

# Some Analysis Things

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In these notes, I make some brief comments on the IB Analysis courses<sup>1</sup>.  
Credit is due to Evan Chen for the style file for these notes<sup>2</sup>.

## §1 Integration

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<sup>1</sup>strictly, just Analysis and Topology and Complex Analysis, but I hoep that the reader agrees that Analysis and Topology is ... more than one courses worth of material!

<sup>2</sup>Available here: <https://github.com/vEnhance/dotfiles/blob/master/texmf/tex/latex/evan/evan.sty>.

**Theorem 1.1** (Interchanging Differentiation and Integration)

Let  $f : \mathbb{R} \times [0, 1] \rightarrow \mathbb{R}$  be a cts function of  $\theta$  and  $t$ , and suppose  $\frac{\partial f}{\partial \theta}$  is also cts. Then

$$\frac{d}{d\theta} \int_0^1 f(\theta, t) dt = \int_0^1 \frac{\partial f}{\partial \theta}(\theta, t) dt. \quad (1)$$

*Proof.* The key idea is that differentiability is a local property, so we can force the domain of  $\frac{\partial f}{\partial \theta}$  to be compact.

WLOG let  $\theta = 0$ . Now  $\forall \varepsilon > 0$ , pick  $\delta > 0$  so that  $\left| \frac{\partial f}{\partial \theta}(\theta, t) - \frac{\partial f}{\partial \theta}(0, t) \right| < \varepsilon$  for  $(\theta, t) \in [-\delta, \delta] \times [0, 1]$ . Then let

$$F(\theta) = \int_0^1 f(\theta, t) dt. \quad (2)$$

Then we can directly calculate (for  $|h| < \delta$ )

$$\frac{1}{h}(F(h) - F(0)) = \int_0^1 \frac{f(h, t) - f(0, t)}{h} dt = \int_0^1 \frac{\partial f}{\partial \theta}(\theta_t, t) dt. \quad (3)$$

where the last equality follows from the mean value theorem.  $\theta_t \in (-|h|, |h|)$  is a function of  $t$ <sup>a</sup>.

But our choice of delta means that this differs by at most  $\varepsilon$  from

$$\int_0^1 \frac{\partial f}{\partial \theta}(0, t) dt, \quad (4)$$

so we're done. □

<sup>a</sup>Extra: can we always make it a cts function of  $t$ ?

**Theorem 1.2** (CIF for Derivatives)

Let  $U$  be a domain and  $f : U \rightarrow \mathbb{C}$  holomorphic. Let  $D(0, 1) \subseteq U$  and  $w \in D(0, 1)$ . Then

$$f^{(n)}(w) = \frac{1}{n!} \oint_{\partial D(0,1)} \frac{f(z)}{(z-w)^{n+1}} dz. \quad (5)$$

*Proof.* Case  $n = 1$  is the ordinary integral formula, and we show how the  $n = 2$  arises by applying (1.1). The higher order cases arise similarly.

We can write the  $n = 1$  case as

$$\oint_{\partial D(0,1)} \frac{f(z)}{z-w} dz = \int_0^1 \frac{f(\gamma(t))\gamma'(t)}{\gamma(t)-w} dt. \quad (6)$$

where  $\gamma(t) = e^{2\pi i t}$ . We can now directly apply the previous result, since the integrand's partial  $w$  derivative is indeed cts, provided we localise to a ball around  $w$  (so that the  $\frac{1}{\gamma(t)-w}$  term doesn't get very large). □

## §2 Differentiation

I don't think this part of the course needs to be anywhere near as feared as it is currently.

**Definition 2.1** (Norm). Let  $V$  be a real vector space. A norm on  $V$  is a function  $\|\cdot\| : V \rightarrow \mathbb{R}$  satisfying

- $\|v\| \geq 0$ , with equality iff  $v = 0$ .
- $\|v + w\| \leq \|v\| + \|w\|$ .
- $\|\lambda v\| = |\lambda| \|v\|$ .

**Remark 2.2.** This naturally induces a topology on  $V$  by turning it into a metric space with distance function  $d(v, w) = \|v - w\|$ .

**Theorem 2.3** (Only one norm)

Let  $V$  be a finite dimensional vector space. Then all norms on  $V$  are Lipschitz equivalent.

*Proof.* Fix a basis  $e_1, \dots, e_n$  of  $V$ . Then let  $\|\cdot\|_2$  be the Euclidean norm, i.e

$$\|\lambda_1 e_1 + \dots + \lambda_n e_n\|_2 = \sqrt{\lambda_1^2 + \dots + \lambda_n^2}. \quad (7)$$

We show that all norms are Lipschitz equivalent to the Euclidean norm, and since Lipschitz equivalence is an equivalence relation, this will suffice.

- $\exists m > 0$  such that  $\|v\| \leq m\|v\|_2$  for all  $v \in V$ :

Direct application of the triangle inequality. Let

$$E = \max_{i=1}^n \|e_i\| > 0. \quad (8)$$

Then if  $v = \lambda_1 e_1 + \dots + \lambda_n e_n$ , and

$$\Lambda = \max_{i=1}^n |\lambda_i|, \quad (9)$$

then

$$\|v\| \leq |\lambda_1| \|e_1\| + \dots + |\lambda_n| \|e_n\| \leq E(|\lambda_1| + |\lambda_2| + \dots + |\lambda_n|) \quad (10)$$

where the first inequality follows from the triangle inequality. So

$$\|v\| \leq nE\Lambda. \quad (11)$$

Now  $\|v\|_2 \geq \Lambda$  and hence  $m = nE$  works (note we need  $\Lambda$  independence, but  $E$  dependence is fine since the former is a property of the specific  $v$ , but the latter a property of the norm).

- $\exists M > 0$  such that  $\|v\| \geq M\|v\|_2$  for all  $v \in V$ :

Less obvious.

□

Why does this matter? Recall the definition of differentiability

**Definition 2.4.**  $f : \mathbb{R}^m \rightarrow \mathbb{R}^n$  is differentiable at  $p$  with derivative the linear map  $T(p)(h)$  if

$$\lim_{h: \|h\|_2 \rightarrow 0} \frac{\|f(p+h) - f(p) - T(p)(h)\|_2}{\|h\|_2}. \quad (12)$$

The important thing to notice here is that while  $f$  and  $T(p)$  are both maps from  $\mathbb{R}^m$ , the former exclusively maps from points very close to  $p$ , and the latter exclusively maps from points very close to 0. If this isn't clear, chapter 42 of [1] (Napkin) provides a far better explanation than what I can give (and includes pictures!).

To be concrete however, working with a multi-dimensional limit is hard. In practice, we are likely to generally use the following result (cross-posted from my Geometry notes)

**Theorem 2.5** (Computing derivatives in  $n$  dimensions.)

Suppose that  $U \subseteq^\circ \mathbb{R}^m$  and  $f : U \rightarrow \mathbb{R}$  has continuous partial derivatives at  $p \in U$ . Then  $f$  is differentiable at  $p$ , with derivative

$$Df(p)(h) = \frac{\partial f}{\partial x_1} h_1 + \cdots + \frac{\partial f}{\partial x_m} h_m. \quad (13)$$

*Proof.* Decompose as a telescoping sum

$$f(x_1 + h_1, \dots, x_m + h_m) - f(x_1, \dots, x_m) \quad (14)$$

$$= f(x_1 + h_1, \dots, x_m + h_m) - f(x_1 + h_1, \dots, x_{m-1} + h_{m-1}, x_m) \quad (15)$$

$$+ f(x_1 + h_1, \dots, x_{m-1} + h_{m-1}, x_m) - f(x_1 + h_1, \dots, x_{m-1}, x_m) \quad (16)$$

$$+ \cdots \quad (17)$$

$$+ f(x_1 + h_1, x_2, \dots, x_m) - f(x_1, \dots, x_m). \quad (18)$$

And now by an ‘ $m$ - $\varepsilon$ ’ proof (i.e  $m$  applications of the triangle inequality), we use continuity of the  $m$  partial derivatives to deduce the desired derivative expression.

However, actually doing this is somewhat more subtle than it may seem at first (for example, to me!), since the limit  $\|h\|_2 \rightarrow 0$  need be independent of the relative ‘speeds’ that each  $h$  component approach 0.

We illustrate how we do this in the case  $m = 2$ ; the argument is easily generalised to the higher order cases.

**Example 2.6**

WLOG show differentiability at 0.

The quantity  $Q$  that we need show approaches 0 as  $h_1 \rightarrow 0$  and  $h_2 \rightarrow 0$  is

$$Q = \frac{f(h_1, h_2) - f(0, 0) - h_1 \frac{\partial f}{\partial x}|_0 - h_2 \frac{\partial f}{\partial y}|_0}{\sqrt{h_1^2 + h_2^2}} \quad (19)$$

Now by the triangle inequality, and that  $\sqrt{h_1^2 + h_2^2} \geq |h_1|, |h_2|$ ,

$$|Q| \leq \left| \frac{f(h_1, h_2) - f(0, h_2) - h_1 \frac{\partial f}{\partial x}|_0}{h_1} \right| + \left| \frac{f(0, h_2) - f(0, 0) - h_2 \frac{\partial f}{\partial y}|_0}{h_2} \right| \quad (20)$$

as  $h_2 \rightarrow 0$ , clearly the latter term approaches 0 (this is easy due to the remark (2.8) made below). However, for the first term the ‘different speeds’ issue immediately arises. To remedy this, what we can do is localise to an open set  $U \subseteq^\circ \mathbb{R}^2$  around 0 so that when  $(x, y) \in U$ ,

$$\left| \frac{\partial f}{\partial x}|_{(x,y)} - \frac{\partial f}{\partial x}|_0 \right| < \varepsilon. \quad (21)$$

Then this means that if  $(h_1, h_2) \in U$ ,

$$\frac{f(h_1, h_2) - f(0, h_2)}{h_1} \quad (22)$$

differs from  $\frac{\partial f}{\partial x}|_0$  by at most  $\varepsilon$ , since otherwise by the *mean value theorem*, somewhere in  $U$  we would have  $\frac{\partial f}{\partial x}$  differing by more than  $\varepsilon$  from  $\frac{\partial f}{\partial x}|_0$ .

**Remark 2.7.** This is basically the same sort of thing we did in IA DEs when we integrated from  $(x_1, y_1)$  to  $(x_2, y_2)$  by first travelling from  $(x_1, y_1)$  to  $(x_2, y_1)$  (with fixed  $y$  value) and then from  $(x_2, y_1)$  to  $(x_2, y_2)$  (with fixed  $x$  value), like a staircase.

**Remark 2.8.** Actually, the last term in the telescoping series,  $f(x_1 + h_1, x_2, \dots, x_m) - f(x_1, \dots, x_m)$  is the partial derivative (in  $x_1$ ) of  $f$  at  $(x_1, \dots, x_n)$ . So we only need continuity of  $n - 1$  of the partial derivatives at a point, and existence of the last, in order to deduce differentiability at a point.

Look again at (2.4). There's nothing special about  $\|\cdot\|_2$ ! For our purposes, a more useful norm is the **operator norm**.

**Definition 2.9** (Operator norm). The **operator norm** on the space of linear maps  $L(\mathbb{R}^m, \mathbb{R}^n)$  is the value

$$\sup_{x \neq 0} \frac{\|L(x)\|_2}{\|x\|_2}. \quad (23)$$

This has the property that it is *sub-multiplicative*, i.e.  $\|AB\| \leq \|A\| \|B\|$  (exercise to reader).

**Example 2.10 (2019 P1L)**

The matrix function  $f(M) = M^{-1}$  is differentiable at  $I$ .

*Proof.* The following proof relies on the fact that the set of  $n \times n$  invertible matrices is an open subset in the set of  $n \times n$  matrices.

This is quite a nice result in its own right, and follows from the following lemma

**Lemma 2.11**

The determinant map  $\det : M_{n \times n}(\mathbb{R}) \rightarrow \mathbb{R}$  is cts.

*Proof.* The map is simply a polynomial in the  $n^2$  variables, which (check if unsure) are cts.  $\square$

Due to this, since  $\mathbb{R} \setminus \{0\}$  is open, so the set of invertible matrices is open since it is  $\det^{-1}(\mathbb{R} \setminus \{0\})$ .

Back to the long question: we first (e.g by checking the one-dimensional case) convince ourselves that the answer is  $-I$ . This leaves us to verify that

$$\frac{\|(I + H)^{-1} - I + H\|}{\|H\|} \rightarrow 0. \quad (24)$$

Now we can check that

$$(I + H)^{-1} - I + H = H^2(I + H)^{-1}. \quad (25)$$

So using the sub-multiplicative property, since

$$\frac{\|(I + H)^{-1} - I + H\|}{\|H\|} \leq \|H(I + H)^{-1}\| \leq \|H\| \|(I + H)^{-1}\| \quad (26)$$

we only need show that that  $(I + H)^{-1}$  has bounded norm for sufficiently small  $H^a$ . But this is true since suppose  $\|v\|_2 = \|w\|_2 = 1$  and

$$(I + H)^{-1}v = \lambda w, \quad (27)$$

Then

$$\frac{1}{\lambda}v = Iw + Hw. \quad (28)$$

Now the magnitude of the RHS is going to arbitrarily close to 1 due to the bounded operator norm of  $H$ . So  $\lambda$  can't grow large.  $\square$

<sup>a</sup>I can't reason this part as cleanly as I hope the rest of the proof is reasoned!

## §3 Notation and Glossary

### §3.1 Notation

### §3.2 Glossary

- Cts: continuous.



## References

- [1] Evan Chen (2021), *An Infinitely Large Napkin*, <https://venhance.github.io/napkin/Napkin.pdf>.