

# LESSON 2 - 10 MARCH 2021

## HILBERT SPACES

HILBERT  $\subseteq$  BANACH  $\subseteq$  COMPLETE METRIC  $\subseteq$  TOPOLOGICAL

## INNER PRODUCT SPACES

Let  $H$  be a real vector space. A function  $\langle \cdot, \cdot \rangle : H \times H \rightarrow \mathbb{R}$  is an INNER PRODUCT if

- $\langle x, y \rangle = \langle y, x \rangle$ ,  $\forall x, y \in H$  (Symmetric)
- $\langle \lambda x + \mu y, z \rangle = \lambda \langle x, z \rangle + \mu \langle y, z \rangle$ ,  $\forall \lambda, \mu \in \mathbb{R}, x, y, z \in H$  (Bilinear)
- $\langle x, x \rangle \geq 0$ ,  $\forall x$  and  $\langle x, x \rangle = 0$  iff  $x = 0$  (Positive definite)

The pair  $(H, \langle \cdot, \cdot \rangle)$  is called an INNER PRODUCT SPACE

REMARK  $(H, \langle \cdot, \cdot \rangle)$  inner product space. Then  $\|x\| = \sqrt{\langle x, x \rangle}$  defines a norm on  $X$ .

CANACHY-SCHWARTZ INEQUALITY  $H$  inner prod space. Then  $|\langle x, y \rangle| \leq \|x\| \|y\|$

HILBERT SPACE  $(H, \langle \cdot, \cdot \rangle)$  inner product space. We say that  $H$  is a HILBERT SPACE if  $(H, \|\cdot\|)$  is COMPLETE, with  $\|x\| = \sqrt{\langle x, x \rangle}$

BASIS  $H$  Hilbert. A set of elements  $\{e_n\}_{n \in \mathbb{N}} \subseteq H$  is called BASIS if

- $\langle e_i, e_j \rangle = 0$  for all  $i \neq j$ ,  $\langle e_i, e_i \rangle = 1 \quad \forall i$
- $\text{span}\{e_i\}$  is dense in  $H$

where for a set  $A \subseteq H$  we define  $\text{span}A = \left\{ \sum_{j=1}^n \lambda_j x_j \mid \lambda_j \in \mathbb{R}, x_j \in A, n \in \mathbb{N} \right\}$

THEOREM If  $H$  is Hilbert separable then  $\exists \{e_n\} \subseteq H$  basis

NOTATION Given a basis  $\{e_n\}$  and  $x \in H$  we define  $x_k := \langle x, e_k \rangle$   
k-th coordinate of  $x$  wrt  $\{e_n\}$

PROPOSITION If Hilbert with basis  $\{e_n\}$ . Then

$$1) \quad \|v\|^2 = \sum_{n=1}^{+\infty} v_n^2, \quad v_n := \langle v, e_n \rangle$$

$$2) \quad \langle v, w \rangle = \sum_{n=1}^{+\infty} v_n w_n$$

For a separable Hilbert space there is a natural correspondence between  $H$  and  $\ell^2$ . Thus we can think of  $H$  as  $\mathbb{R}^\infty$ . To make this statement precise we need the following definition

(Recall:  $\ell^2 = \{(x_1, x_2, \dots) \mid \sum_{j=1}^{+\infty} |x_j|^2 < +\infty\}$ . This is normed by  $\|x\| = \left(\sum_{j=1}^{+\infty} |x_j|^2\right)^{1/2}$  and is Hilbert with inner product  $\langle x, y \rangle = \sum_{j=1}^{+\infty} x_j y_j$ )

DEF  $X$  normed space,  $\{x_n\} \subseteq X$ . We say that

$$\sum_{n=1}^{+\infty} x_n = x_0$$

if  $s_n \rightarrow x_0$ , where  $s_n := \sum_{j=1}^n x_j$  partial sums.

THEOREM  $H$  Hilbert with basis  $\{e_n\}$ . Let  $\{v_n\} \subseteq \mathbb{R}$ . Then

$$\sum_{n=1}^{+\infty} v_n e_n \text{ converges in } H \iff \sum_{n=1}^{+\infty} v_n^2 \text{ converges in } \mathbb{R}$$

In particular  $H \cong \ell^2$ , with isomorphism  $v \in H \mapsto (v_1, v_2, \dots, v_n, \dots) \in \ell^2$

Another nice aspect of Hilbert spaces is that  $H = H^*$  dual space.

THEOREM (RIESZ)  $H$  Hilbert. Define the map  $\bar{\Phi}: H \rightarrow H^*$

$$\bar{\Phi}(x)(z) := \langle x, z \rangle, \quad \forall z \in H$$

Then  $\bar{\Phi}$  is invertible and  $\|\bar{\Phi}(x)\|_{H^*} = \|x\|_H$ .  
Thus  $H \cong H^*$  isomorphic.

In particular  $H$  can be identified with  $H^*$ . Thus weak\* and weak topologies coincide, and we can characterize weak convergence by

$$x_n \rightarrow x_0 \iff \langle x_n, z \rangle \rightarrow \langle x_0, z \rangle, \forall z \in H.$$

## FURTHER PROPERTIES OF WEAK CONVERGENCE IN HILBERT

PROP  $H$  Hilbert with basis  $\{e_n\}$ . If  $x_n \rightarrow x_0$  then

$$(x_n)_k \rightarrow (x_0)_k, \forall k \in \mathbb{N}$$

### WARNING

We know that  $x_n \rightarrow x_0$  does not imply  $x_n \rightarrow x_0$ . However it is also not true that  $\|x_n\| \rightarrow \|x_0\|$  (i.e. the norm is not weakly continuous).

However, the following proposition relating strong convergence  $\Rightarrow$  weak convergence holds.

PROP

$H$  Hilbert. Then

$$x_n \rightarrow x_0 \iff x_n \rightarrow x_0 \text{ and } \|x_n\| \rightarrow \|x_0\|$$



NOTE: This is not saying that  
 $\|x_n - x_0\| \rightarrow 0$

Another useful proposition is that weak convergence can be tested against a dense subset

PROP

$H$  Hilbert. Assume that  $\{x_n\} \subseteq H$  is bounded, i.e.

$\sup_n \|x_n\| < +\infty$ . Suppose that  $W \subseteq H$  is s.t.  $\overline{\text{span } W} = H$  and

$$\langle x_n, w \rangle \rightarrow \langle x_0, w \rangle, \forall w \in W.$$

Then  $x_n \rightarrow x_0$ .

## COROLLARY

H Hilbert with basis  $\{e_n\}$ . Let  $\{x_n\}$  be BOUNDED.  
Then if

$$(x_n)_k \rightarrow (x)_k, \forall k \in \mathbb{N}$$

We have  $x_n \rightarrow x_0$ .

[Proof: take  $N = \{e_n\}$ ]

## EXAMPLE

$X = C[a, b]$  with  $\|u\|_\infty := \max_{x \in [a, b]} |u(x)|$

$Y = C^1[a, b]$  with  $\|u\|_1 := \|u\|_\infty + \|u'\|_\infty$

Then  $(X, \|\cdot\|_\infty)$  and  $(Y, \|\cdot\|_1)$  are Banach spaces, but not Hilbert spaces.



Hint to show this: in an inner product space the parallelogram law holds:

$$\|x+y\|^2 + \|y-x\|^2 = 2(\|x\|^2 + \|y\|^2) \quad \forall x, y \in H$$

## 1. CALCULUS IN NORMED SPACES

Reference : S. KESAVAN

"NONLINEAR FUNCTIONAL ANALYSIS,  
A FIRST COURSE"

HINDUSTAN BOOK AGENCY, 2004

Throughout this section  $X, Y$  are real NORMED SPACES,  $U \subseteq X$  is OPEN and  $F: U \rightarrow Y$  is a given function.

GOAL: Construct a theory of differentiation for maps  $F: U \rightarrow Y$

### DEFINITION 1.1

We say that  $F$  is FÉCHET DIFFERENTIABLE at  $u_0 \in U$  if  $\exists A_{u_0} \in \mathcal{J}(X, Y)$  s.t.

$$\lim_{\|v\|_X \rightarrow 0} \frac{\|F(u_0 + v) - F(u_0) - A_{u_0}(v)\|_Y}{\|v\|_X} = 0.$$

### REMARK

If  $F$  is diff. at  $u_0$  then  $A_{u_0} \in \mathcal{J}(X, Y)$  satisfying

(\*) is UNIQUE (Check it by exercise)

### NOTATION

If  $F$  is diff. at  $u_0 \in U$  we call  $A_{u_0}$  the FÉCHET DERIVATIVE (or just derivative) of  $F$  at  $u_0$ . We denote

$$F'(u_0) := A_{u_0} \in \mathcal{J}(X, Y)$$

### NOTE

This generalizes diff. for maps  $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ . In this case the differential is  $F'(u)(v) = DF(u)v$ , with  $DF(u) \in \mathbb{R}^{m \times n}$  matrix of partial derivatives of  $F$ , i.e.

$$[DF(u)]_{ij} = \frac{\partial F_i}{\partial x_j}(u), \quad F = (F_1, \dots, F_m)$$

## DEFINITION 1.2

Assume that  $F$  is diff. at  $u_0 \in U$ . Then we can define the map

$$F': U \rightarrow J(X, Y)$$

$$u \mapsto F'(u)$$

If  $F'$  is continuous we say that  $F \in C^1(U, Y)$

$\uparrow$   
WRT norm on  $X$  and  
operator norm on  $J(X, Y)$

$\uparrow$   
In words we  
say  $F$  is  
continuously diff.

## EXAMPLES

The most common examples throughout the course will be real valued functions, i.e.,  $Y = \mathbb{R}$

1)  $X$  normed,  $U \subseteq X$  open,  $F: U \rightarrow \mathbb{R}$ . If  $F$  is diff at  $u_0 \in U$  then  $F'(u_0) \in J(X, \mathbb{R}) = X^*$

Then if  $F$  diff in  $U$ , the derivative defines an application  $F': U \rightarrow X^*$  ( $u \mapsto F'(u) \in X^*$ )

2)  $H$  Hilbert,  $U \subseteq H$  open,  $F: U \rightarrow \mathbb{R}$ . If  $F$  is diff. at  $u_0 \in U$  then  $F'(u_0) \in J(H, \mathbb{R})$ . By Riesz's Thm  $\exists! z_0 \in H$  s.t.

$$F'(u_0)(w) = \langle z_0, w \rangle, \quad \forall w \in H$$

We denote  $z_0 := \text{grad } F(u_0)$  (gradient of  $F$  at  $u_0$ ).

### PROPOSITION 1.3

Assume that  $F$  is diff at  $u_0 \in U$ . Then  $F$  is continuous at  $u_0$ .

Proof Introduce the notation  $\circ(\|v\|_X)$  for a quantity such that

$$\frac{\circ(\|v\|_X)}{\|v\|_X} \rightarrow 0 \text{ as } \|v\|_X \rightarrow 0, \text{ since } U \text{ is open and } u_0 \in U \text{ then } \exists \varepsilon > 0$$

such that  $B_\varepsilon(u_0) \subseteq U$ . Let  $v \in B_\varepsilon(0)$  so that  $u_0 + v \in B_\varepsilon(u_0) \subseteq U$ : then

$$\begin{aligned} \|F(u_0 + v) - F(u_0)\|_Y &\leq \|F(u_0 + v) - F(u_0) - A_{u_0}(v)\|_Y + \|A_{u_0}(v)\|_Y \\ &\leq \circ(\|v\|_X) + \|A_{u_0}\|_{J(X,Y)} \|v\|_X \quad (\text{since } A_{u_0} \in J(X,Y)) \\ &= \circ(\|v\|_X) \rightarrow 0 \text{ as } \|v\|_X \rightarrow 0 \end{aligned} \quad \square$$

### THEOREM 1.4 (CHAIN RULE)

$X, Y, Z$  normed,  $U \subseteq X$ ,  $V \subseteq Y$  open,  $F: U \rightarrow V$ ,  $G: V \rightarrow Z$ . Assume  $F$  is diff at  $u_0 \in U$  and  $G$  is diff. at  $v_0 := F(u_0) \in V$ . Then  $G \circ F: U \rightarrow Z$  is diff. at  $u_0$  with

$$(G \circ F)'(u_0) = G'(v_0) \circ F'(u_0) \in J(X, Z)$$

$\uparrow$   
Composition of linear continuous operators in  $J(X, Y)$  and  $J(Y, Z)$

The proof is very simple, and we thus omit it. If you are interested you can find it in the book of KESAVAN, PROPOSITION 1.1.1 page 7

### DEFINITION 1.5

We say that  $F$  is GATEAUX DIFFERENTIABLE at  $u_0 \in U$  in the DIRECTION  $v \in X$  if

$$F'_g(u_0)(v) := \lim_{t \rightarrow 0} \frac{F(u_0 + tv) - F(u_0)}{t} \in Y \quad ,$$

Here  $t \in \mathbb{R}$

i.e., if the above limit exists.

Note The Gâteaux derivative (G.-derivative) generalizes the directional derivative for maps  $F: \mathbb{R}^n \rightarrow \mathbb{R}$ . In this case we have

$$F'_G(u)(\sigma) = \frac{\partial F}{\partial \sigma}(u) = JF(u_0) \cdot \sigma$$

↑  
If  $F$  diff.  
at  $u$

↑  
Scalar product  
in  $\mathbb{R}^n$

WARNING  $F'_G(u): X \rightarrow Y$  is always linear. But in general we don't have that  $F'_G(u) \in \mathcal{J}(X, Y)$ , as it happens for Fréchet derivatives.

PROPOSITION 1.6 If  $F$  diff. at  $u_0 \in U$  then  $F$  is G.-diff. at  $u_0$  in every direction  $\sigma$  and

$$F'_G(u_0)(\sigma) = F'(u_0)(\sigma)$$

Gâteaux derivative can be computed by Fréchet derivative,  
and viceversa

Proof 
$$\frac{F(u_0 + t\sigma) - F(u_0)}{t} =$$

$$= \frac{F(u_0 + t\sigma) - F(u_0) - F'(u_0)(t\sigma)}{t \|\sigma\|_X} \|\sigma\|_X + \frac{F'(u_0)(t\sigma)}{t}$$

$\underbrace{\quad}_{\rightarrow 0 \text{ as } t \rightarrow 0, \text{ since } F \text{ is diff. at } u_0}$  (note this is converging in  $Y$ )  
to 0

$$= o(t) + F'(u_0)(\sigma) \rightarrow F'(u_0)(\sigma) \text{ as } t \rightarrow 0$$

Here we used that  $F'(u_0)$   
is lim. operator, so  $F'(u_0)(t\sigma) =$   
 $tF'(u_0)(\sigma)$

□

## WARNING

The converse of prop 1.6 does not hold, i.e.

Gâteaux diff.  $\not\Rightarrow$  Fréchet diff.

For example take  $F: \mathbb{R}^2 \rightarrow \mathbb{R}$

$$F(x, y) := \begin{cases} \frac{x^5}{(y-x^2)^2+x^4} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0) \end{cases}$$

It is easy to check that  $F'_g(0)(r) = 0$ ,  $\forall r \in \mathbb{R}$ . So  $F$  is G-diff at  $0 = (0, 0)$  in every direction. Thus, if  $F$  was Fréchet diff we would have (by Prop 1.6) that  $F'(0)(r) = 0$ . Thus by def of derivative

(\*) 
$$\lim_{\substack{\|v\|_{\mathbb{R}^2} \rightarrow 0 \\ v \in \{y=x^2\}}} \frac{|F(v) - F(0)|}{\|v\|_{\mathbb{R}^2}} = 0$$

However if in the above limit we consider  $v \in \{y=x^2\}$   
we obtain

$$\lim_{\substack{\|v\|_{\mathbb{R}^2} \rightarrow 0 \\ v \in \{y=x^2\}}} \frac{|F(v) - F(0)|}{\|v\|_{\mathbb{R}^2}} = 1,$$

which contradicts (\*). Thus  $F$  is not Fréchet diff at 0.

## REMARK

Proposition 1.6 is very useful to guess the Fréchet derivative of a function  $F: U \rightarrow Y$ , as the Gâteaux derivative can be computed via a formula

## EXAMPLE

$X = C[0,1]$ , with norm  $\| \cdot \|_\infty$ . Define  $F: X \rightarrow \mathbb{R}$  by

$$F(u) = \int_0^1 \sin(u(x)) dx$$

What could be the derivative of  $F$ ? Let us compute the Gâteaux derivative at  $u$  in the direction  $v$ :

$$\lim_{t \rightarrow 0} \frac{F(u+tv) - F(u)}{t} = \lim_{t \rightarrow 0} \int_0^1 \frac{\sin(u+tv) - \sin(u)}{t} dx$$

From the Chain Rule (THEOREM 1.4) we have:

$$\lim_{t \rightarrow 0} \frac{\sin(u(x)+tv(x)) - \sin(u(x))}{t} = \cos(u(x)) v(x), \text{ uniformly in } x \in [0,1]$$

Thus we can pass the limit under the integral and obtain

$$F'_g(u)(v) = \int_0^1 \cos(u(x)) v(x) dx. \quad (\text{$F'_g(u)$ is linear!})$$

If  $F$  is Fréchet diff. then by PROPOSITION 1.6 we must have  $F'(u) = F'_g(u)$ . So

We guess that  $F'_g(u)$  is the Fréchet derivative of  $F$  at  $u$ . Indeed notice that  $F'_g(u) \in J(X, \mathbb{R})$ , since

$$|F'_g(u)(v)| \leq \|v\|_\infty \int_0^1 |\cos(u(x))| dx \leq \|v\|_\infty \Rightarrow \sup_{\|v\|_X \leq 1} |F'_g(u)(v)| < +\infty.$$

Moreover it is easy to see that

$$\lim_{\|v\|_X \rightarrow 0} \frac{|F(u+v) - F(u) - F'_g(u)(v)|}{\|v\|_X} = 0$$

Showing that  $F$  is Fréchet diff. at  $u$  with  $F'(u)(v) = \int_0^1 \cos(u(x)) v(x) dx$ .

## QUESTION

Assume that  $F: V \rightarrow Y$  is Gâteaux differentiable. Under which assumptions on  $F$  are we guaranteed Fréchet differentiability?

To answer the above question, we need the following theorem.

### THEOREM 1.7 (MEAN VALUE)

Suppose  $F$  is G-diff. in  $J$  in every direction. Let  $x_1, x_2 \in U$  be such that the segment

$$[x_1, x_2] := \{x_1 + t(x_2 - x_1), t \in [0, 1]\} \subseteq U$$

Assume also that  $F'_g(u) : X \rightarrow Y$  is s.t.  $F'_g(u) \in J(X, Y)$ ,  $\forall u \in J$ .

Then

$$\|F(x_2) - F(x_1)\|_Y \leq \sup_{u \in [x_1, x_2]} \|F'_g(u)\|_{J(X, Y)} \|x_2 - x_1\|_X$$

Proof If  $F(x_1) = F(x_2)$  then the thesis is trivial. Thus assume that  $F(x_1) \neq F(x_2)$ . We now employ the following:

FACT (COROLLARY OF HAHN-BANACH)  $Y$  normed space,  $z \in Y, z \neq 0$ . Then  $\exists \Lambda \in Y^*$  s.t.

$$\|\Lambda\|_{Y^*} = 1 \text{ and } \Lambda(z) = \|z\|_Y.$$

Thus let  $\Lambda \in Y^*$  be such that  $\Lambda(F(x_2) - F(x_1)) = \|F(x_2) - F(x_1)\|_Y$  and  $\|\Lambda\|_{Y^*} = 1$ .

Define the segment function  $\alpha : [0, 1] \rightarrow U$  by  $\alpha(t) := x_1 + t(x_2 - x_1)$ . Consider the map  $H : [0, 1] \rightarrow \mathbb{R}$  defined by

$$H := \Lambda \circ F \circ \alpha.$$

The thesis is obtained by applying the classical Mean Value Theorem to  $H$ . Thus, all we need to show is that  $H$  is differentiable.

WARNING: It would be tempting to apply the Chain Rule of Theorem 1.4 to  $H$ . However  $F$  is only Gâteaux differentiable, and in general the Chain Rule does not apply in this case.

We will check by hand that  $H$  is differentiable. Thus let  $t \in [0, 1]$ , and  $\tau \neq 0$  be such that  $(t+\tau) \in [0, 1]$ . We have

$$\textcircled{*} \quad \frac{H(t+\tau) - H(t)}{\tau} = \Lambda \left[ \frac{F(\alpha(t+\tau)) - F(\alpha(t))}{\tau} \right] \quad (\text{being } \Lambda \text{ linear})$$

Note that  $F(\alpha(t+\tau)) = F(\alpha(t) + \tau(x_2 - x_1))$ . Since  $F$  is  $\hat{g}$ -tame and diff. at  $\alpha(t)$  we then get

$$\frac{F(\alpha(t+\tau)) - F(\alpha(t))}{\tau} = \frac{F(\alpha(t) + \tau(x_2 - x_1)) - F(\alpha(t))}{\tau} \rightarrow F'_g(\alpha(t))(x_2 - x_1)$$

as  $\tau \rightarrow 0$ . Note that by definition the above convergence is WRT the norm of  $Y$ . As  $\Lambda \in Y^*$  is continuous, by taking the limit as  $\tau \rightarrow 0$  in  $\textcircled{*}$  we get

$$H'(t) = \lim_{\tau \rightarrow 0} \frac{H(t+\tau) - H(t)}{\tau} = \Lambda \left[ F'_g(\alpha(t))(x_2 - x_1) \right].$$

In particular  $H$  is diff. in  $[0, 1]$ . Therefore we can apply the Mean Value Theorem to find  $\xi \in (0, t)$  such that

$$\textcircled{**} \quad H(1) - H(0) = H'(\xi) \quad (= H'(\xi)(1-0))$$

Now

$$\begin{aligned} H(1) - H(0) &= \Lambda [F(\alpha(1))] - \Lambda [F(\alpha(0))] \\ &= \Lambda [F(x_2)] - \Lambda [F(x_1)] \\ &= \Lambda [F(x_2) - F(x_1)] \\ &= \|F(x_2) - F(x_1)\|_Y \end{aligned}$$

by the properties of  $\Lambda$ .

On the other hand, as we computed, we have

$$H'(\xi) = \Lambda [F_g^1(\alpha(\xi)) (x_2 - x_1)]$$

and so

$$|H'(\xi)| \leq \|\Lambda\|_{Y^*} \|F_g^1(\alpha(\xi))\|_{J(x,Y)} \|x_2 - x_1\|_X$$

(Using that  $\Lambda$   
and  $F_g^1(\alpha(\xi))$   
are linear and  
bounded)

$$\leq \sup_{u \in [x_1, x_2]} \|F_g^1(u)\|_{J(x,Y)} \|x_2 - x_1\|_X$$

(Using that  $\|\Lambda\|_{Y^*} = 1$   
and that  $\alpha(\xi) \in [x_1, x_2]$ )

From  $\star\star$  we then get

$$\|F(x_2) - F(x_1)\|_Y = H(1) - H(0) = H'(\xi)$$

$$\leq \sup_{u \in [x_1, x_2]} \|F_g^1(u)\|_{J(x,Y)} \|x_2 - x_1\|_X$$

□

COROLLARY 1.8 (of MEAN VALUE) Make the same assumptions of Theorem 1.7. Then

$$\|F(x_2) - F(x_1) - F_g^1(x_1)(x_2 - x_1)\|_Y \leq \sup_{u \in [x_1, x_2]} \|F_g^1(u) - F_g^1(x_1)\|_{J(x,Y)} \|x_2 - x_1\|_X$$

Proof Define  $H: U \rightarrow Y$  by  $H(u) := F(u) - F_g^1(x_1)(u)$ . Note that

$F_g^1(x_1) \in J(x, Y)$  by assumption. In particular  $F_g^1(x_1): X \rightarrow Y$  is Fréchet differentiable with derivative constantly equal to itself.

(Check it by exercise:  $X, Y$  normed spaces,  $T \in J(X, Y)$ . Then  $T$  is Fréchet diff with  $T'(u) = T$ ,  $\forall u \in X$ ).

Therefore  $H$  is Gâteaux diff. with  $H_g^1(u) = F_g^1(u) - F_g^1(x_1)$ . Thus  $H_g^1(u) \in J(x, Y)$  for all  $u \in U$ . Thus  $H$  satisfies assumptions of THEOREM 1.7. Applying THM 1.7 to  $H$  we conclude.

□

We are finally ready to answer our question:

"When does Gâteaux diff. imply Fréchet diff.?"

### THEOREM 1.9

Assume that  $F: U \rightarrow Y$  is Gâteaux diff. at every point of  $U$  and in every direction. Also suppose that  $F'_g(u) \in J(X, Y)$  for all  $u \in U$ . Define the map

$$\begin{aligned} F'_g: U &\longrightarrow J(X, Y) \\ u &\mapsto F'_g(u) \end{aligned}$$

If  $F'_g$  is continuous at  $u_0$  then  $F$  is Fréchet diff. at  $u_0$  and

continuity is wrt norm on  $X$  and operator norm on  $J(X, Y)$

$$F'(u_0)(v) = F'_g(u_0)(v), \quad \forall v \in X$$

Proof Apply COROLLARY 1.8 to points  $x_1 := u_0$ ,  $x_2 := u_0 + v$ . Since  $U$  is open, for  $v$  s.t.  $\|v\|_X$  is sufficiently small we have  $[x_1, x_2] \subseteq U$ . By COROLLARY 1.8 we have

$$\textcircled{x} \|F(u_0 + v) - F(u_0) - F'_g(u_0)(v)\|_Y \leq \sup_{u \in [x_1, x_2]} \|F'_g(u) - F'_g(u_0)\|_{J(X, Y)} \|v\|_X$$

Recall that  $[x_1, x_2] = [u_0, u_0 + v]$ . As  $F'_g$  is continuous at  $u_0$  we have

$$\sup_{u \in [x_1, x_2]} \|F'_g(u) - F'_g(u_0)\|_{J(X, Y)} \rightarrow 0 \quad \text{as } \|v\|_X \rightarrow 0$$

(in practice this implies  $[x_1, x_2] \rightarrow \{u_0\}$ )

Therefore, dividing  $\textcircled{x}$  by  $\|v\|_X$  and taking the limit as  $\|v\|_X \rightarrow 0$  concludes. □