MATH 20550 Differentiable, Directional Derivatives and Gradients Fall 2016

1. Differentiable Functions

A real-valued function of a vector variable, $F(\mathbf{x})$ is differentiable at a point **a** if and only if

$$\lim_{\mathbf{h}\to\mathbf{0}}\frac{f(\mathbf{a}+\mathbf{h})-f(\mathbf{a})-\nabla f(\mathbf{a})\cdot\mathbf{h}}{|\mathbf{h}|}=0$$

As we say in the section on limits, if the limit exists, then its value can be calculated. To do this, fix a vector $\mathbf{h} \neq \mathbf{0}$ and let $t\mathbf{h}$ be a line to $\mathbf{0}$. Since there is an $|\mathbf{h}|$ in our formula it is useful to compute $\lim_{t \to 0^+}$.

If f is differentiable, the limit is 0 along every line so

$$0 = \lim_{t \to 0^+} \frac{f(\mathbf{a} + t\mathbf{h}) - f(\mathbf{a}) - \nabla f(\mathbf{a}) \bullet (t\mathbf{h})}{t |\mathbf{h}|}$$

or

$$\lim_{t \to 0^+} \frac{f(\mathbf{a} + t\mathbf{h}) - f(\mathbf{a})}{t |\mathbf{h}|} = \nabla f(\mathbf{a}) \bullet \frac{\mathbf{h}}{|\mathbf{h}|}$$

Similalrly

$$\lim_{t \to 0^{-}} \frac{f(\mathbf{a} + t\mathbf{h}) - f(\mathbf{a})}{-t |\mathbf{h}|} = -\nabla f(\mathbf{a}) \bullet \frac{\mathbf{h}}{|\mathbf{h}|}$$

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(*)
$$\lim_{t \to 0} \frac{f(\mathbf{a} + t\mathbf{h}) - f(\mathbf{a})}{t |\mathbf{h}|} = \nabla f(\mathbf{a}) \cdot \frac{\mathbf{h}}{|\mathbf{h}|}$$

If we let $\mathbf{h} = \mathbf{e}_i$, (*) shows that

$$\lim_{t \to 0} \frac{f(\mathbf{a} + t\mathbf{e}_i) - f(\mathbf{a})}{t} = \frac{\partial f}{\partial x_i}(\mathbf{a})$$

and therefore, if f is differentiable at \mathbf{a} , all the partials exist at \mathbf{a} .

The existence of partial derivatives is insufficient to make the function differentiable, just as the existence of limits along some lines in insufficient to guarantee the limit exists.

Theorem. If a function has all partial derivatives continuous in a neighborhood of \mathbf{a} then f is differentiable in that neighborhood.

As a practical matter, write down the partials and argue that they are continuous using your theorems on when a function is continuous.

1.1. Directional Derivatives. By definition, if \mathbf{u} is a unit vector, the *directional derivative* of f in the direction \mathbf{u} at the point \mathbf{a} is

$$D_{\mathbf{u}}(f)(\mathbf{a}) = \lim_{t \to 0} \frac{f(\mathbf{a} + t\mathbf{u}) - f(\mathbf{a})}{t}$$

It follows that if f is differentiable at \mathbf{a} ,

$$D_{\mathbf{u}}(f)(\mathbf{a}) = \nabla f(\mathbf{a}) \bullet \mathbf{u}$$

1.2. Properties of the Gradient.

- $D_{\mathbf{u}}(f)(\mathbf{a})$ is the instantaneous rate of change of f at \mathbf{a} in the direction \mathbf{u} . If $D_{\mathbf{u}}(f)(\mathbf{a})$ is positive, the value of f is increasing; if $D_{\mathbf{u}}(f)(\mathbf{a})$ is negative, the value of f is decreasing.
- The direction of fastest increase in f at a point \mathbf{a} is the direction of the gradient. (Or meaningless if $\nabla f(\mathbf{a}) = \mathbf{0}$.) The value of this rate is $|\nabla f(\mathbf{a})|$.
- The direction of fastest decrease in f at a point **a** is the negative of the direction of the gradient. (Or meaningless if $\nabla f(\mathbf{a}) = \mathbf{0}$.) The value of this rate is $-|\nabla f(\mathbf{a})|$.
- In two variables, if $\mathbf{r}(t)$ is a smooth parametrization of a level curve of f(x, y) then $\mathbf{r}'(t_0)$ is perpendicular to the gradient $\nabla f(\mathbf{r}(t_0))$.
- In three variables, if $\mathbf{r}(t)$ is a smooth parametrization of a curve lying in the surface f(x, y, z) = c then $\mathbf{r}'(t_0)$ is perpendicular to the gradient $\nabla f(\mathbf{r}(t_0))$.

2. More on level curves

Suppose f(x, y) = c is a level curve. Then at any point **a** on the level curve, $\nabla f(\mathbf{a})$ is orthogonal to the level curve. Hence as long as $\nabla f(\mathbf{a}) \neq \mathbf{0}$, $\mathbf{N} = \frac{\nabla f(\mathbf{a})}{|\nabla f(\mathbf{a})|}$ is a unit normal to the curve and this normal is the direction of fastest increase in the value of f.

It is also possible to work out the remaining invariants of a plane curve for this sort of example. In particular, $\mathbf{B} = \langle 0, 0, 1 \rangle$ and so a vector in the direction of \mathbf{T} is $\mathbf{N} \times \mathbf{B} = |\mathbf{i} \quad \mathbf{i} \quad \mathbf{k}|$

$$\begin{vmatrix} \partial f & \partial f \\ \partial x & \partial f & 0 \\ 0 & 0 & 1 \end{vmatrix} = \left\langle \frac{\partial f}{\partial y}, -\frac{\partial f}{\partial x}, 0 \right\rangle.$$

The remaining invariant is the curvature. Suppose $\frac{\partial f}{\partial x}\Big|_{\mathbf{x}=\mathbf{a}=\langle x_0,y_0\rangle} \neq 0$. Then the Implicit Function Theorem says that near this point on the level curve, there exists a differentiable function $\alpha(x)$ such that $\mathbf{r} = \langle x, \alpha(x) \rangle$ is the level curve. The curvature is $|\alpha''(x_0)|$

$$\frac{\left|\alpha\left(x_{0}\right)\right|}{\left(\sqrt{1+\left(\alpha'(x_{0})\right)^{2}}\right)^{3}}.$$

By the Chain Rule, $\nabla f(\langle x, \alpha(x) \rangle) \bullet \langle 1, \alpha'(x) \rangle = 0$ so

$$\frac{\partial f}{\partial x}\left(\langle x, \alpha(x) \rangle\right) + \frac{\partial f}{\partial y}\left(\langle x, \alpha(x) \rangle\right) \alpha'(x) = 0$$

$$(1) \quad \frac{\partial^2 f}{\partial x \partial x} + \frac{\partial^2 f}{\partial y \partial x} \alpha'(x) + \left(\frac{\partial^2 f}{\partial y \partial x} + \frac{\partial^2 f}{\partial y \partial y} \alpha'(x)\right) \alpha'(x) + \frac{\partial f}{\partial y} \left(\langle x, \alpha(x) \rangle\right) \alpha''(x) = 0$$

Multiply (1) by $\left(\frac{\partial f}{\partial y}\right)^2$ and assume mixed partials are equal:

(2)
$$\frac{\partial^2 f}{\partial x \partial x} \left(\frac{\partial f}{\partial y}\right)^2 - 2 \frac{\partial^2 f}{\partial y \partial x} \frac{\partial f}{\partial x} \frac{\partial f}{\partial y} + \frac{\partial^2 f}{\partial y \partial y} \left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^3 \alpha''(x) = 0$$

and so

$$\kappa(x_0, y_0) = \frac{\left|\frac{\partial^2 f}{\partial x \partial x} \left(\frac{\partial f}{\partial y}\right)^2 - 2\frac{\partial^2 f}{\partial y \partial x} \frac{\partial f}{\partial x} \frac{\partial f}{\partial y} + \frac{\partial^2 f}{\partial y \partial y} \left(\frac{\partial f}{\partial x}\right)^2}{\left(\sqrt{\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2}\right)^3}$$

Consider this example of the level curves of $f(x, y) = x^2 - y^2$ from worksheet 4.



A second example: the graph and level curves of $f(x, y) = x^3 - 2xy + y^2$



3. The curve of intersection of two implicit surfaces

An implicit surface is the set of solutions to an equation f(x, y, z) = c. Given two implicit surfaces f(x, y, z) = c and g(x, y, z) = d they typically intersect in a curve. If **a** is a point on both surfaces we can try to compute the typical invariants of the intersect curve at this point.

If we parametrize the curve by $\mathbf{r}(t)$ then $f(\mathbf{r}(t)) = c$ and $\nabla f(\mathbf{r}(t)) \cdot \mathbf{r}'(t) = 0$. In particular $\mathbf{r}'(t)$ is perpendicular to $\nabla f(\mathbf{r}(t))$. Similarly $\mathbf{r}'(t)$ is perpendicular to $\nabla g(\mathbf{r}(t))$. Hence we will take $\mathbf{r}'(t)$ parallel to $\nabla f(\mathbf{r}(t)) \times \nabla g(\mathbf{r}(t))$ and so

$$\mathbf{T}(\mathbf{a}) = \frac{\nabla f(\mathbf{a}) \times \nabla g(\mathbf{a})}{\left|\nabla f(\mathbf{a}) \times \nabla g(\mathbf{a})\right|}$$

If we assume $\nabla f(\mathbf{a}) \times \nabla g(\mathbf{a}) \neq \mathbf{0}$ it follows that $\nabla f(\mathbf{a}) \neq \mathbf{0}$ and $\nabla g(\mathbf{a}) \neq \mathbf{0}$. The Implicit Function Theorem says that around \mathbf{a} , f(x, y, z) = c is a graph $z = \alpha(x, y)$ and g(x, y, z) = d is a graph $z = \beta(x, y)$. The intersection of the two surfaces is the set of all points where $\alpha(x, y) = \beta(x, y)$ and (under some non-zero hypothesis of partials) there is a function $\gamma(x)$ so that $\alpha(x, \gamma(x)) = \beta(x, \gamma(x))$. The curve is paramterized by

$$\mathbf{r}(x) = \left\langle x, \gamma(x), \alpha(x, \gamma(x)) \right\rangle = \left\langle x, \gamma(x), \beta(x, \gamma(x)) \right\rangle$$

$$\begin{split} f(\langle x, y, \alpha(x, y) \rangle) &= c \\ \nabla f(\langle x, y, \alpha(x, y) \rangle) \bullet \left\langle 1, 0, \frac{\partial \alpha}{\partial x} \right\rangle &= 0 \\ \nabla f(\langle x, y, \alpha(x, y) \rangle) \bullet \left\langle 0, 1, \frac{\partial \alpha}{\partial y} \right\rangle &= 0 \text{ so} \\ \left\langle 1, 0, \frac{\partial \alpha}{\partial x} \right\rangle \times \left\langle 0, 1, \frac{\partial \alpha}{\partial y} \right\rangle &= \begin{vmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ 1 & 0 & \frac{\partial \alpha}{\partial x} \\ 0 & 1 & \frac{\partial \alpha}{\partial y} \end{vmatrix} = \begin{vmatrix} 0 & \frac{\partial \alpha}{\partial x} \\ 1 & \frac{\partial \alpha}{\partial y} \end{vmatrix} \mathbf{i} - \begin{vmatrix} 1 & \frac{\partial \alpha}{\partial x} \\ 0 & \frac{\partial \alpha}{\partial y} \end{vmatrix} \mathbf{j} + \begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} \mathbf{k} \\ &= \left\langle -\frac{\partial \alpha}{\partial x}, -\frac{\partial \alpha}{\partial y}, 1 \right\rangle \end{split}$$

is parallel to ∇f so

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$$\nabla f = \left\langle -\frac{\partial \alpha}{\partial x} \cdot \frac{\partial f}{\partial z}, -\frac{\partial \alpha}{\partial y} \cdot \frac{\partial f}{\partial z}, \frac{\partial f}{\partial z} \right\rangle \quad \text{so} \quad \frac{\partial \alpha}{\partial x} = -\frac{\frac{\partial f}{\partial x}}{\frac{\partial f}{\partial z}} \quad \text{and} \quad \frac{\partial \alpha}{\partial y} = -\frac{\frac{\partial f}{\partial y}}{\frac{\partial f}{\partial z}}$$
$$\nabla g = \left\langle -\frac{\partial \beta}{\partial x} \cdot \frac{\partial g}{\partial z}, -\frac{\partial \beta}{\partial y} \cdot \frac{\partial g}{\partial z}, \frac{\partial g}{\partial z} \right\rangle \quad \text{so} \quad \frac{\partial \beta}{\partial x} = -\frac{\frac{\partial g}{\partial x}}{\frac{\partial g}{\partial y}} \quad \text{and} \quad \frac{\partial \beta}{\partial y} = -\frac{\frac{\partial g}{\partial y}}{\frac{\partial g}{\partial y}}$$

a f

$$\overline{\partial z} \qquad \overline{\partial z}$$

Since $\alpha(x, \gamma(x)) = \beta(x, \gamma(x))$, differentiating with respect to x gives $\nabla \alpha \cdot \langle 1, \gamma' \rangle = \nabla \beta \cdot \langle 1, \gamma' \rangle$

$$\gamma'(x) = \frac{\frac{\partial \alpha}{\partial x} - \frac{\partial \beta}{\partial x}}{\frac{\partial \beta}{\partial y} - \frac{\partial \alpha}{\partial y}} = \frac{\frac{\partial g}{\partial x} - \frac{\partial f}{\partial x}}{\frac{\partial f}{\partial y} - \frac{\partial g}{\partial y}}$$

With these formulas and the Chain Rule one can calculate \mathbf{r}' , \mathbf{r}'' and \mathbf{r}''' in terms of the (higher order) partials of f and g and hence obtain the curvature, the torsion and the Frenet-Serret frame at a point on the intersection curve.

4. Approximations (Extra)

Recall that a function is *differentiable* if and only if

$$\lim_{\mathbf{h}\to\mathbf{0}}\frac{\left|\left(f(\mathbf{a}+\mathbf{h})-f(\mathbf{a})\right)-\nabla f(\mathbf{a})\cdot\mathbf{h}\right|}{|\mathbf{h}|}=0$$

This in turn means that if $|\mathbf{h}|$ is small,

(*)
$$(f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a})) - \nabla f(\mathbf{a}) \cdot \mathbf{h} \approx 0$$

4.1. Approximation values. Suppose you have a function $f(\mathbf{x})$ and you know $f(\mathbf{a})$ but you want to know $f(\mathbf{b})$ for some other input **b**. Write

$$f(\mathbf{b}) = f(\mathbf{a} + (\mathbf{b} - \mathbf{a}))$$

Then, letting $\mathbf{h} = \mathbf{b} - \mathbf{a}$ and rewriting (*)

$$f(\mathbf{b}) = f(\mathbf{a} + (\mathbf{b} - \mathbf{a})) \approx f(\mathbf{a}) + \nabla f(\mathbf{a}) \cdot (\mathbf{b} - \mathbf{a})$$

4.2. Solve equations. Try to solve $f(\mathbf{x}) = 0$. Pick a point \mathbf{a}_0 and evaluate $f(\mathbf{a}_0)$. If this is not 0 solve

$$f(\mathbf{a}_0) + t_0 \nabla f(\mathbf{a}_0) \bullet \nabla f(\mathbf{a}_0) = 0$$

for t_0 . As long as $\nabla f(\mathbf{a}_0) \neq \mathbf{0}$ you can do this. Since $f(\mathbf{a}_0) + t_0 \nabla f(\mathbf{a}_0) \bullet \nabla f(\mathbf{a}_0) = 0$, if we take

$$\mathbf{a}_1 = \mathbf{a}_0 + t_0 \nabla f(\mathbf{a}_0)$$

and approximate $f(\mathbf{a}_1)$ we get

$$f(\mathbf{a}_1) \approx f(\mathbf{a}_0) + t_0 \nabla f(\mathbf{a}_0) \bullet \nabla f(\mathbf{a}_0) = 0$$

So we might expect that \mathbf{a}_1 is a better approximation to the solution than was our initial guess \mathbf{a}_1 . Now iterate.

In summary:

Pick \mathbf{a}_0 , compute

$$\mathbf{a}_1 = \mathbf{a}_0 - \frac{f(\mathbf{a}_0)}{\nabla f(\mathbf{a}_0) \bullet \nabla f(\mathbf{a}_0)} \nabla f(\mathbf{a}_0)$$

and iterate.

Example: Find a solution to f(x, y, z) = 0 where

$$f(x, y, z) = x^2 + xyz - xz^3 - 4$$

Compute $\nabla f = \langle 2x + yz - z^3, xz, xy - 3xz^2 \rangle$.

	х	у	z	f(x,y,z)	$f_x(x,y,z)$	$f_y(x,y,z)$	f_z(x,y,z)	f/Grad•Grad
a_0	1	2	3	-24	-19	3	-25	-0.024120603
a_1	0.5417085427	2.072361809	2.3969849246	-8.4760490463	-7.7211278229	1.2984672104	-8.21460289	-0.0658172711
a_2	0.0335249794	2.1578233775	1.8563221791	-4.0790391515	-2.3240947064	0.0622331629	-0.2742334171	-0.7442832331
a_3	-1.6962597426	2.2041424771	1.6522148448	0.3505362172	-4.2610416023	-2.8025855273	10.1526220846	0.00271551
a_4	-1.6846888414	2.2117529262	1.6246452978	0.008853864	-4.0642700231	-2.7370218045	9.6139533387	7.60395100E-005
a_5	-1.6843797963	2.211961048	1.6239142575	6.13224384871103E-006	-4.0591447638	-2.7352883663	9.5998388293	5.28116489E-008
a_6	-1.684379582	2.2119611925	1.6239137505	2.94875235340442E-012	-4.059141211	-2.7352871642	9.5998290437	2.53950719E-014
a_7	-1.684379582	2.2119611925	1.6239137505	0	-4.059141211	-2.7352871642	9.5998290437	0



4.3. Points of intersection of two implicit surfaces. Suppose you want a point on f(x, y, z) = c and g(x, y, z) = d. Solve

$$(f(x, y, z) - c)^{2} + (g(x, y, z) - d)^{2} = 0$$