MATH358 - Advanced Calculus

Based on lectures from Winter 2026 by Prof. John Toth.
Notes by Louis Meunier

Contents

§1 Differentiation

We say $\Omega \subset \mathbb{R}^n$ a *domain* if it is open and connected.

↪**Definition 1.1** (Differentiation): Let $f = (f_1, \ldots, f_m)^T : \Omega \to \mathbb{R}^m$, Ω a domain in $\mathbb{R}^n$ and $f_j : \Omega \to \mathbb{R}$. We say f *differentiable* at $x_0 \in \Omega$ if there exists a linear map $L : \mathbb{R}^n \to \mathbb{R}^m$ such that

$$\lim_{x \to x_0} \frac{\|f(x) - f(x_0) - L(x - x_0)\|}{\|x - x_0\|} = 0.$$

Remark 1.1: Note that the first norm on $\mathbb{R}^m$, the second on $\mathbb{R}^n$.

Remark 1.2: In terms of ε, δ, the definition says that $\forall\, \varepsilon > 0$, $\exists\, \delta > 0$ such that if $x \in \Omega \cap B(x_0, \delta)$, then $\|f(x) - f(x_0) - L(x - x_0)\| < \varepsilon \|x - x_0\|$.

↪**Theorem 1.1**: L as above is unique if it exists.

Proof. Suppose $L_1 \neq L_2$ both satisfy the definition. Then, for all $\varepsilon > 0$, there exists $\delta > 0$ such that if $0 < \|x - x_0\| < \delta$, then

$$\begin{aligned} \|(L_1 - L_2)(x - x_0)\| &\leq \|f(x) - f(x_0) - L_1(x - x_0)\| + \|f(x) - f(x_0) - L_2(x - x_0)\| \\ &\leq \varepsilon \|x - x_0\|, \end{aligned}$$

by differentiability (and the previous remark). In particular, $\|(L_1 - L_2)u\| < \varepsilon$ for all unit vectors u, which implies $\|(L_1 - L_2)u\| = 0$ and thus $L_1 = L_2$. ■

↪**Definition 1.2**: If f differentiable at x_0, we'll write $Df(x_0) = L$ for the *differential* of f at x_0.

↪**Proposition 1.1**: f differentiable at x_0 implies f continuous at x_0. In fact, f is Lipschitz at x_0.

Proof. Let $\delta > 0$ such that $\|x - x_0\| < \delta$ implies $\|f(x) - f(x_0) - Df(x_0)(x - x_0)\| < \|x - x_0\|$, which implies

$$\|f(x) - f(x_0)\| \leq \|Df(x_0)(x - x_0)\| + \|x - x_0\| \leq (\|L\| + 1)\|x - x_0\|,$$

which readily proves the statement. ■

↪**Proposition 1.2**: f differentiable at a point x_0 iff each of its component functions are differentiable at x_0.

↪**Definition 1.3**: For $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$, define the *partial derivative*

$$\frac{\partial f_j}{\partial x_i}(x_1, \ldots, x_m) := \lim_{h \to 0} \frac{\left[f_j(x_1, \ldots, x_i + h, \ldots, x_m) - f_j(x_1, \ldots, x_i, \ldots, x_m)\right]}{h},$$

if the limit exists.

↪**Proposition 1.3**: Let $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$ be differentiable at x_0. Then, $\frac{\partial f_j}{\partial x_i}(x_0)$ exists for each $i = 1, ..., n$ and $j = 1, ..., m$, and

$$L = Df(x_0) = \left(\frac{\partial f_j}{\partial x_i}(x_0)\right)_{\substack{1 \leq i \leq n \\ 1 \leq j \leq m}}.$$

We call this matrix the Jacobian or derivative of f at x_0.

Proof. Write $L = \left(a_{ji}\right)$ in the standard basis $e_1, ..., e_n$ for $\mathbb{R}^n$. Let $\varepsilon > 0$, fix some i with $1 \leq i \leq n$, and set $x := x_0 + he_i$, with $|h| < \delta$ sufficiently small. By differentiability,

$$\frac{\|f(x) - f(x_0) - L(x - x_0)\|}{\|x - x_0\|} = \left(\sum_{j=1}^{m}\left[\frac{f_j(x) - f_j(x_0)}{h} - a_{ji}\right]^2\right)^{1/2}.$$

Since the limit as $h \to 0$ of the above ratio must be zero, the limit of each term in the summation as $h \to 0$ must be zero as well (being a sum of nonnegative terms), i.e.

$$\lim_{h \to 0} \frac{f_j(x) - f_j(x_0)}{h} = a_{ji} \quad \forall j = 1, ..., m.$$

But the limit on the left is just $\frac{\partial f_j}{\partial x_i}(x_0)$, which proves all of the claims in turn. ■

Remark 1.3: This proposition says that f differentiable at x_0 implies $\frac{\partial f_j}{\partial x_i}(x_0)$ exists for all i, j. The converse need not be true. Consider

$$f(x,y) := \begin{cases} \frac{xy}{\sqrt{x^2+y^2}} & (x,y) \neq 0 \\ 0 & \text{else} \end{cases}.$$

⊛ **Example 1.1**: Another counterexample as in the previous remark is the function

$$f(x,y) := \begin{cases} \frac{x^2 y}{x^2+y^2} & (x,y) \neq (0,0) \\ 0 & (x,y) = (0,0) \end{cases}.$$

Claim 1: f continuous at $(0,0)$. We have, for $(x,y) \neq (0,0)$,

$$\begin{aligned} |f(x,y) - f(0,0)| &= \left|\frac{x^2 y}{x^2+y^2}\right| \\ &= \frac{x^2|y|}{x^2+y^2} \\ &\leq |y| \to 0 \text{ as } (x,y) \to (0,0), \end{aligned}$$

so we have continuity indeed.

Claim 2: $\partial_x f, \partial_y f$ exist at the origin, and are equal to zero. Note that $f(x,0) = 0$ for $x \neq 0$, and $f(0,0) = 0$, so it follows that $\partial_x f(0,0) = 0$. Similarly for $\partial_y f(0,0)$.

Claim 3: f is not differentiable at $(0,0)$. Suppose otherwise. Then, $L = Df(0,0) = (0,0)$, so

$$\begin{aligned} 0 &= \lim_{(x,y)\to(0,0)} \frac{|f(x,y) - f(0,0) - Df(0,0)(x,y)|}{\|(x,y)\|} \\ &= \lim_{(x,y)\to(0,0)} \frac{|f(x,y)|}{\|(x,y)\|} \\ &= \lim_{(x,y)\to(0,0)} \frac{x^2|y|}{(x^2+y^2)\cdot\sqrt{x^2+y^2}} \\ &= \lim_{(x,y)\to(0,0)} \frac{x^2|y|}{(x^2+y^2)^{3/2}}. \end{aligned}$$

Suppose $y = x$ in the final term (i.e., we approach the limit on a diagonal), and $x > 0$, then this ratio simplifies

$$\frac{x^3}{(2x^2)^{3/2}} = \frac{1}{2^{3/2}} \neq 0,$$

so we have a contradiction.

We can get a partial converse, however, if we assume continuity.

↪**Theorem 1.2**: Let $f = (f_1, ..., f_m) : \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$. Suppose each $\frac{\partial f_j}{\partial x_i}$ is continuous at some $x^0 \in \Omega$. Then, f is differentiable at x^0.

Proof. We use MVT, and suppose $n = 2, m = 1$ for simplicity of notation, so that $f : \Omega \subset \mathbb{R}^2 \to \mathbb{R}$. We write $x = (x_1, x_2) \in \Omega, x^0 = (x_1^0, x_2^0)$. Let $\varepsilon > 0$. By assumption, there exists a $\delta > 0$ such that

$$\|y - x^0\| < \delta \Rightarrow \left|\frac{\partial f}{\partial x_i}(y) - \frac{\partial f}{\partial x_i}(x^0)\right| \leq \frac{\varepsilon}{2}, \quad i = 1, 2.$$

We write

$$\begin{aligned} f(x) - f(x^0) &= f(x_1, x_2) - f\left(x_1^0, x_2\right) + f\left(x_1^0, x_2\right) - f\left(x_1^0, x_2^0\right) \\ \text{(MVT, coordinate-wise)} \quad &= \frac{\partial f}{\partial x_1}(z_1, x_2)\left(x_1 - x_1^0\right) + \frac{\partial f}{\partial x_2}\left(x_1^0, z_2\right)\left(x_2 - x_2^0\right), \end{aligned}$$

for some z_1 between x_1 and x_1^0 and some z_2 between x_2 and x_2^0. Thus,

$$\begin{aligned} f(x) - f(x^0) - Df(x^0)(x - x^0) &= f(x) - f(x^0) - \left(\partial_{x_1} f(x^0), \partial_{x_2} f(x^0)\right) \cdot (x - x^0) \\ &= \frac{\partial f}{\partial x_1}(z_1, x_2)\left(x_1 - x_1^0\right) + \frac{\partial f}{\partial x_2}\left(x_1^0, z_2\right)\left(x_2 - x_2^0\right) \\ &\quad - \frac{\partial f}{\partial x_1}(x^0)\left(x_1 - x_1^0\right) - \frac{\partial f}{\partial x_2}(x^0)\left(x_2 - x_2^0\right) \\ &= \left[\partial_{x_1} f(z_1, x_2) - \partial_{x_1} f\left(x_1^0, x_2^0\right)\right]\left(x_1 - x_1^0\right) \\ &\quad + \left[\partial_{x_2}\left(x_1^0, z_2\right) - \partial_{x_2} f\left(x_1^0, x_2^0\right)\right]\left(x_2 - x_2^0\right). \end{aligned}$$

By choice of z_1, z_2 and for (x_1, x_2) in $B(x^0, \delta)$, we know $(z_1, x_2) \in B(x^0, \delta)$ and $\left(x_1^0, z_2\right) \in B(x^0, \delta)$ as well, so we can appeal to continuity. In addition, its clear that $\left|x_i - x_i^0\right| \leq \|x - x^0\|$. Thus, using continuity, we find

$$|f(x) - f(x^0) - Df(x^0)(x - x^0)| \leq \left(\frac{\varepsilon}{2} + \frac{\varepsilon}{2}\right)\|x - x^0\| = \varepsilon\|x - x^0\|,$$

so dividing both sides by $\|x - x^0\|$ immediately gives the result. ■

↪**Definition 1.4**: Suppose $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}$ has continuous $\frac{\partial f}{\partial x_i}$ at all points in Ω. Then, we say f is *continuously differentiable* (in Ω), and we write $f \in C^1(\Omega)$.

Remark 1.4: Continuity of partial derivatives is sufficient, but not necessary, for differentiability. For instance,

$$f(x, y) = \begin{cases} \frac{x^2 y^2}{x^2 + y^4} & (x, y) \neq (0, 0) \\ 0 & (x, y) = (0, 0) \end{cases}.$$

On readily computes $\partial_x f(0, 0) = \partial_y f(0, 0) = 0$, but along the parabola $x = t^2, y = t$ $(t \neq 0)$,

$$\partial_x f(t^2, t) = \frac{1}{2},$$

so $\partial_x f$ can't be continuous. However, f is still differentiable at $(0, 0)$: we claim $L = 0$, then

$$\frac{|f(x, y) - f(0, 0) - L(x, y)|}{\|(x, y)\|} = \frac{|f(x, y)|}{(x^2 + y^2)^{\frac{1}{2}}} = \frac{x^2 y^2}{(x^2 + y^4)(x^2 + y^2)^{\frac{1}{2}}} \leq \frac{y^2}{|x^2 + y^2|^{\frac{1}{2}}} \leq |y| \underset{(x,y)\to 0}{\to} 0.$$

↪**Proposition 1.4** (Basic Properties of Differentiation):

1. If $f, g : \Omega \to \mathbb{R}^m$ both differentiable at $x^0 \in \Omega$, then so is $F = f + g$, and
$$D(f+g)(x^0) = Df(x^0) + Dg(x^0).$$
2. If $f, g : \Omega \to \mathbb{R}^m$ both differentiable at $x^0 \in \Omega$, then so is $F = fg : \Omega \to \mathbb{R}$, and
$$DF(x^0) = f(x^0)Dg(x^0) + g(x^0)Df(x^0).$$
3. $f, g : \Omega \to \mathbb{R}$ both differentiable at x^0 with $g(x^0) \neq 0$, then so is $F = \frac{f}{g}$, and
$$DF(x^0) = \frac{DF(x^0)}{g(x^0)} - \frac{f(x^0)Dg(x^0)}{g^2(x^0)}.$$
4. (Chain Rule) Given $f : \Omega \subset \mathbb{R}^n \to \tilde{\Omega} \subset \mathbb{R}^m$ and $g : \tilde{\Omega} \to \mathbb{R}^k$, with f differentiable at x^0 and g differentiable at $y^0 = f(x^0)$, then $H = g \circ f : \Omega \subset \mathbb{R}^n \to \mathbb{R}^k$ is differentiable at x^0, and
$$DH(x^0) = Dg(y^0) \cdot Df(x^0),$$
in which one should read the "$\cdot$" as matrix multiplication.

Proof. 1., 2., 3. left as an exercise. We prove 4., the Chain Rule, for it is realistically the most interesting. Set $L := Dg(y_0) \cdot Df(x_0)$, and we'll write $y = f(x)$ (so in particular $y_0 = f(x_0)$, as in the statement). We need to show
$$\lim_{x \to x_0} \frac{\|H(x) - H(x_0) - L(x - x_0)\|}{\|x - x_0\|} = 0.$$
Let us work the numerator:
$$\begin{aligned} H(x) - H(x_0) - L(x - x_0) &= g(y) - g(y_0) - Dg(y_0)Df(x_0)(x - x_0) \\ &= g(y) - g(y_0) - Dg(y_0)(y - y_0) \\ &\quad + Dg(y_0)(y - y_0) - Dg(y_0)Df(x_0)(x - x_0) \\ &= g(y) - g(y_0) - Dg(y_0)(y - y_0) \\ &\quad + Dg(y_0)(f(x) - f(x_0) - Df(x_0)(x - x_0)). \end{aligned}$$
This means
$$\begin{aligned} \|H(x) - H(x_0) - L(x - x_0)\| &\leq \overbrace{\|g(y) - g(y_0) - Dg(y_0)(y - y_0)\|}^{=:(A)} \\ &\quad + \|Dg(y_0)\| \overbrace{\|f(x) - f(x_0) - Df(x_0)(x - x_0)\|}^{=:(B)}. \end{aligned}$$
By differentiability of f at x_0, $(B) \to 0$ as $\|x - x_0\| \to 0$. We also have that, since f differentiable it is Lipschitz continuous, there is some $C > 0$ such that for $\|x - x_0\|$ sufficiently small,
$$(A) = \|y - y_0\| \cdot \frac{(A)}{\|y - y_0\|} \leq C\|x - x_0\| \frac{A}{\|y - y_0\|}.$$

By differentiability of g, the ratio $\frac{\|A\|}{\|y-y_0\|} \to 0$ as $\|y - y_0\| \to 0$. By continuity of f, $\|y - y_0\| = \|f(x) - f(x_0)\|$ will become small as $\|x - x_0\| \to 0$, so that we have in all $\frac{A}{\|x-x_0\|} \to 0$ as $\|x - x_0\| \to 0$. ■

Exercise 1.1: Let f differentiable in $\mathbb{R}^2$ and $g(r, \theta) := (r\cos\theta, r\sin\theta)$ with $(r, \theta) \in (0, \infty) \times [0, 2\pi)$. Let $F(r, \theta) = f(g(r, \theta))$. Compute $\frac{\partial F}{\partial \theta}$ and $\frac{\partial F}{\partial r}$.

§1.1 Aside on Tangent Planes

Let $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}$ differentiable on Ω. Then $Df(x) = \left(\frac{\partial f}{\partial x_1}(x), ..., \frac{\partial f}{\partial x_2}(x)\right) =: \nabla f(x)$, called the *gradient* of f. Let $S := \{(x, z) \in \Omega \times \mathbb{R} : z = f(x)\}$ be the *graph* of f. Then, for $x^0 \in \mathbb{R}$,

$$T_{x_0}S = \left\{(x, z) \in \mathbb{R}^n \times \mathbb{R} : z = f\left(x^0\right) + \sum_{j=1}^{n} \frac{\partial f}{\partial x_j}\left(x^0\right)\left(x_j - x_j^0\right)\right\}$$

is the *tangent plane* to S at x^0.

To see this, let $v \in \mathbb{R}^n$ be a unit vector and $x \in \Omega$. Define $g(t) := f(x + tv)$ for $f : \Omega \to \mathbb{R}$ differentiable (for t sufficiently small, $x + tv$ remains in Ω by openness). We find

$$g'(t) = \langle \nabla f(x + tv), v \rangle$$

for t sufficiently small.

↪Proposition 1.5: Suppose $\nabla f(x) \neq 0$. Then, $\nabla f(x)$ points in the direction of steepest increase of f.

PROOF. For v a unit vector, the *directional derivative* in the direction of v is $D_v f(x) = \langle \nabla f(x), v \rangle = \|\nabla f(x)\| \cos(\theta)$ where θ the angle between $\nabla f(x)$ and v. This is maximized when $\theta = 0$, i.e. when $v = \frac{\nabla f(x_0)}{\|\nabla f(x_0)\|}$. ■

We can rewrite the graph S as the level set $\{(x, z) \in \Omega \times \mathbb{R} \mid g(x, z) = 0\}$ where $g(x, z) := z - f(x)$. Heuristically, $\nabla g(x_0, z_0)$ should be *normal* to the surface S at (x_0, z_0) (for steepest increase). As such, we define

$$T_{(x_0,z_0)}S := \{\nabla g(x_0, z_0) \cdot (x - x_0, z - z_0) = 0\}.$$

Note that

$$\nabla g(x_0, z_0) = \left(-\partial_{x_1} f(x_0), ..., -\partial_{x_n} f(x_0), 1\right),$$

so that

$$T_{(x_0,z_0)} = \{z - z_0 = \nabla f(x_0) \cdot (x - x_0)\},$$

which gives the definition from above.

§1.2 Clairault's Theorem

Here, the question is, given $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}$ twice differentiable, when can we exchange order of second-order partial derivatives, i.e. when is

$$\frac{\partial^2 f}{\partial x_j \partial x_i} = \frac{\partial^2 f}{\partial x_i \partial x_j}, \quad \forall\, i,j = 1, \dots, n?$$

We need to establish first a generalization of the mean-value theorem. First, note that if

$$\gamma : (a,b) \to \mathbb{R}^n, \quad g : \Omega \subset \mathbb{R}^n \to \mathbb{R}$$

are two differentiable functions with $\gamma((a,b)) \subset \Omega$, then by the chain rule, if we put $H(t) := g(\gamma(t))$,

$$\frac{\partial H}{\partial t} = Dg(\gamma(t)) \cdot D\gamma(t), \quad D\gamma(t) = (\gamma_1'(t), \dots, \gamma_n'(t)).$$

↪**Theorem 1.3** (Mean-Value Theorem): Let $B \subset \mathbb{R}^n$ be a ball and $f : B \to \mathbb{R}$ be differentiable for all $x \in B$. Then, for any $x, y \in B$, there exists $z \in B$ such that

$$f(x) - f(y) = Df(z) \cdot (x - y).$$

In particular, $|f(x) - f(y)| \leq \|Df(z)\|\|x - y\|$.

Proof. Let $x, y \in B$ fixed and let $\gamma(t) := tx + (1-t)y$ for $t \in [0,1]$. We see that $\gamma(t) \in B$ for all $t \in [0,1]$, and that $D\gamma(t) = x - y$. Set $F(t) := f(\gamma(t))$ (i.e., we restrict f to its values along the straight line along x and y), noting $F : \mathbb{R} \to \mathbb{R}$. So, by 1-dimensional mean-value theorem, there is some $t^* \in [0,1]$ such that

$$\begin{aligned} f(x) - f(y) = F(1) - F(0) = F'(t^*) &= Df\left(\underbrace{t^* x + (1-t^*)y}_{=:z \in B}\right) \cdot D\gamma(t) \\ &= Df(z) \cdot (x - y). \end{aligned}$$

■

Let $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$ differentiable. Remember that $Df : \Omega \to \mathbb{R}^{m \times n}$.

↪**Definition 1.5**: We say f twice differentiable at x if Df exists locally to x and Df is differentiable at x. We write

$$D^2 f = D(Df),$$

and similarly

$$D^k f := D\left(D^{k-1} f\right)$$

with an analogous definition.

↪**Definition 1.6**: Given $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}$, we see that $f \in C^k(\Omega)$ for $k \in \mathbb{Z}_+$ if all the partial derivatives to order k exist and are continuous in Ω.

↪**Definition 1.7**: If $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}$ twice differentiable, the *Hessian matrix* is given by

$$H_f(x) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1 \partial x_1} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_n} \end{pmatrix}.$$

Exercise 1.2: Let $f(x,y) := \begin{cases} \frac{(xy)(x^2-y^2)}{x^2+y^2} & (x,y)\neq(0,0) \\ 0 & (x,y)=(0,0) \end{cases}$ and compute $H_f(x,y)$.

↪**Theorem 1.4** (Clairault): Let $f : \Omega \subset \mathbb{R}^n \to \mathbb{R}$ be twice differentiable at $x \in \Omega$. Then,

$$\frac{\partial^2 f}{\partial x_i \partial x_j}(x) = \frac{\partial^2 f}{\partial x_j \partial x_i}(x), \quad \forall\, i,j = 1,\ldots,n.$$

↪**Corollary 1.1**: If $\frac{\partial^2 f}{\partial x_i \partial x_j}$ are all continuous at $x \in \Omega$ for $i,j = 1,\ldots,n$, then $\frac{\partial^2 f}{\partial x_i \partial x_j}(x) = \frac{\partial^2 f}{\partial x_j \partial x_i}(x)$.

Proof. (Of Clairault's) It's enough to consider $n = 2$. Fix $(x,y) \in \Omega$, and note that for $s,t \in \mathbb{R}$ sufficiently small, $(x+s, y+t) \in \Omega$ by openness. Set

$$\begin{aligned} \Delta(s,t) &:= f(x+s,y+t) - f(x,y+t) - f(x+s,y) + f(x,y) \\ &= g_t(x+s) - g_t(x), \quad g_t(u) := f(u,y+t) - f(u,y). \end{aligned}$$

By the mean-value theorem, there is some $\xi_{s,t}$ between x and $x+s$ suh that

$$\Delta(s,t) = \frac{\partial g_t}{\partial x}(\xi_{s,t}) \cdot s = \left[\frac{\partial f}{\partial x}(\xi_{s,t}, y+t) - \frac{\partial f}{\partial x}(\xi_{s,t}, y)\right] s. \quad (\ddagger)$$

By assumption, $\frac{\partial f}{\partial x}$ is differentiable at (x,y), so

$$\frac{\partial f}{\partial x}(z_1,z_2) = \frac{\partial f}{\partial x}(x,y)(z_1 - x) + \frac{\partial^2}{\partial x^2}(x,y)(z_2 - y) + E_1(z_1,z_2), \quad (\dagger)$$

where

$$\frac{|E_1(z_1,z_2)|}{\sqrt{(z_1-x)^2 + (z_2-y)^2}} \to 0, \quad \text{as } (z_1,z_2) \to (x,y).$$

Evaluating $(\dagger)$ at $(z_1,z_2) = (\xi_{s,t}, y+t)$ and $(\xi_{s,t}, y)$, and plugging into $(\ddagger)$ yields

$$\Delta(s,t) = \left(\frac{\partial^2 f}{\partial y \partial x}(x,y)t + E_1(\xi_{s,t}, y+t) - E_1(\xi_{s,t}, y)\right)s.$$

Let $s = t$ and let $t \to 0$. We claim that

$$\frac{\partial^2 f}{\partial y \partial x}(x,y) = \lim_{s=t\to 0} \frac{\Delta(s,t)}{st} = \frac{\partial^2 f}{\partial x \partial y}(x,y)$$

■