# Lecture XI Notes

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## 1 The Multivariable Chain Rule – 14.5

In three dimensions, the formula for the chain rule looks as follows:

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

Conditions z = f(x, y); x = x(t), y = y(t)

Example:

$$z = x^{2}y + 3xy^{4}; x = \sin 2t, y = \cos t$$
$$\frac{dz}{dt} = (2xy + 3y^{4})(2\cos 2t) + (x^{2} + 12xy^{3})(-\sin t)$$

Another case looks as follows:

If z = f(x, y); x = g(s, t), y = h(s, t), then:

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

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

The general case of the chain rule, where u is a multivariable function of n variables  $x_1$ ,  $x_2, \ldots x_n$ , and each x is a function of m variables  $t_1, \ldots t_m$ , then:

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

### 1.1 Implicit Differentiation

If  $\frac{\partial f}{\partial y} \neq 0$ :

$$\frac{dy}{dx} = -\frac{\frac{\partial f}{\partial x}}{\frac{\partial f}{\partial y}}$$

In addition to this:

$$\frac{\partial z}{\partial x} = -\frac{\frac{\partial f}{\partial x}}{\frac{\partial f}{\partial z}}$$

and

$$\frac{\partial z}{\partial y} = -\frac{\frac{\partial f}{\partial y}}{\frac{\partial f}{\partial z}}$$

# 2 The Directional Derivative – 14.6

The definition of a directional derivative is:

$$D_{\hat{u}} = \lim_{h \to 0} \frac{f(x_o + ha, y_o + hb) - f(x_o, y_o)}{h}$$

In more simplified terms:

$$D_{\hat{u}} = \frac{\partial f}{\partial x}a + \frac{\partial f}{\partial y}b$$

## 2.1 The Gradient Vector

The gradient vector is written  $\overrightarrow{\nabla} f$ , and:

$$\overrightarrow{\nabla} f = \frac{\partial f}{\partial x} \hat{\mathbf{i}} + \frac{\partial f}{\partial y} \hat{\mathbf{j}}$$

In the case that u = f(x, y, z):

$$\overrightarrow{\nabla} f = \frac{\partial f}{\partial x} \hat{\mathbf{i}} + \frac{\partial f}{\partial y} \hat{\mathbf{j}} + \frac{\partial f}{\partial z} \hat{\mathbf{k}}$$

and

$$D_{\hat{u}} = \overrightarrow{\nabla} f(x, y, z) \cdot \hat{u}$$

The greatest value of the gradient, written as  $|\overrightarrow{\nabla} f|$ , occurs in the same direction for  $\hat{\bf u}$  and  $\overrightarrow{\nabla} f$