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18.02 Multivariable Calculus Fall 2007

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### 18.02 Lecture 11. – Tue, Oct 2, 2007

### Differentials.

Recall in single variable calculus:  $y = f(x) \Rightarrow dy = f'(x) dx$ . Example:  $y = \sin^{-1}(x) \Rightarrow x = \sin y$ ,  $dx = \cos y \, dy$ , so  $dy/dx = 1/\cos y = 1/\sqrt{1-x^2}$ .

Total differential:  $f = f(x, y, z) \Rightarrow df = f_x dx + f_y dy + f_z dz$ .

This is a new type of object, with its own rules for manipulating it (df is not the same as  $\Delta f$ ! The textbook has it wrong.) It encodes how variations of f are related to variations of x, y, z. We can use it in two ways:

1. as a placeholder for approximation formulas:  $\Delta f \approx f_x \Delta x + f_y \Delta y + f_z \Delta z$ .

2. divide by dt to get the **chain rule**: if x = x(t), y = y(t), z = z(t), then f becomes a function of t and  $\frac{df}{dt} = f_x \frac{dx}{dt} + f_y \frac{dy}{dt} + f_z \frac{dz}{dt}$ 

Example:  $w = x^2y + z$ ,  $dw = 2xy dx + x^2 dy + dz$ . If x = t,  $y = e^t$ ,  $z = \sin t$  then the chain rule gives  $dw/dt = (2te^t) 1 + (t^2) e^t + \cos t$ , same as what we obtain by substitution into formula for w and one-variable differentiation.

Can justify the chain rule in 2 ways:

1. dx = x'(t) dt, dy = y'(t) dt, dz = z'(t) dt, so substituting we get  $dw = f_x dx + f_y dy + f_z dz = f_x x'(t) dt + f_y y'(t) dt + f_z z'(t) dt$ , hence dw/dt.

2. (more rigorous):  $\Delta w \simeq f_x \Delta x + f_y \Delta y + f_z \Delta z$ , divide both sides by  $\Delta t$  and take limit as  $\Delta t \to 0$ .

# Applications of chain rule:

Product and quotient formulas for derivatives: f = uv, u = u(t), v = v(t), then  $d(uv)/dt = f_u u' + f_v v' = vu' + uv'$ . Similarly with g = u/v,  $d(u/v)/dt = g_u u' + g_v v' = (1/v) u' + (-u/v^2) v' = (u'v - uv')/v^2$ .

Chain rule with more variables: for example w = f(x, y), x = x(u, v), y = y(u, v). Then  $dw = f_x dx + f_y dy = f_x(x_u du + x_v dv) + f_y(y_u du + y_v dv) = (f_x x_u + f_y y_u) du + (f_x x_v + f_y y_v) dv$ . Identifying coefficients of du and dv we get  $\partial f/\partial u = f_x x_u + f_y y_u$  and similarly for  $\partial f/\partial v$ . It's not legal to "simplify by  $\partial x$ ".

Example: polar coordinates:  $x = r \cos \theta$ ,  $y = r \sin \theta$ . Then  $f_r = f_x x_r + f_y y_r = \cos \theta f_x + \sin \theta f_y$ , and similarly  $f_{\theta}$ .

## 18.02 Lecture 12. – Thu, Oct 4, 2007

Handouts: PS4 solutions, PS5.

#### Gradient.

Recall chain rule:  $\frac{dw}{dt} = w_x \frac{dx}{dt} + w_y \frac{dy}{dt} + w_z \frac{dz}{dt}$ . In vector notation:  $\frac{dw}{dt} = \nabla w \cdot \frac{d\vec{r}}{dt}$ . Definition:  $\nabla w = \langle w_x, w_y, w_z \rangle$  – GRADIENT VECTOR.

Theorem:  $\nabla w$  is perpendicular to the level surfaces w = c.

Example 1: w = ax + by + cz, then w = d is a plane with normal vector  $\nabla w = \langle a, b, c \rangle$ .

Example 2:  $w = x^2 + y^2$ , then w = c are circles,  $\nabla w = \langle 2x, 2y \rangle$  points radially out so  $\perp$  circles. Example 3:  $w = x^2 - y^2$ , shown on applet (Lagrange multipliers applet with g disabled).  $\nabla w$  is a vector whose value depends on the point (x, y) where we evaluate w.

Proof: take a curve  $\vec{r} = \vec{r}(t)$  contained inside level surface w = c. Then velocity  $\vec{v} = d\vec{r}/dt$  is in the tangent plane, and by chain rule,  $dw/dt = \nabla w \cdot d\vec{r}/dt = 0$ , so  $\vec{v} \perp \nabla w$ . This is true for every  $\vec{v}$  in the tangent plane.

Application: tangent plane to a surface. Example: tangent plane to  $x^2 + y^2 - z^2 = 4$  at (2, 1, 1): gradient is  $\langle 2x, 2y, -2z \rangle = \langle 4, 2, -2 \rangle$ ; tangent plane is 4x + 2y - 2z = 8. (Here we could also solve for  $z = \sqrt{x^2 + y^2 - 4}$  and use linear approximation formula, but in general we can't.)

(Another way to get the tangent plane: dw = 2x dx + 2y dy - 2z dz = 4dx + 2dy - 2dz. So  $\Delta w \approx 4\Delta x + 2\Delta y - 2\Delta z$ . The level surface is  $\Delta w = 0$ , its tangent plane approximation is  $4\Delta x + 2\Delta y - 2\Delta z = 0$ , i.e. 4(x-2) + 2(y-1) - 2(z-1) = 0, same as above).

**Directional derivative.** Rate of change of w as we move (x, y) in an arbitrary direction.

Take a unit vector  $\hat{u} = \langle a, b \rangle$ , and look at straight line trajectory  $\vec{r}(s)$  with velocity  $\hat{u}$ , given by  $x(s) = x_0 + as$ ,  $y(s) = y_0 + bs$ . (unit speed, so s is arclength!)

Notation:  $\frac{dw}{ds}_{|\hat{u}}$ .

Geometrically: slice of graph by a vertical plane (not parallel to x or y axes anymore). Directional derivative is the slope. Shown on applet (Functions of two variables), with  $w = x^2 + y^2 + 1$ , and rotating slices through a point of the graph.

Know how to calculate dw/ds by chain rule:  $\frac{dw}{ds}_{|\hat{u}} = \nabla w \cdot \frac{d\vec{r}}{ds} = \nabla w \cdot \hat{u}.$ 

Geometric interpretation:  $dw/ds = \nabla w \cdot \hat{u} = |\nabla w| \cos \theta$ . Maximal for  $\cos \theta = 1$ , when  $\hat{u}$  is in direction of  $\nabla w$ . Hence: direction of  $\nabla w$  is that of fastest increase of w, and  $|\nabla w|$  is the directional derivative in that direction. We have dw/ds = 0 when  $\hat{u} \perp \nabla w$ , i.e. when  $\hat{u}$  is tangent to direction of level surface.

### 18.02 Lecture 13. – Fri, Oct 5, 2007 (estimated – written before lecture)

Practice exams 2A and 2B are on course web page.

#### Lagrange multipliers.

Problem: min/max when variables are constrained by an equation g(x, y, z) = c.

Example: find point of xy = 3 closest to origin ? I.e. minimize  $\sqrt{x^2 + y^2}$ , or better  $f(x, y) = x^2 + y^2$ , subject to g(x, y) = xy = 3. Illustrated using Lagrange multipliers applet.

Observe on picture: at the minimum, the level curves are tangent to each other, so the normal vectors  $\nabla f$  and  $\nabla g$  are parallel.

So: there exists  $\lambda$  ("multiplier") such that  $\nabla f = \lambda \nabla g$ . We replace the constrained min/max problem in 2 variables with equations involving 3 variables  $x, y, \lambda$ :

$\int f_x$	$=\lambda g_x$			$2x = \lambda y$
$\left\{ f_{y} \right\}$	$=\lambda g_y$	i.e. here	{	$2y = \lambda x$
g =	= <i>c</i>			xy = 3.

In general solving may be hard and require a computer. Here, linear algebra:  $\begin{cases} 2x - \lambda y = 0 \\ -\lambda x + 2y = 0 \end{cases}$ 

requires either x = y = 0 (impossible, since xy = 3), or det  $= 4 - \lambda^2 = 0$ . So  $\lambda = \pm 2$ . No solutions for  $\lambda = -2$ , while  $\lambda = 2$  gives  $(\sqrt{3}, \sqrt{3})$  and  $(-\sqrt{3}, -\sqrt{3})$ . (Checked on applet that  $\nabla f = 2\nabla g$  at minimum).

Why the method works: at constrained min/max, moving in any direction along the constraint surface g = c should give df/ds = 0. So, for any  $\hat{u}$  tangent to  $\{g = c\}, \frac{df}{ds}|_{\hat{u}} = \nabla f \cdot \hat{u} = 0$ , i.e.  $\hat{u} \perp \nabla f$ . Therefore  $\nabla f$  is normal to tangent plane to g = c, and so is  $\nabla g$ , hence the gradient vectors are parallel.

Warning: method doesn't say whether we have a min or a max, and second derivative test doesn't apply with constrained variables. Need to answer using geometric argument or by comparing values of f.

Advanced example: surface-minimizing pyramid.

Triangular-based pyramid with given triangle as base and given volume V, using as little surface area as possible.

Note:  $V = \frac{1}{3}A_{base}h$ , so height h is fixed, top vertex moves in a plane z = h.

We can set up problem in coordinates: base vertices  $P_1 = (x_1, y_1, 0)$ ,  $P_2$ ,  $P_3$ , and top vertex P = (x, y, h). Then areas of faces  $= \frac{1}{2} |\vec{PP_1} \times \vec{PP_2}|$ , etc. Calculations to find critical point of function of (x, y) are very hard.

Key idea: use variables adapted to the geometry, instead of (x, y): let  $a_1, a_2, a_3$  = lengths of sides of the base triangle;  $u_1, u_2, u_3$  = distances in the xy-plane from the projection of P to the sides of the base triangle. Then each face is a triangle with base length  $a_i$  and height  $\sqrt{u_i^2 + h^2}$  (using Pythagorean theorem).

So we must minimize  $f(u_1, u_2, u_3) = \frac{1}{2}a_1\sqrt{u_1^2 + h^2} + \frac{1}{2}a_2\sqrt{u_2^2 + h^2} + \frac{1}{2}a_3\sqrt{u_3^2 + h^2}$ .

Constraint? (asked using flashcards; this was a bad choice, very few students responded at all.) Decomposing base into 3 smaller triangles with heights  $u_i$ , we must have  $g(u_1, u_2, u_3) = \frac{1}{2}a_1u_1 + \frac{1}{2}a_2u_2 + \frac{1}{2}a_3u_3 = A_{base}$ .

Lagrange multiplier method:  $\nabla f = \lambda \nabla g$  gives

$$\frac{a_1}{2}\frac{u_1}{\sqrt{u_1^2+h^2}} = \lambda \frac{a_1}{2}, \quad \text{similarly for } u_2 \text{ and } u_3.$$

We conclude  $\lambda = \frac{u_1}{\sqrt{u_1^2 + h^2}} = \frac{u_2}{\sqrt{u_2^2 + h^2}} = \frac{u_3}{\sqrt{u_3^2 + h^2}}$ , hence  $u_1 = u_2 = u_3$ , so *P* lies above the incenter.