# Chain Rule and the Gradient

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## The gradient

**Definition:** Let f be a function of n variables:  $x_1, x_2, \ldots, x_n$ , then the **gradient** is

$$\nabla f(x_1,\ldots,x_n) = \langle \frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}\ldots, \frac{\partial f}{\partial x_n} \rangle$$

# Linear approximations for functions of n variables

For a function of n variables we can use this notation to write the linear approximation at  $\mathbf{a} = (a_1, a_2, \dots, a_n)$ 

$$L(x_1,\ldots,x_n) = f(\mathbf{a}) + \nabla f(\mathbf{a}) \cdot \langle x_1 - a_1,\ldots,x_n - a_n \rangle$$

#### Definition of differentiable

**Definition:** We say that a function of n variables, f is **differentiable** at a point  $\mathbf{a} = (a_1, a_2, \ldots, a_n)$  if all the partial derivatives  $\frac{\partial f}{\partial x_i}(\mathbf{a})$  exist and the linear approximation  $L(x_1, \ldots, x_n)$  is a **good approximation**. I.e.

$$\lim_{(x_1,\dots,x_n)\to \mathbf{a}} \frac{f(x_1,\dots,x_n) - L(x_1,\dots,x_n)}{\|(x_1,\dots,x_n) - (a_1,\dots,a_n)\|} = 0$$

#### Chain Rule - Case 1

Suppose that z = f(x, y) is a differentiable function of two variables and x = g(t) and y = h(t) are both differentiable functions of t. Then the derivative of f with respect to tis:

$$\frac{df}{dt} = \frac{\partial f}{\partial x}\frac{dx}{dt} + \frac{\partial f}{\partial y}\frac{dy}{dt}$$

$$\frac{df}{dt} = \nabla f \cdot \langle \frac{dx}{dt}, \frac{dy}{dt} \rangle$$

### Chain Rule - Case 2

Suppose that z = f(x, y) is a differentiable function of x and y and x = g(s, t) and y = h(s, t) are differentiable functions of s and t. Then

$$\frac{\partial f}{\partial s} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial s}$$
$$\frac{\partial f}{\partial t} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial t}$$

Or in vector notation

$$\frac{\partial f}{\partial s} = \nabla f \cdot \langle \frac{\partial x}{\partial s}, \frac{\partial y}{\partial s} \rangle$$
 and  $\frac{\partial f}{\partial t} = \nabla f \cdot \langle \frac{\partial x}{\partial t}, \frac{\partial y}{\partial t} \rangle$ 

#### Chain Rule in general

Suppose f is a differentiable function in nvariables:  $x_1, x_2, \ldots, x_n$  and each  $x_i$  is a differentiable function in m variables  $t_1, t_2, \ldots, t_m$ . Then f is a function of  $t_1, t_2, \ldots, t_m$  and for each  $i = 1, \cdots, m$ :

$$\frac{\partial f}{\partial t_i} = \frac{\partial f}{\partial x_1} \frac{\partial x_1}{\partial t_i} + \frac{\partial f}{\partial x_2} \frac{\partial x_2}{\partial t_i} + \dots + \frac{\partial f}{\partial x_n} \frac{\partial x_n}{\partial t_i}$$

In vector notation

$$\frac{\partial f}{\partial t_i} = \nabla f \cdot \langle \frac{\partial x_1}{\partial t_i}, \frac{\partial x_2}{\partial t_i}, \dots, \frac{\partial x_n}{\partial t_i} \rangle$$

# **Implicit Differentiation**

Suppose that y is defined implicitly by the equation

$$F(x,y) = 0$$

Then

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