

Week 9

Partial Derivatives

Chain Rule

Differentiability

If a function $f(x, y)$ in two variables is differentiable at (a, b) , then geometrically it means that there exists a tangent plane to the graph $z = f(x, y)$ of f at the point $(a, b, f(a, b))$ on the graph. An equation which describes this tangent plane is:

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

or equivalently:

$$-f_x(a, b)(x - a) - f_y(a, b)(y - b) + (z - f(a, b)) = 0.$$

Hence, this is the plane in \mathbb{R}^3 which contains the point $(a, b, f(a, b))$, and has $\vec{n} = \langle -f_x(a, b), -f_y(a, b), 1 \rangle$ as a normal vector.

Let $f(x, y)$ be a function in 2 variables. The existence of $f_x(a, b)$ and $f_y(a, b)$ does ***not*** guarantee the differentiability of f at (a, b) . However,

Theorem.

If f_x and f_y are continuous on an open region containing (a, b) , then f is differentiable at (a, b) .

Theorem.

If f is differentiable at P , then it is continuous at P .

Higher Order Partial Derivatives

Since, $\frac{\partial f}{\partial x_i}$ is itself a function in n variables, we can consider its partial derivative with respect to any of the variables x_j . And we can further consider partial derivatives of *that* partial derivative, and so on. The notation is as follows:

$$\frac{\partial^2 f}{\partial x_i^2} = f_{x_i x_i} := \frac{\partial}{\partial x_i} \left(\frac{\partial f}{\partial x_i} \right).$$

For $j \neq i$,

$$\frac{\partial^2 f}{\partial x_j \partial x_i} = f_{x_i x_j} := \frac{\partial}{\partial x_j} \left(\frac{\partial f}{\partial x_i} \right).$$

For $m \in \mathbb{N}$,

$$\frac{\partial^m f}{\partial x_i^m} = \underbrace{f_{x_i x_i \dots x_i}}_{m \text{ times}} := \frac{\partial}{\partial x_i} \left(\frac{\partial^{m-1} f}{\partial x_i^{m-1}} \right)$$

For $i_1, i_2, \dots, i_m \in \{1, 2, 3, \dots, n\}$,

$$\frac{\partial^m f}{\partial x_{i_m} \partial x_{i_{m-1}} \partial x_{i_{m-2}} \dots \partial x_{i_1}} = f_{x_{i_1} x_{i_2} \dots x_{i_m}} := \frac{\partial}{\partial x_{i_m}} \left(\frac{\partial^{m-1} f}{\partial x_{i_{m-1}} \partial x_{i_{m-2}} \dots \partial x_{i_1}} \right).$$

Theorem.

Let x and y be two of the variables of a function f . If f_{xy} and f_{yx} are continuous on an open region containing a point P , then:

$$f_{xy}(P) = f_{yx}(P).$$

Chain Rule

Theorem. If $f(x_1, x_2, \dots, x_n)$ is a differentiable function in n variables, and each $x_i = x_i(s_1, s_2, \dots, s_m)$ ($i = 1, 2, \dots, n$) is a differentiable function in m variables, then f is differentiable as a function in s_1, s_2, \dots, s_m with:

$$\frac{\partial f}{\partial s_i} = \sum_{j=1}^n \frac{\partial f}{\partial x_j} \frac{\partial x_j}{\partial s_i}.$$

Definition.

The **gradient** of $f(x_1, x_2, \dots, x_n)$ at $P \in \mathbb{R}^n$ is:

$$\nabla f(P) = \langle f_{x_1}(P), f_{x_2}(P), \dots, f_{x_n}(P) \rangle \in \mathbb{R}^n$$

Hence,

$$\frac{\partial f}{\partial s_i} = \nabla f \cdot \frac{\partial \vec{x}}{\partial s_i},$$

where:

$$\frac{\partial \vec{x}}{\partial s_i} := \left\langle \frac{\partial x_1}{\partial s_i}, \frac{\partial x_2}{\partial s_i}, \dots, \frac{\partial x_n}{\partial s_i} \right\rangle.$$

