# Lecture 18

# 13.6. Directional Derivatives and Gradients

Goals: (1) Find and use directional derivatives of a function of two variables.

- (2) Find the gradient of a function of two variables.
- (3) Use the gradient of a function of two variables in applications.
- (4) Find directional derivatives and gradients of functions of three variables.

## Questions:

• What are the partial derivatives of a function z = f(x, y)?

### 13.6.1. Directional derivative of a function of two variables

Use the unit vector to study all directions of line paths.

(1) Definition:

Let f be a function of two variables x, y and let  $\vec{u} = \cos \theta \vec{i} + \sin \theta \vec{j}$  be a unit vector. Then the *directional derivative* of f in the direction of  $\vec{u}$ , denoted by  $D_{\vec{u}}f$  is

$$D_{\vec{u}}f(x,y) = \lim_{t \to 0} \frac{f(x + t\cos\theta, y + t\sin\theta) - f(x,y)}{t}$$

provided this limit exists.

(2) Formula 1:

$$D_{\vec{u}}f(x,y) = f_x(x,y)\cos\theta + f_y(x,y)\sin\theta$$

<u>Note</u>: There are infinitely many directional derivatives at a certain point. In particular, when  $\theta = 0$ ,  $\vec{u} = \vec{i}$ ,  $D_{\vec{i}}f = f_x$ ; when  $\theta = \frac{\pi}{2}$ ,  $\vec{u} = \vec{j}$ ,  $D_{\vec{i}}f = f_y$ .

- (3) Examples 1, 2: finding a directional derivative (p. 935)
  - Try exercises 1-8, 13-16, 17-18

#### 13.6.2. The gradient of a function of two variables

The gradient of a function is a vector that can be used to find the directional derivative of the function.

(1) Let z = f(x, y) be a function whose partial derivatives exist. Then the **gradient** of f, denoted by  $\nabla f(x, y)$ , or grad f(x, y), is the vector-valued function:

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

(2) Formula 2:

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

- (3) Example 3: finding the gradient of a function (p. 936)
  - Try exercises 21-24
- (4) Example 4: using  $\nabla f(x, y)$  to find a directional derivative (p. 937)

• Try exercises 27-30

## 13.6.3. Properties of the gradient

An advanced topic: Operator is a mapping from a set to another set. Here are three examples:

 $\nabla$  is an "operator" which maps <u>scalar-valued</u> functions to <u>vector-valued</u> functions. One can consider a regular function to be a special "operator" which maps the <u>domain</u> (numbers) to the <u>range</u> (still numbers). Another example of "operator" is  $\frac{d}{dx}$ , which maps <u>functions</u> to <u>functions</u>. That is, when you apply  $\frac{d}{dx}$  to a function, you get another function!

- (1) Let f be differentiable at the point (x, y).
  - If  $\nabla f(x,y) = \vec{0}$ , then  $D_{\vec{u}}f(x,y) = 0$  for all  $\vec{u}$ .
  - If  $\nabla f(x,y) \neq \vec{0}$ ,  $\nabla f(x,y)$  gives the direction of *maximum* increase of f. Furthermore, the *maximum* value of  $D_{\vec{u}}f(x,y)$  is  $\|\nabla f(x,y)\|$ . In other words, f increases most rapidly in the direction of the gradient  $\nabla f(x,y)$ . See Figure 13.49 (p. 937). Notice that this is a local solution, meaning that different point will lead to different maximum increase.
  - Similarly,  $-\nabla f(x, y)$  gives the direction of *minimum* increase of f. Furthermore, the *minimum* value of  $D_{\vec{u}}f(x, y)$  is  $-\|\nabla f(x, y)\|$ .
- (2) An observation of gradient:

If f is differentiable at  $(x_0, y_0)$  and  $\nabla f(x_0, y_0) \neq \vec{0}$ , then  $\nabla f(x_0, y_0)$  is *normal* to the level curve through  $(x_0, y_0)$ . See Example 6 and Figure 13.52.

- (3) Example 5: finding the direction of max increase (p. 938).
  - Try exercises 31-36
- (4) Example 6: finding the path of a heat-seeking particle (p. 939).
  - Try exercises 55-58
- (5) Example 7: finding a normal vector  $\nabla f(x, y)$  to a level curve (p. 940).
  - Try exercises 59-62

### 13.6.4. Directional derivative of a function of three variables

Notice that we are no longer able to visualize the directional derivative for a function in three variables.

(1) Definition:

Let f be a function of three variables x, y, z and let  $\vec{u} = a\vec{i} + b\vec{j} + c\vec{k}$  be a unit vector. Then the *directional derivative* of f in the direction of  $\vec{u}$ , denoted by  $D_{\vec{u}}f$  is

$$D_{\vec{u}}f(x, y, z) = \lim_{t \to 0} \frac{f(x + ta, y + tb, z + tc) - f(x, y, z)}{t}$$

provided this limit exists.

(2) Formula 3:

$$D_{\vec{u}}f(x,y,z) = af_x(x,y,z) + bf_y(x,y,z) + cf_z(x,y,z)$$

Note: Here we have  $a^2 + b^2 + c^2 = 1$ , as  $\vec{u}$  is a unit vector. Try exercises 9-12, 19-20

## 13.6.5. The gradient of a function of three variables and properties

This is a natural extension of gradient of a function of two variables.

(1) The gradient is:

$$\nabla f(x, y, z) = f_x(x, y, z) \, \vec{i} + f_y(x, y, z) \, \vec{j} + f_z(x, y, z) \, \vec{k}$$

(2) Formula 4:

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

- (3) Let f be differentiable at the point (x, y, z).
  - If  $\nabla f(x, y, z) = \vec{0}$ , then  $D_{\vec{u}} f(x, y, z) = 0$  for all  $\vec{u}$ .
  - If  $\nabla f(x,y,z) \neq \vec{0}$ ,  $\nabla f(x,y,z)$  gives the direction of *maximum* increase of f. Furthermore, the *maximum* value of  $D_{\vec{u}}f(x,y,z)$  is  $\|\nabla f(x,y,z)\|$ . In other words, f increases most rapidly in the direction of the gradient  $\nabla f(x,y,z)$ . Notice that this is a local solution, meaning that different point will lead to different maximum increase.
  - Similarly,  $-\nabla f(x, y, z)$  gives the direction of *minimum* increase of f. Furthermore, the *minimum* value of  $D_{\vec{v}}f(x, y, z)$  is  $-\|\nabla f(x, y, z)\|$ .
- (4) Example 8: finding the gradient (p. 941)
  - Try exercises 25-26, 37-40

#### 13.6.6. **Homework Set #18**

- Read 13.6 (pages 933-941).
- Do exercises on pages 942-944: 5, 7, 9, 11, 13, 15, 17, 19, 21, 23, 25, 27, 29, 31, 35, 37, 39, 53, 55, 57, 65, 73-76