

# Math 213 - Directional Derivatives and the Gradient Vector

Peter A. Perry

University of Kentucky

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# Homework

- Remember that WebWork B4 is due Wednesday
- Start working on practice problems in section 14.6, 11-23 (odd), 29, 31, 33, 42, 43, 54, 55, 57
- Re-read section 14.6
- Read ahead for Wednesday - section 14.7

## Unit II: Differential Calculus of Several Variables

- Lecture 12 Functions of Several Variables
- Lecture 13 Partial Derivatives
- Lecture 14 Tangent Planes and Linear Approximation
- Lecture 15 The Chain Rule
- Lecture 16 Directional Derivatives and the Gradient
- Lecture 17 Maximum and Minimum Values, I
- Lecture 18 Maximum and Minimum Values, II
- Lecture 19 Lagrange Multipliers
  
- Lecture 20 Double Integrals
- Lecture 21 Double Integrals over General Regions
- Lecture 22 Double Integrals in Polar Coordinates
  
- Lecture 23 Exam II Review

## Goals of the Day

- Understand what a *directional derivative* of a function of two and three variables is, and how to compute it
- Understand what the *gradient vector*  $\nabla f$  is, and how it's related to directional derivatives
- Understand that the gradient vector:
  - Points in the direction of maximum change of  $f$
  - Has magnitude equal to that maximal rate of change
  - Is perpendicular to level curves (two variables) or level surfaces (three variables)

# Review

## The Chain Rule, 2 Variables (Case 1) If

$$z = f(x, y), \quad x = g(t), \quad y = h(t),$$

then

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

Suppose that

$$f_x(1, 3) = -2, \quad f_y(1, 3) = 4,$$

$$x(t) = t, \quad y(t) = 3t.$$

What is the derivative of  $f(x(t), y(t))$  at  $t = 1$ ?

# Review

**The Chain Rule, 2 Variables (Case 1)** If

$$z = f(x, y), \quad x = g(t), \quad y = h(t),$$

then

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

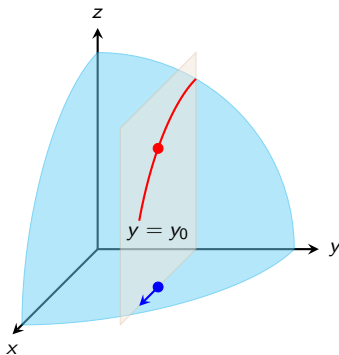
If  $x(t) = x_0 + at$ ,  $y(t) = y_0 + bt$ , what is  $(d/dt)f(x(t), y(t))$  at  $t = 0$ ?

# Review

Suppose  $f(x, y)$  is a function of two variables, and  $(x_0, y_0)$  is a point in its domain. The partial derivatives of  $f$  with respect to  $x$  and  $y$  are given by:

$$f_x(x_0, y_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + h, y_0) - f(x_0, y_0)}{h}$$

(change in direction of  $\langle 1, 0 \rangle$ )



# Review

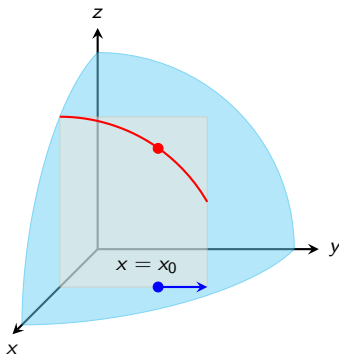
Suppose  $f(x, y)$  is a function of two variables, and  $(x_0, y_0)$  is a point in its domain. The partial derivatives of  $f$  with respect to  $x$  and  $y$  are given by:

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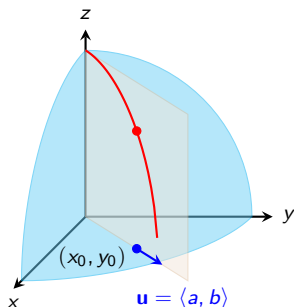
$$f_y(x_0, y_0) = \lim_{h \rightarrow 0} \frac{f(x_0, y_0 + h) - f(x_0, y_0)}{h}$$

(change in direction of  $\langle 0, 1 \rangle$ )





# The Directional Derivative



Suppose  $f(x, y)$  is a function of two variables,  $(x_0, y_0)$  is a point in its domain, and  $\mathbf{u} = \langle a, b \rangle$  is a unit vector.

The directional derivative of  $f$  in the direction  $\mathbf{u} = \langle a, b \rangle$  at  $(x_0, y_0)$  is

$$D_{\mathbf{u}}f(x_0, y_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + ah, y_0 + bh) - f(x_0, y_0)}{h}$$

# Computing the Directional Derivative

Remember that

$$D_{\mathbf{u}}f(x_0, y_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + ah, y_0 + bh) - f(x_0, y_0)}{h}$$

If  $\mathbf{u} = \langle a, b \rangle$ , this is the same as the derivative of  $f(x_0 + at, y_0 + bt)$  at  $t = 0$ .

We can compute this by the chain rule and get

$$D_{\mathbf{u}}f(x_0, y_0) = af_x(x_0, y_0) + bf_y(x_0, y_0)$$

1. Find the directional derivative of  $f(x, y) = xy^3 - x^2$  at  $(1, 2)$  in the direction  $\mathbf{u} = \langle 1/2, \sqrt{3}/2 \rangle$
2. Find the directional derivative of  $f(x, y) = x^2 \ln y$  at  $(3, 1)$  in the direction of  $\mathbf{u} = (-5/13)\mathbf{i} + (12/13)\mathbf{j}$

# The Gradient Vector

We can look at the formula

$$D_{\mathbf{u}}f(x_0, y_0) = af_x(x_0, y_0) + bf_y(x_0, y_0)$$

in a new way by introducing the *gradient of  $f$  at  $(x_0, y_0)$* : if

$$(\nabla f)(x_0, y_0) = f_x(x_0, y_0)\mathbf{i} + f_y(x_0, y_0)\mathbf{j}$$

then

$$D_{\mathbf{u}}f(x_0, y_0) = \nabla f \cdot \mathbf{u}$$

1. Find the gradient vector for  $f(x, y) = x/y$  at  $(2, 1)$
2. Find the directional derivative of  $f$  at  $(2, 1)$  in the direction  $\mathbf{u} = \frac{3}{5}\mathbf{i} + \frac{4}{5}\mathbf{j}$

# Maximum Rate of Change

Remember that

$$\mathbf{a} \cdot \mathbf{b} = |\mathbf{a}| |\mathbf{b}| \cos(\theta)$$

where  $\theta$  is the angle between  $\mathbf{a}$  and  $\mathbf{b}$ .

The dot product has its maximum value when the vector  $\mathbf{a}$  points in the same direction as  $\mathbf{b}$ .

So, the directional derivative  $D_{\mathbf{u}}f(x_0, y_0)$  has its maximum when  $\mathbf{u}$  points in the same direction as  $\nabla f(x_0, y_0)$ . In this direction,  $D_{\mathbf{u}}f(x_0, y_0) = |\nabla f(x_0, y_0)|$

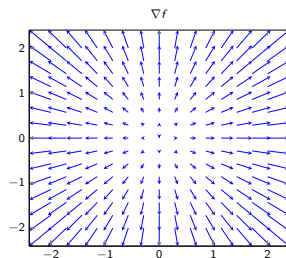
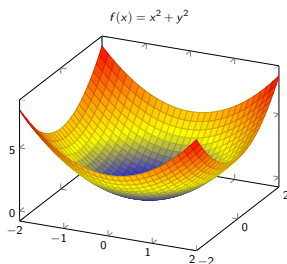
1. Find the maximum rate of change of  $f(x, y) = xe^{xy}$  at  $(0, 2)$  and find the direction where it occurs.

# Sneak Preview: The Gradient Vector Field

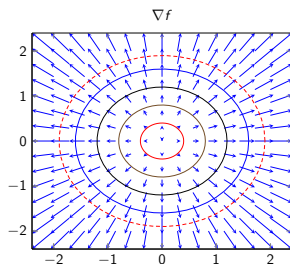
If  $f(x, y)$  is a function two variables, the *gradient vector field*

$$\nabla f(x, y) = \frac{\partial f}{\partial x}(x, y)\mathbf{i} + \frac{\partial f}{\partial y}(x, y)\mathbf{j}$$

moves in the direction of greatest change of  $f$



# The Gradient and Level Curves



The dot product of two vectors  $\mathbf{a}$  and  $\mathbf{b}$  is zero when  $\mathbf{a}$  and  $\mathbf{b}$  are perpendicular.

The directional derivative  $D_{\mathbf{u}}f$  must be zero when  $\mathbf{u}$  points along a level curve.

So, the gradient of a function  $f$  must be perpendicular to the level curves of  $f$

# Summary

- The *gradient* of a function  $f(x, y)$  at  $(x, y) = (x_0, y_0)$  is the vector

$$\nabla f(x_0, y_0) = f_x(x_0, y_0)\mathbf{i} + f_y(x_0, y_0)\mathbf{j}$$

- If  $\mathbf{u} = a\mathbf{i} + b\mathbf{j}$  is a unit vector, then the directional derivative of  $f$  at  $(x_0, y_0)$  in the direction  $\mathbf{u}$  is

$$D_{\mathbf{u}}f(x_0, y_0) = \nabla f(x_0, y_0) \cdot \mathbf{u}$$

- The gradient  $\nabla f(x_0, y_0)$  points in the direction of greatest change of  $f$  at  $(x_0, y_0)$ . The magnitude of the gradient is equal to the greatest change.
- The gradient  $\nabla f(x_0, y_0)$  is perpendicular to the level curve of  $f$  passing through  $(x_0, y_0)$

## Directional Derivatives of $f(x, y, z)$

The directional derivative of  $f(x, y, z)$  at  $(x_0, y_0, z_0)$  in the direction  $\mathbf{u} = \langle a, b, c \rangle$  is given by

$$D_{\mathbf{u}}f(x_0, y_0, z_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + ah, y_0 + bh, z_0 + ch) - f(x_0, y_0, z_0)}{h}$$

To compute it, we introduce the gradient vector

$$\nabla f(x_0, y_0, z_0) = f_x(x_0, y_0, z_0)\mathbf{i} + f_y(x_0, y_0, z_0)\mathbf{j} + f_z(x_0, y_0, z_0)\mathbf{k}$$

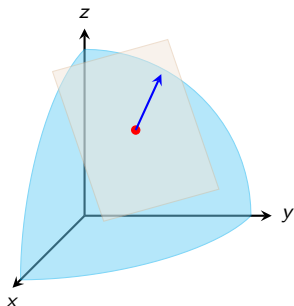
Then, if  $\mathbf{u} = \langle a, b, c \rangle$  is a unit vector,

$$D_{\mathbf{u}}f(x_0, y_0, z_0) = \nabla f(x_0, y_0, z_0) \cdot \mathbf{u}.$$

1. Using the gradient vector, find the directional derivative of  $f(x, y, z) = y^2 e^{xyz}$  at  $(0, 1, -1)$  in the direction  $\mathbf{u} = \frac{3}{13}\mathbf{i} + \frac{4}{13}\mathbf{j} + \frac{12}{13}\mathbf{k}$ .
2. Find the maximum rate of change of  $f(xy, z) = x \ln(yz)$  at  $(1, 2, 1/2)$  and find the direction in which it occurs.



# Tangent Planes to Level Surfaces



The gradient of a function of two variables is perpendicular to level curves of that function.

The gradient of a function of three variables is perpendicular to level surfaces of that function.

This means that  $\nabla f(x_0, y_0, z_0)$  is normal to the tangent plane to  $f$  at  $(x_0, y_0, z_0)$

1. Find the equations of the tangent plane and the normal line to the surface  $x = y^2 + z^2 + 1$  at  $(3, 1, -1)$ .
2. Are there any points on the hyperboloid  $x^2 - y^2 = -z^2 = 1$  where the tangent plane is parallel to the plane  $z = x + y$ ?