are tangent—but there are many such points, in fact an infinite number, as we’ve only shown a few of the level curves. All along the line \( y = x \) are points at which two level curves are tangent. While this might seem to be a show-stopper, it is not. The gradient of 2x + 2y is \((2, 2)\), and the gradient of \( f \) is \((\lambda, \lambda)\). They are parallel when \((2, 2) = \lambda(\lambda, \lambda)\), that is, when \(2 = \lambda y \) and \(2 = \lambda x \). We have two equations in two unknowns, which typically results in many solutions (as we expected). A third equation will reduce the number of solutions; the third equation is the original constraint, 100 = 2x + 2y.

14. Chapter 14 Partial Differentiation

15. Given the three points (1, 4), (5, 2), and (3, -2), \((x - 1)^2 + (y - 4)^2 + (x - 5)^2 + (y - 2)^2 + (x - 3)^2 + (y + 2)^2\) is the sum of the squares of the distances from point \((x, y)\) to the three points. Find \(x\) and \(y\) so that this quantity is minimized. 

16. Suppose that \( f(x, y) = x^2 + 2xy + y^2 \). Find and classify the critical points, and discuss how they change when \( k \) takes on different values.

17. Find the shortest distance from the point \((0, 0)\) to the parabola \( y = x^2 \).

18. Find the shortest distance from the point \((0, 0, 0)\) to the paraboloid \( z = x^2 + y^2 \).

19. Find the volume of the largest rectangular box with edges parallel to the axes that can be inscribed in the ellipsoid \(2x^2 + 72y^2 + 18z^2 = 288 \).

14.8 Lagrange Multipliers

14.8.1 Recall example 14.7.8: the diagonal of a box is 1; we seek to maximize the volume. The constraint is \(1 = \sqrt{x^2 + y^2 + z^2} \), which is the same as \(1 = \sqrt{x^2 + y^2 + z^2} \). So we have the following system to solve:

\[
\begin{align*}
2 &= 3y \\
2 &= 2x \\
100 &= 2x + 2y.
\end{align*}
\]

In the first two equations, \(\lambda\) can’t be 0, so we may divide it by it to get \(x = y = 2/\lambda\). Substituting into the third equation we get

\[
\frac{2}{\lambda} + \frac{2}{\lambda} + 100 = \lambda
\]

so \(x = y = 25\). Note that we are not really interested in the value of \(\lambda\)—it is a clever tool, the Lagrange multiplier, introduced to solve the problem. In many cases, however, it is easier to find \(\lambda\) than to find everything else without using \(\lambda\).

The same method works for functions of three variables, except of course everything is one dimension higher: the function to be optimized is a function of three variables and the constraint represents a surface—for example, the function may represent temperature, and we may be interested in the maximum temperature on some surface, like a sphere. The points we seek are those at which the constraint surface is tangent to a level surface of the function. Once again, we consider the constraint surface to be a level surface of some function, and we look for points at which the two gradients are parallel, giving us three equations in four unknowns. The constraint provides a fourth equation.
$x^2 + y^2 + z^2$. The function to maximize is $xyz$. The two gradient vectors are $(2x, 2y, 2z)$ and $(y, x, xy)$, so the equations to be solved are

\[
\begin{align*}
yz &= 2x\lambda \\
xz &= 2y\lambda \\
xy &= 2z\lambda \\
1 &= x^2 + y^2 + z^2
\end{align*}
\]

If $\lambda = 0$ then at least two of $x$, $y$, $z$ must be 0, giving a volume of 0, which will not be the maximum. If we multiply the first two equations by $x$ and $y$ respectively, we get

\[
\begin{align*}
xyz &= 2x^2\lambda \\
xyz &= 2y^2\lambda
\end{align*}
\]

so $2x\lambda = 2y\lambda$ or $x^2 = y^2$, in the same way we can show $x^2 = z^2$. Hence the fourth equation becomes $1 = x^2 + x^2 + x^2$ or $x = 1/\sqrt{3}$, and so $x = y = z = 1/\sqrt{3}$ gives the maximum volume. This is of course the same answer we obtained previously.

Another possibility is that we have a function of three variables, and we want to find a maximum or minimum value not on a surface but on a curve; often the curve is the intersection of two surfaces, so that we really have two constraint equations, say $g(x, y, z) = c_1$ and $h(x, y, z) = c_2$. It turns out that at points on the intersection of the surfaces where $f$ has a maximum or minimum value,

\[
\nabla f = \lambda \nabla g + \mu \nabla h
\]

As before, this gives us three equations, one for each component of the vectors, but now in five unknowns, $x$, $y$, $z$, $\lambda$, and $\mu$. Since there are two constraint functions, we have a total of five equations in five unknowns, and so can usually find the solutions we need.

**Example 14.8.2** The plane $x + y - z = 1$ intersects the cylinder $x^2 + y^2 = 1$ in an ellipse. Find the points on the ellipse closest to and furthest from the origin.

We want the extreme values of $f = \sqrt{x^2 + y^2 + z^2}$ subject to the constraints $g = x^2 + y^2 = 1$ and $h = x + y - z = 1$. To simplify the algebra, we may use instead $f = x^2 + y^2 + z^2$, since this has a maximum or minimum value at exactly the points at which $\nabla f = \lambda \nabla g + \mu \nabla h$ does. The gradients are

\[
\begin{align*}
\nabla f &= (2x, 2y, 2z) \\
\nabla g &= (2x, 2y, 0) \\
\nabla h &= (1, 1, -1)
\end{align*}
\]

\[
1 = x^2 + y^2 = 1
\]

Solving these gives $x = 1, y = 0, x = 0, y = 1$, so the points of interest are $(1,0,0)$ and $(0,1,0)$, which are both distance 1 from the origin. If $x = y$, the fourth equation is $2x^2 = 1$, giving $x = y = \pm 1/\sqrt{2}$, and from the fifth equation we get $z = -1 \pm \sqrt{2}$. The distance from the origin to $(1/\sqrt{2}, 1/\sqrt{2}, -1)$ is $\sqrt{\frac{1}{2} + \frac{1}{2} + 1} = 1.08$ and the distance from the origin to $(-1/\sqrt{2}, -1/\sqrt{2}, -1)$ is $\sqrt{\frac{1}{2} + \frac{1}{2} + 1} = 2.08$. Thus, the points $(1,0,0)$ and $(0,1,0)$ are closest to the origin and $(-1/\sqrt{2}, -1/\sqrt{2}, -1)$ is farthest from the origin. This applet shows the cylinder, the plane, the four points of interest, and the origin.

**Exercises 14.8**

1. A six-sided rectangular box is to hold 1 cubic meter; what shape should the box be to minimize surface area?
2. The post office will accept packages whose combined length and girth are at most 130 inches (girth is the maximum distance around the package perpendicular to the length). What is the largest volume that can be sent in a rectangular box?
3. The bottom of a rectangular box costs twice as much per unit area as the sides and top. Find the shape for a given volume that will minimize cost.
4. Using Lagrange multipliers, find the shortest distance from the point $(x_0, y_0, z_0)$ to the plane $ax + by + cz = d$.
5. Find all points on the surface $xy - x^2 + 1 = 0$ that are closest to the origin.
6. The material for the bottom of an aquarium costs half as much as the high strength glass for the four sides. Find the shape of the cheapest aquarium that holds a given volume $V$.
7. The plane $x - y + z = 2$ intersects the cylinder $x^2 + y^2 = 4$ in an ellipse. Find the points on the ellipse closest to and farthest from the origin.
8. Find three positive numbers whose sum is 48 and whose product is as large as possible.
9. Find all points on the plane $x + y + z = 5$ in the first octant at which $f(x, y, z) = xy^2 z^3$ has a maximum value.

10. Find the points on the surface $x^2 - y^2 = 5$ that are closest to the origin.
11. A manufacturer makes two models of an item, standard and deluxe. It costs 840 to manufacture the standard model and 860 for the deluxe. A market research firm estimates that if the standard model is priced at $x$ dollars and the deluxe at $y$ dollars, then the manufacturer will sell $500(y - x)$ of the standard items and 45,000 + $500(x - 2y)$ of the deluxe each year. How should the items be priced to maximize profit?
12. A length of sheet metal is to be made into a water trough by bending up two sides as shown in figure 14.8.3. Find $x$ and $\phi$ so that the trapezoid-shaped cross section has maximum area, when the width of the metal sheet is 27 inches (that is, $2x + y = 27$).

**Figure 14.8.3** Cross-section of a trough.

13. Find the maximum and minimum values of $f(x, y, z) = 6x + 3y + 2z$ subject to the constraint $g(x, y, z) = 4x^2 + 2y^2 + z^2 - 70 = 0$.
14. Find the maximum and minimum values of $f(x, y) = x^3$ subject to the constraint $g(x, y) = x^2 + y^2 = 16$. 
15. Find the maximum and minimum values of $f(x, y) = xy + \sqrt{y^2 - x^2 - y^2}$ when $x^2 + y^2 \leq 9$.
16. Find three real numbers whose sum is 9 and the sum of whose squares is as small as possible.
17. Find the dimensions of the closed rectangular box with maximum volume that can be inscribed in the unit sphere.