

1 Functions of Several Variables

1.1 Graphs and Level Curves

Definition

A function $z = f(x, y)$ assigns to each point (x, y) in a set D in \mathbb{R}^2 a unique real number z in a subset of \mathbb{R} . The set D is the domain of f . The range of f is the set of all real numbers z that are assumed as the points (x, y) vary over the domain.

Example

Find the domain of

$$f(x, y) = \frac{1}{\sqrt{x^2 + y^2 - 25}}$$

Note that the denominator cannot be zero or is negative.

Therefore, the domain is $D = (x, y) : x^2 + y^2 > 25$.

Definition

For a function of two variables $f(x, y)$ a level curve is the set of points (x, y) in the xy -plane where $f(x, y)$ is equal to a constant z_0 .

We can extend our definitions to functions of more than two variables.

Definition

For a function of three variables $f(x, y, z)$ a level surface is the set of points (x, y, z) in xyz -space where $f(x, y, z)$ is equal to a constant w_0 .

We can describe level surfaces as well.

1.2 Limits and Continuity

We can write the limit of two variables as:

$$\lim_{(x,y) \rightarrow (a,b)} f(x, y) = \lim_{P \rightarrow P_0} f(x, y) = L$$

There are some laws of limits.

Theorem 1.1

Let a, b , and c be real numbers.

- Constant function $f(x, y) = c$:

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

- Linear function $f(x, y) = x$:

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

- Linear function $f(x, y) = y$:

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

If we let the limit of a function $f(x, y) = L$ and the limit of a function $g(x, y) = M$, we can see:

- The sum of the limits is $L + M$.
- The difference of the limits is $L - M$.
- If we multiply a function by a constant c , the limit is cL .
- The product of the limits is LM .
- The difference of the limits is $\frac{L}{M}$.
- Putting a limit to a power n is L^n .

Definition

Let R be a region in \mathbb{R}^2 . An interior point P of R lies entirely within R , which means it is possible to find a disk centered at P that contains only points of R .

A boundary point Q of R lies on the edge of R in the sense that every disk centered at Q contains at least one point in R and at least one point not in R .

Definition

A region is open if it consists entirely of interior points. A region is closed if it contains all its boundary points.

The two-path test for nonexistence of limits:

If $f(x, y)$ approaches the two different values as (x, y) approaches (a, b) along two different paths in the domain of f , then the limit does not exist.

Definition

The function f is continuous at the point (a, b) provided:

- f is defined at (a, b) .
- The limit for f exists.
- The two quantities must be equal.

1.3 Partial Derivatives

The idea of a partial derivative is to differentiate f with respect to one variable, treating the other variable as a constant.

Definition

The partial derivative of f with respect to x at the point (a, b) is:

$$f_x(a, b) = \lim_{h \rightarrow 0} \frac{f(a + h, b) - f(a, b)}{h}$$

The partial derivative of f with respect to y at the point (a, b) is:

$$f_y(a, b) = \lim_{h \rightarrow 0} \frac{f(a, b + h) - f(a, b)}{h}$$

provided these limits exist.

The notation for these are:

$$f_x = \frac{\partial f}{\partial x} \quad f_y = \frac{\partial f}{\partial y}$$

We can first partially differentiate f with respect to x and then with respect to x denoted as:

$$\frac{\partial^2 f}{\partial x^2}$$

We can do it first with respect to x and then with respect to y denoted as:

$$\frac{\partial^2 f}{\partial y \partial x}$$

We can also differentiate with respect to y and then with respect to x :

$$\frac{\partial^2 f}{\partial x \partial y}$$

Lastly, we can differentiate with respect to y twice:

$$\frac{\partial^2 f}{\partial y^2}$$

Theorem 1.2: Equality of Mixed Partial Derivatives

Assume f is defined on an open set D of \mathbb{R}^2 , and that f_{xy} and f_{yx} are continuous throughout D . Then $f_{xy} = f_{yx}$ at all points of D .

1.4 The Chain Rule

Theorem 1.3

Let z be a differentiable function on x and y on its domain, where x and y are differentiable functions of t on an interval I . Then

$$\frac{dz}{dt} = \frac{\partial z}{\partial x} \frac{dx}{dt} + \frac{\partial z}{\partial y} \frac{dy}{dt}$$

Let's suppose that we have a function $w = f(x, y, z)$ where x, y, z depend on t .

Then we have:

$$\frac{dw}{dt} = \frac{\partial w}{\partial x} \frac{dx}{dt} + \frac{\partial w}{\partial y} \frac{dy}{dt} + \frac{\partial w}{\partial z} \frac{dz}{dt}$$

Theorem 1.4

Let z be a differentiable function of x and y , where x and y are differentiable functions of s and t . Then

$$\frac{\partial z}{\partial s} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial s}$$

and

$$\frac{\partial z}{\partial t} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial t}$$

Theorem 1.5

Let F be differentiable on its domain and suppose $F(x, y) = 0$ defines y as a differentiable function of x . Provided $F_y \neq 0$,

$$\frac{dy}{dx} = -\frac{F_x}{F_y}$$

1.5 Directional Derivatives and the Gradient

Suppose we have the function $f(x, y)$.

$f_x(x, y)$ measures the rate of change in the direction of positive x -axis and $f_y(x, y)$ measures the rate of change in the direction of positive y -axis.

Definition

Let f be differentiable at (a, b) and let $\mathbf{u} = \langle u_1, u_2 \rangle$ be a unit vector in the xy -plane. The directional derivative of f at (a, b) in the direction of \mathbf{u} is

$$D_{\mathbf{u}}f(a, b) = \lim_{h \rightarrow 0} \frac{f(a + hu_1, b + hu_2) - f(a, b)}{h}$$

provided the limit exists.

The directional derivative $D_{\mathbf{u}}f(a, b)$ measures the rate of change of f at the point (a, b) in the direction of \mathbf{u} .

Theorem 1.6

Let f be differentiable at (a, b) and let $\mathbf{u} = \langle u_1, u_2 \rangle$ be a unit vector in the xy -plane. The directional derivative of f at (a, b) in the direction of \mathbf{u} is

$$D_{\mathbf{u}}f(a, b) = \langle f_x(a, b), f_y(a, b) \rangle \cdot \langle u_1, u_2 \rangle$$

Definition

Let f be differentiable at the point (x, y) . The gradient of f at (x, y) is the vector-valued function

$$\nabla f(x, y) = \langle f_x(x, y), f_y(x, y) \rangle = f_x(x, y)\mathbf{i} + f_y(x, y)\mathbf{j}$$

The applications of this are:

Directional derivatives: $D_{\mathbf{u}}f(a, b) = \nabla f(a, b) \cdot \mathbf{u}$

The gradient gives the direction of maximum increase of a function.

Gradient is normal to level curves.

Theorem 1.7

Let f be differentiable at (a, b) with $\nabla f(a, b) \neq 0$.

f has its maximum rate of increase at (a, b) in the direction of the gradient $\nabla f(a, b)$. The rate of change in this direction is $|\nabla f(a, b)|$. (This will be maximized if $\theta = 0$.)

f has its maximum rate of decrease at (a, b) in the direction of $-\nabla f(a, b)$. The rate of change in this direction is $-|\nabla f(a, b)|$. (This will be minimized if $\theta = \pi$.)

The directional derivative is zero in any direction orthogonal to $\nabla f(a, b)$.

Theorem 1.8

Given a function f differentiable at (a, b) , the line tangent to the level curve of f at (a, b) is orthogonal to the gradient $\nabla f(a, b)$, provided $\nabla f(a, b) \neq 0$.

We can now consider gradients of functions of three variables.

Theorem 1.9

Let f be differentiable at (a, b, c) and let $\mathbf{u} = \langle u_1, u_2, u_3 \rangle$ be a unit vector. The directional derivative of f at (a, b, c) in the direction of \mathbf{u} is

$$\begin{aligned} D_{\mathbf{u}}f(a, b, c) &= \nabla f(a, b, c) \cdot \mathbf{u} \\ &= \langle f_x(a, b, c), f_y(a, b, c), f_z(a, b, c) \rangle \cdot \langle u_1, u_2, u_3 \rangle \end{aligned}$$

Assuming $\nabla f(a, b, c) \neq 0$, the gradient in three dimensions has the following properties.

f has its maximum rate of increase at (a, b, c) in the direction of the gradient $\nabla f(a, b, c)$, and the rate of change in this direction is $|\nabla f(a, b, c)|$.

f has its maximum rate of decrease at (a, b, c) in the direction of $-\nabla f(a, b, c)$, and the rate of change in this direction is $-|\nabla f(a, b, c)|$.

The directional derivative is zero in any direction orthogonal to $\nabla f(a, b, c)$.

1.6 Tangent Planes and Linear Approximation

Definition

Let F be differentiable at the point $P_0(a, b, c)$ with $\nabla F(a, b, c) \neq 0$. The plane tangent to the surface $F(x, y, z) = 0$ at P_0 , called the tangent plane, is the plane passing through P_0 orthogonal to $\nabla F(a, b, c)$. An equation of the tangent plane is

$$F_x(a, b, c)(x - a) + F_y(a, b, c)(y - b) + F_z(a, b, c)(z - c) = 0$$

Definition

Let f be differentiable at the point (a, b) . An equation of the plane tangent to the surface $z = f(x, y)$ at the point $(a, b, f(a, b))$ is

$$z = f_x(a, b)(x - a) + f_y(a, b)(y - b) + f(a, b)$$

One of the main applications of tangent planes is linear approximation for a function near a point.

Definition

Let f be differentiable at (a, b) . The linear approximation to the surface $z = f(x, y)$ at the point $(a, b, f(a, b))$ is the tangent plane at that point, given by the equation

$$L(x, y) = f_x(a, b)(x - a) + f_y(a, b)(y - b) + f(a, b)$$

For a function of three variables, the linear approximation to $w = f(x, y, z)$ at the point $(a, b, c, f(a, b, c))$ is given by

$$L(x, y, z) = f_x(a, b, c)(x - a) + f_y(a, b, c)(y - b) + f_z(a, b, c)(z - c) + f(a, b, c)$$

The idea for linear approximations of functions gives the definition of the differential.

Definition

Let f be differentiable at the point (x, y) . The change in $z = f(x, y)$ as the independent variables change from (x, y) to $(x + dx, y + dy)$ is denoted Δz and is approximated by the differential dz .

$$\Delta z \approx dz = f_x(x, y)dx + f_y(x, y)dy$$

Note that Δz is the change in f on the surface and is equal to $f(x, y) - f(a, b)$.

Note that dz is the change in l on the tangent plane and is equal to $L(x, y) - f(a, b)$.

1.7 Maximum/Minimum Problems

Definition

Suppose (a, b) is a point in a region R on which $f(x, y)$ is defined. If $f(x, y) \leq f(a, b)$ for all (x, y) in the domain of f and in some open disk centered at (a, b) , then $f(a, b)$ is a local maximum value of f . If $f(x, y) \geq f(a, b)$ for all (x, y) in the domain of f and in some open disk centered at (a, b) , then $f(a, b)$ is a local minimum value of f . Local maximum and local minimum values are also called local extreme values or local extrema.

Theorem 1.10

If f has a local maximum or minimum value at (a, b) and the partial derivatives f_x and f_y exist at (a, b) , then $f_x(a, b) = f_y(a, b) = 0$.

Definition

An interior point (a, b) in the domain of f is a critical point of f if either:

- $f_x(a, b) = f_y(a, b) = 0$ or
- at least one of the partial derivatives f_x and f_y does not exist at (a, b) .

Theorem 1.11

Suppose the second partial derivatives of f are continuous throughout an open disk centered at the point (a, b) , where $f_x(a, b) = f_y(a, b) = 0$. Let $D(x, y) = f_{xx}(x, y)f_{yy}(x, y) - (f_{xy}(x, y))^2$.

- If $D(a, b) > 0$ and $f_{xx}(a, b) < 0$, then f has a local maximum value at (a, b) .
- If $D(a, b) > 0$ and $f_{xx}(a, b) > 0$, then f has a local minimum value at (a, b) .
- If $D(a, b) < 0$, then f has a saddle point at (a, b) .
- If $D(a, b) = 0$, then the test is inconclusive.

Definition

Let f be defined on a set R in \mathbb{R}^2 containing the point (a, b) . If $f(a, b) \geq f(x, y)$ for every (x, y) in R , then $f(a, b)$ is an absolute maximum value of f on R . If $f(a, b) \leq f(x, y)$ for every (x, y) in R , then $f(a, b)$ is an absolute minimum value of f on R .

We can describe the procedure for finding the absolute maximum/minimum values on closed bounded sets:

Let f be continuous on a closed bounded set R in \mathbb{R}^2 . To find the absolute maximum and minimum values of f on R :

- Determine the values of f at all critical points in R .
- Find the maximum and minimum values of f on the boundary of R .
- The greatest function value found in the previous steps is the absolute maximum value of f on R , and the least function value found is the absolute minimum value of f on R .

1.8 Lagrange Multipliers

The goal is to maximize (or minimize) a function $f(x, y)$ subject to the constraint $g(x, y) = 0$.

To find absolute extrema on closed and bounded constraint curves using Lagrange Multipliers:

Let the objective function f and the constraint function g be differentiable on a region of \mathbb{R}^2 with $\nabla g(x, y) \neq 0$ on the curve $g(x, y) = 0$. To locate the absolute maximum and minimum values of f subject to the constraint $g(x, y) = 0$, carry out the following steps:

1. Find the values of x , y , and λ (if they exist) that satisfy the equations

$$\nabla f(x, y) = \lambda \nabla g(x, y) \quad \text{and} \quad g(x, y) = 0$$

2. Evaluate f at the values (x, y) found in Step 1 and at the endpoints of the constraint curve (if they exist). Select the largest and smallest corresponding function values. These values are the absolute maximum and minimum values of f subject to the constraint.

(This section is kinda annoying, I don't like the amount of steps each question has) Basically, I hate optimization.