

## 14.1 Functions of Several Variables

### New Definitions

**Level Curves**  $f(x, y) = k$  where  $k$  is a constant.

**Level Surfaces**  $f(x, y, z) = k$

## 14.2 Limits and Continuity

### Definition of Limits

Let  $f$  be a function of two variables whose domain  $D$  includes points arbitrarily close to  $(a, b)$ .

$$\lim_{(x,y) \rightarrow (a,b)} f(x, y) = L \quad (1)$$

if for every number  $\varepsilon > 0$  there is a corresponding number  $\delta > 0$  such that

$$\text{if } (x, y) \in D \text{ and } 0 < \sqrt{(x-a)^2 + (y-b)^2} < \delta \text{ then } |f(x, y) - L| < \varepsilon \quad (2)$$

### When the Limit Does Not Exist

If we can find two different paths of approach along which the function  $f(x, y)$  has different limits, then it follows that  $\lim_{(x,y) \rightarrow (a,b)} f(x, y)$  does not exist.

**Limit Laws** Approaching along axes and a constant

$$\lim_{(x,y) \rightarrow (a,b)} x = a \quad \lim_{(x,y) \rightarrow (a,b)} y = b \quad \lim_{(x,y) \rightarrow (a,b)} c = c \quad (3)$$

The **Squeeze Theorem** also holds.

### Continuity

The direct substitution property: A function  $f$  is called **continuous at**  $(a, b)$  if

$$\lim_{(x,y) \rightarrow (a,b)} f(x, y) = f(a, b) \quad (4)$$

We say  $f$  is **continuous on**  $D$  if  $f$  is continuous at every point  $(a, b)$  in  $D$ .

## 14.3 Partial Derivatives

### Partial Derivatives

$$f_x(a, b) = g'(a) \quad \text{where} \quad gx = f(x, b) \quad (1)$$

### Definition of Derivatives

The partial derivatives are defined by

$$\begin{aligned} f_x(x, y) &= \lim_{h \rightarrow 0} \frac{f(x+h, y) - f(x, y)}{h} \\ f_y(x, y) &= \lim_{h \rightarrow 0} \frac{f(x, y+h) - f(x, y)}{h} \end{aligned} \quad (2)$$

### Notation

If  $z = f(x, y)$

$$f_x(x, y) = f_x = \frac{\partial f}{\partial x} = \frac{\partial}{\partial x} f(x, y) = \frac{\partial z}{\partial x} = f_1 = D_1 f = D_x f$$

$$f_y(x, y) = f_y = \frac{\partial f}{\partial y} = \frac{\partial}{\partial y} f(x, y) = \frac{\partial z}{\partial y} = f_2 = D_2 f = D_y f$$

### Finding Partial Derivatives

To find  $f_x$ , treat  $y$  as a constant and differentiate  $f(x, y)$  with respect to  $x$ .

To find  $f_y$ , treat  $x$  as a constant and differentiate  $f(x, y)$  with respect to  $y$ .

### Interpretations

The partial derivatives  $f_x(a, b)$  and  $f_y(a, b)$  can be interpreted geometrically as the slopes of the tangent lines at  $P(a, b, c)$  to the traces  $C_1$  and  $C_2$  of  $S$  in the planes  $y = b$  and  $x = a$ .

Partial derivatives can also be interpreted as *rates of change*. If  $z = f(x, y)$ , then  $\partial z / \partial x$  represents the rate of change of  $z$  with respect to  $x$  when  $y$  is fixed. Similarly,  $\partial z / \partial y$  represents the rate of change of  $z$  with respect to  $y$  when  $x$  is fixed.

### More Than Two Variables

Treated similarly to two variable functions. If  $w = f(x, y, z)$ , then  $f_x = \partial w / \partial x$  with respect to  $x$  when  $y$  and  $z$  are held constant. Similar notation follows.

## Higher Derivatives

**Second Partial Derivatives** The partial derivatives of the partial derivatives,  $(f_x)_x, (f_x)_y, (f_y)_x, (f_y)_y$ . The notation is similar to the notation of higher derivatives of functions of one variable.

**Clairaut's Theorem** Suppose  $f$  is defined on a disk  $D$  that contains the point  $(a, b)$ . If the functions  $f_{xy}$  and  $f_{yx}$  are both continuous on  $D$ , then

$$f_{xy}(a, b) = f_{yx}(a, b) \quad (3)$$

## 14.4 Tangent Planes and Linear Approximations

### Tangent Planes

Suppose a surface  $S$  had equation  $z = f(x, y)$  with continuous first derivatives. Let  $C_1$  and  $C_2$  be the curves obtained by intersecting the vertical planes  $y = y_0$  and  $x = x_0$  with the surface  $S$ . Let  $T_1$  and  $T_2$  be the tangent lines to the curves  $C_1$  and  $C_2$  at  $P$ .

**Tangent Plane** The plane to  $S$  at  $P$  that contains tangent lines  $T_1$  and  $T_2$ .

At  $P(x_0, y_0, z_0)$

$$z - z_0 = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) \quad (1)$$

### Linear Approximations

At point  $(a, b, f(a, b))$ , the **linearization** is

$$z = L(x, y) = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b) \quad (2)$$

The approximation of  $f(x, y)$  is called the **linear approximation**.

### Differentiable Functions

**Theorem for differentiability** If the partial derivatives  $f_x$  and  $f_y$  exist near  $(a, b)$  and are continuous at  $(a, b)$ , then  $f$  is differentiable at  $(a, b)$ .

Even if *directional derivatives* exist at a point in every direction, the function still may not be differentiable at that point. But if a function is differentiable at a point, then *all* directional derivatives will exist at that point.

### Three or More Variables

Defined similarly.

## 14.5 The Chain Rule

### Chain Rule for Functions of Two Variables

**Chain Rule Case 1** Suppose  $z = f(x, y)$ ,  $x = g(t)$ ,  $y = h(t)$ , and all are differentiable. Then

$$\frac{dz}{dt} = \frac{\partial z}{\partial x} \frac{dx}{dt} + \frac{\partial z}{\partial y} \frac{dy}{dt} \quad (1)$$

**Chain Rule Case 2** Suppose  $z = f(x, y)$ ,  $x = g(s, t)$ ,  $y = h(s, t)$ , and all are differentiable. Then

$$\frac{dz}{ds} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial s} \quad \frac{dz}{dt} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial t} \quad (2)$$

A tree diagram can help remember the Chain Rule (supposedly).

### Implicit Differentiation

Writing  $F(x, y) = 0$  which defines  $y$  implicitly in terms of  $x$ , that is  $y = f(x)$  so  $F(x, f(x)) = 0$ , then

$$\frac{dy}{dx} = - \frac{\frac{\partial F}{\partial x}}{\frac{\partial F}{\partial y}} = - \frac{F_x}{F_y} \quad (3)$$

For  $z$  given implicitly as  $z = f(x, y)$  by  $F(x, y, z) = 0$ , that is  $F(x, y, f(x, y)) = 0$ , we get similar formulas:

$$\frac{\partial z}{\partial x} = - \frac{F_x}{F_z} \quad \frac{\partial z}{\partial y} = - \frac{F_y}{F_z} \quad (4)$$

Note in both cases the fraction appears inverted, but that is a consequence of solving for the implicit derivative(s).

## 14.6 Directional Derivatives and the Gradient Vector

### Directional Derivatives

For differentiable  $f$ ,  $f$  has a directional derivative in the direction of the unit vector  $\vec{u} = \langle a, b \rangle$  and

$$D_{\vec{u}}f(x, y) = f_x(x, y)a + f_y(x, y)b \quad (1)$$

If  $\vec{u}$  makes angle  $\theta$  with the positive x-axis, we can write  $\vec{u} = \langle \cos \theta, \sin \theta \rangle$ .

**The Gradient Vector** Since  $D_{\vec{u}}f(x, y) = \langle f_x(x, y), f_y(x, y) \rangle \cdot \langle a, b \rangle$ , we give the first vector a special name, **grad**  $f$  or  $\nabla f$  ("del" or "nabla").

$$\nabla f(x, y) = \langle f_x(x, y), f_y(x, y) \rangle \quad (2)$$

$$D_{\vec{u}}f(x, y) = \nabla f(x, y) \cdot \vec{u} \quad (3)$$

### Functions of Three Variables

Defined similarly.

$$\nabla f = \langle f_x, f_y, f_z \rangle \quad (4)$$

$$D_{\vec{u}}f(x, y, z) = \nabla f(x, y, z) \cdot \vec{u} \quad (5)$$

### Maximizing the Directional Derivatives

Suppose  $f$  is differentiable.

$$\max(D_{\vec{u}}f(\vec{x})) = |\nabla f(\vec{x})| \quad (6)$$

This occurs when  $\vec{u}$  has the same direction as  $\nabla f(\vec{x})$ , or in my own notation

$$\vec{u} = k\nabla f(\vec{x}) \quad (7)$$

Normalize  $\nabla f(\vec{x})$  to get  $\vec{u}$ .

### Tangent Planes to Level Surfaces

If  $\nabla F(x_0, y_0, z_0) \neq 0$ , define the **tangent plane to the level surface**  $F(x, y, z) = k$  at  $P(x_0, y_0, z_0)$  as the plane that passes through  $P$  and has normal vector  $\nabla F(x_0, y_0, z_0)$ .

Using the standard equation of a plane, we can write

$$F_x(x_0, y_0, z_0)(x - x_0) + F_y(x_0, y_0, z_0)(y - y_0) + F_z(x_0, y_0, z_0)(z - z_0) = 0 \quad (8)$$

The **normal line** is perpendicular to the tangent plane and given by  $\nabla F(x_0, y_0, z_0)$ .  
It has symmetric equations

$$\frac{x - x_0}{F_x(x_0, y_0, z_0)} = \frac{y - y_0}{F_y(x_0, y_0, z_0)} = \frac{z - z_0}{F_z(x_0, y_0, z_0)} \quad (9)$$

## 14.7 Maximum and Minimum Values

### Local maximum and minimum

If  $f$  has a local maximum or minimum at  $(a, b)$  and the partial derivatives exist, then

$$f_x(a, b) = 0 \quad \text{and} \quad f_y(a, b) = 0 \quad (1)$$

**Critical Point** More generally, a point  $(a, b)$  is called a critical point if  $f_x(a, b) = 0$  and  $f_y(a, b) = 0$ , or if one of these partial derivatives does not exist.

### 2nd Derivative Test

If  $f_x(a, b) = 0$  and  $f_y(a, b) = 0$ , let

$$D = D(a, b) = f_{xx}(a, b)f_{yy}(a, b) - (f_{xy}(a, b))^2 \quad (2)$$

- (a) If  $D > 0$  and  $f_{xx}(a, b) > 0$ , then  $f(a, b)$  is a local minimum.
- (b) If  $D > 0$  and  $f_{xx}(a, b) < 0$ , then  $f(a, b)$  is a local maximum.
- (c) If  $D < 0$ , then  $f(a, b)$  is a **saddle point** of  $f$ .

### Absolute Maximum and Minimum Values

A **closed set** is one that contains all its boundary points.

**Extreme Value Theorem** There exists an absolute maximum and absolute minimum on a closed, bounded set  $D$ .

**Closed Interval Method** To find the absolute maximum and minimum values of  $f$  on  $D$ :

1. Find the values of  $f$  at the critical points of  $f$  in  $D$ .
2. Find the extreme values of  $f$  on the boundary of  $D$ .
3. The largest of the values from steps 1 and 2 is the absolute maximum, the smallest is the absolute minimum.



## 14.8 Lagrange Multipliers

Geometrically, the maximum or minimum of  $f(x, y)$  subject to  $g(x, y) = k$  is when the two curves touch, that is when they have a common tangent line.

$$\nabla f(x_0, y_0, z_0) = \lambda \nabla g(x_0, y_0, z_0) \quad (1)$$

Lambda is called a **Lagrange multiplier**.

In component form,

$$f_x = \lambda g_x \quad f_y = \lambda g_y \quad f_z = \lambda g_z \quad g(x, y, z) = k \quad (2)$$

This is a system of four equations and four unknowns. For functions of two variables, a similar equation holds.

### Two Constraints

For  $f(x, y, z)$  subject to two constraints  $g(x, y, z) = k$  and  $h(x, y, z) = c$ ,

$$\nabla f(x_0, y_0, z_0) = \lambda \nabla g(x_0, y_0, z_0) + \mu \nabla h(x_0, y_0, z_0) \quad (3)$$

In component form,

$$\begin{aligned} f_x &= \lambda g_x + \mu h_x \\ f_y &= \lambda g_y + \mu h_y \\ f_z &= \lambda g_z + \mu h_z \\ g(x, y, z) &= k \\ h(x, y, z) &= c \end{aligned} \quad (4)$$