

# Math 11, Fall 2007

## Lecture 10

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# Outline

- 1 Review and overview
  - Last class
- 2 Today's material
  - The gradient
- 3 Group Work
- 4 Next class

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# Differentiation

## The chain rule

- $f(x, y), x = x(s), y = y(s)$

$$\frac{\partial f}{\partial s} = \frac{\partial f}{\partial x} \frac{dx}{ds} + \frac{\partial f}{\partial y} \frac{dy}{ds}$$

- $f(x, y), x = x(s, t, r), y = y(s, t, r)$

$$f_s = f_x x_s + f_y y_s$$

$$f_t = f_x x_t + f_y y_t$$

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# The gradient vector field

- We have already seen the importance of the partial derivatives,  $f_x, f_y$
- If  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ , then  $\nabla f = f_x \vec{i} + f_y \vec{j}$  gives a vector field on  $\mathbb{R}^2$  called the gradient vector field.
- If  $f : \mathbb{R}^3 \rightarrow \mathbb{R}$ , then  $\nabla f = f_x \vec{i} + f_y \vec{j} + f_z \vec{k}$

## $\nabla f$ and directional derivatives

**Theorem:** Let  $(x_0, y_0)$  be a point in the plane and  $\vec{u} = \langle a, b \rangle$  a vector. Then, if  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,

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

# Proof of Theorem

Let  $g(t) = f(x_0 + ta, y_0 + tb)$  and compute the derivative of  $g$  at zero:

$$\begin{aligned} g'(0) &= \lim_{h \rightarrow 0} \frac{g(h) - g(0)}{h} \\ &= \lim_{h \rightarrow 0} \frac{f(x_0 + ha, y_0 + hb) - f(x_0, y_0)}{h} = D_{\vec{u}} f(x_0, y_0) \end{aligned}$$

But, we may also view  $g(t)$  as  $f(x, y)$  with  $x = x_0 + ta, y = y_0 + tb$  and compute  $g'(0)$  using the chain rule:

$$\frac{dg}{dt} = f_x x_t + f_y y_t$$

Since  $x_t = a, y_t = b$ , at  $t = 0$  we have

$$g'(0) = f_x(x_0, y_0)a + f_y(x_0, y_0)b = \nabla f(x_0, y_0) \cdot \langle a, b \rangle$$

Q.E.D.



## Direction of maximal ascent

The gradient point in the direction of maximal ascent of the function. In other words, out of all the unit vectors,  $D_{\vec{u}}f$  is largest when  $\vec{u} = \frac{\nabla f}{|\nabla f|}$ .

Why is this true?

What is the direction of maximal descent? What happens at a minimum or maximum?

**Cor:**  $D_{\vec{u}}f = 0$  for all  $\vec{u}$  if and only if  $\nabla f = \vec{0}$

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# Rates of change using the gradient

Let  $f(x, y) = xy - y^2 + x^3$

- What is the direction of maximal ascent at  $(x, y) = (1, 0)$ ?
- If you are walking along the surface along the curve  $\langle x(t), y(t), f(x(t), y(t)) \rangle$  and at  $t = 1$  you are at the point  $(1, 1, 1)$  traveling in the direction  $x'(t) = -1, y'(t) = 0$ , what is your instantaneous rate of change in the  $z$  direction at that time?

# Tangent lines and planes

- Think of  $z = f(x, y)$  as a level set  
 $F(x, y, z) = f(x, y) - z = 0$ . What is  $\nabla F$ ? Can we easily write the tangent plane to the surface using  $\nabla F$ ?
- Show that  $\nabla f$  is orthogonal to the curve  $f(x, y) = 0$  (or, in three variables, the surfaces  $f(x, y, z) = 0$ ).

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# Work for next class

- Reading: 15.7
- f07hw11