

# LESSON 1 - 3 MARCH 2021

CALCULUS OF VARIATIONS: The study of minimization problems

$X = \text{set}$ ,  $F: X \rightarrow \mathbb{R}$  (often  $\mathbb{R} \cup \{\pm\infty\}$ ) function

We want to solve:

$$\min \{F(u) \mid u \in X\}, \quad \underset{\text{argmin}}{\text{argmin}} \{F(u) \mid u \in X\}$$

↑  
This is a real number,  
called MINIMUM

↑  
These are elements of  $X$ ,  
called MINIMIZERS

We will look at the following classes of methods to study minimization problems:

- INDIRECT METHODS
- DIRECT METHODS
- RELAXATION
- $\Gamma$ -CONVERGENCE

Let us see some basic examples to see what the above words mean:

EXAMPLE 1  $F: \mathbb{R} \rightarrow \mathbb{R}$ ,  $F(x) = x^2 - 4x$ . What is  $\min\{F(x) \mid x \in \mathbb{R}\}$ ?

INDIRECT METHOD: Find candidate minimizers. If  $\hat{x}$  is min. then  $F'(\hat{x}) = 0$ . Now  $F'(x) = 2x - 4 = 0$  iff  $x = 2$ . So  $\hat{x} = 2$  is a candidate minimizer and minimum value is  $F(2) = -4$

CLAIM  $\hat{x} = 2$  is the UNIQUE minimizer of  $F$ . Thus

$$\min_{\mathbb{R}} F = -4 \quad , \quad \arg\min_{\mathbb{R}} F = \{2\} \subseteq \mathbb{R}$$

Proof We need to show that

$$1) \quad F(x) \geq F(2) \quad \forall x \in \mathbb{R} \quad (\hat{x} = 2 \text{ is minimizer})$$

$$2) \quad F(x) > F(2) \quad \forall x \in \mathbb{R} \setminus \{2\} \quad (\hat{x} = 2 \text{ is unique min.})$$

$$1) \quad F(x) \geq F(2) \iff x^2 - 4x \geq -4 \iff (x-2)^2 \geq 0 \iff x=2$$

$$2) \quad F(x) > F(2) \iff (x-2)^2 > 0 \iff x \neq 2 \quad \square$$

### EXAMPLE 2

DIRECT METHOD: proving existence of a minimizer by general results

Ex:  $F: \mathbb{R} \rightarrow \mathbb{R}$  continuous and coercive, i.e.,

$$\lim_{|x| \rightarrow +\infty} F(x) = +\infty$$

Then  $\exists$  minimizer by Weierstrass Theorem

### Example 3

RELAXATION: This technique is relevant when a minimizer does not exist, e.g.,

$$\textcircled{*} \quad \min \{(x^2 - 2)^2 \mid x \in \mathbb{Q}\}$$

Solution of  $\textcircled{*}$  would be  $\hat{x} = \pm\sqrt{2}$  which is not in  $\mathbb{Q}$ . So in this case there is no minimum. But we are left with 2 questions

1) What is

$$\inf \{(x^2 - 2)^2 \mid x \in \mathbb{R}\} ?$$

2) If  $\{x_n\}$  is MINIMIZING SEQUENCE, i.e.

$$F(x_n) \rightarrow \inf \{F(x) \mid x \in \mathbb{R}\}, \quad F(x) = (x^2 - 2)^2$$

what can we say about accumulation points of  $\{x_n\}$ ?

Answer: 1) As we guessed, min over  $\mathbb{R}$  would be  $x^* = \pm\sqrt{2}$ , so

$$\inf \{(x^2 - 2)^2 \mid x \in \mathbb{Q}\} = F(\pm\sqrt{2}) = 0$$

2)  $x_n \rightarrow \sqrt{2}$  OR  $x_n \rightarrow -\sqrt{2}$  (up to subsequences)

Relaxation is useful to treat problems such as  $(*)$ . To ensure that a minimizer exists one could, for example,

- Extend  $F$  over some set  $\hat{X}$  with  $\hat{X} \supseteq X$  ( $\hat{X} = \mathbb{R}$  for  $(*)$ )
- Change  $F$  so that a minimizer is more likely to exist

EXAMPLE 4  $\Gamma$ -CONVERGENCE: We have a family of problems

$$\min \{F_n(x) \mid x \in X\}, \quad F_n: X \rightarrow \mathbb{R}, \quad n \in \mathbb{N}$$

What happens as  $n \rightarrow +\infty$ ? We hope to find  $F_\infty: X \rightarrow \mathbb{R}$  such that

$$1) \min \{F_n \mid x \in X\} \rightarrow \min \{F_\infty \mid x \in X\} \text{ as } n \rightarrow +\infty$$

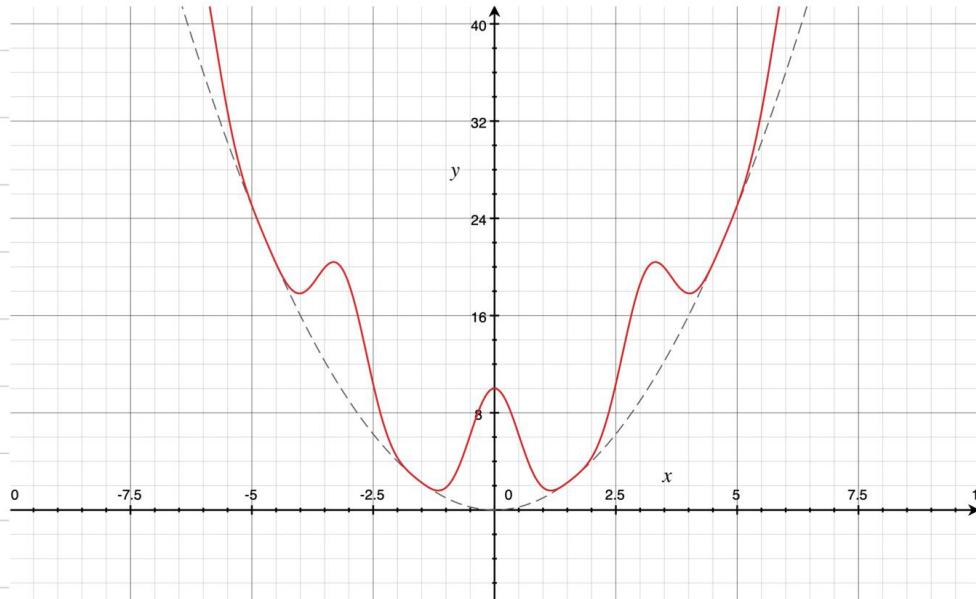
$$2) \text{If } x_n \in \arg \min \{F_n \mid x \in X\} \text{ then } x_n \rightarrow x_\infty \text{ with} \\ x_\infty \in \arg \min \{F_\infty \mid x \in X\}$$

$F_\infty$  is the  $\Gamma$ -limit of  $\{F_n\}$  as  $n \rightarrow +\infty$

For example consider

$$m_n = \min \{ F_n(x) \mid x \in \mathbb{R} \}, \quad F_n(x) = x^2 + n \cos^4 x$$

What is the limit of  $m_n$ ?



Dashed  $y = x^2$

Red  $F_n, n = 10$

- $F_n$  is sum of two positive terms
- $x^2$  small  $\Leftrightarrow x \approx 0$
- $n \cos^4 x$  small  $\Leftrightarrow \cos x \approx 0$

True when  $x = \pm \frac{\pi}{2}$

Indeed one has

1)  $m_n \rightarrow \left(\frac{\pi}{2}\right)^2$  as  $n \rightarrow +\infty$

2)  $\{x_n\}$  minimizing sequence converges (up to subsequences) to  $\pm \frac{\pi}{2}$ .

3) The  $\Gamma$ -limit is

$$F_\infty(x) = \begin{cases} x^2 & \text{if } \cos x = 0 \\ +\infty & \text{otherwise} \end{cases}$$

## INTEGRAL FUNCTIONALS

This course mainly focusses on integral functionals

$X = \text{some functions space}$ , e.g.,

$$X = C^k[a,b] = \{u: [a,b] \rightarrow \mathbb{R} \mid u \text{ k-times continuously differentiable}\}$$

and  $F: X \rightarrow \mathbb{R}$  is of the form

$$F(u) := \int_a^b L(x, u(x), u'(x), \dots, u^{(k)}(x)) dx$$

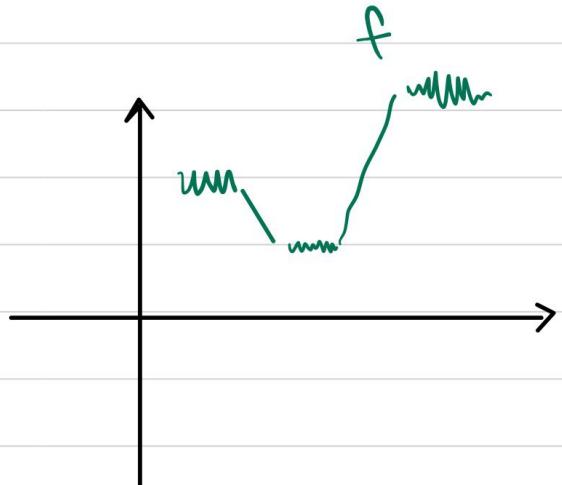
with  $L: [a,b] \times \mathbb{R}^k \rightarrow \mathbb{R}$  LAGRANGIAN

Typically  $L: [a,b] \times \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ ,  $L = L(x, s, p)$

### EXAMPLE 1 (DENOSING)

We receive a signal  $f: [0,1] \rightarrow \mathbb{R}$

which we want to denoise

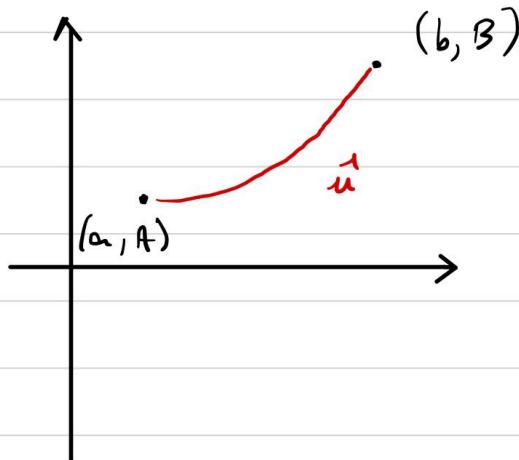


This task is achieved by solving

$$\hat{u} \in \operatorname{argmin} \left\{ \int_0^1 \dot{u}^2 + (u-f)^2 dx \mid u \in C^1[0,1] \right\}$$

- NOTE
- $\dot{u}^2$  penalizes oscillations
  - $u-f$  penalizes discrepancy from the noisy signal  $f$

Example 2 (Hanging Rope) Find the profile of a rope hanging at  $(a, A)$ ,  $(b, B)$



The energy of a profile  $u: [a, b] \rightarrow \mathbb{R}$  is modelled by

$$E(u) = \int_a^b u'^2 + u \, dx, \quad u(a)=A, u(b)=B$$

↑  
elastic energy      ↴ potential energy

- Note
- 1)  $u$  can be negative which lowers  $E$
  - 2) Due to boundary conditions, if  $u < 0$  then  $u' > 0 \Rightarrow E$  higher

The solution

$$\hat{u} \in \arg\min \left\{ E(u) \mid u \in C^1[a, b], u(a)=A, u(b)=B \right\}$$

will be a balance between (1) and (2)

### PROBLEMS WE WILL NOT TALK ABOUT

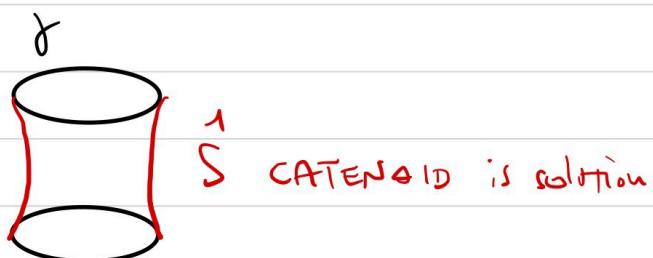
1) GEOMETRIC FUNCTIONALS:

- DIDO'S PROBLEM:  $\min \left\{ \text{Area}(\partial V) \mid V \subseteq \mathbb{R}^3, \text{Vol}(V)=1 \right\}$

Intuitively the sol is a sphere. However proving it when not requiring regularity on  $V$  requires advanced tools (Geometric Measure Theory)

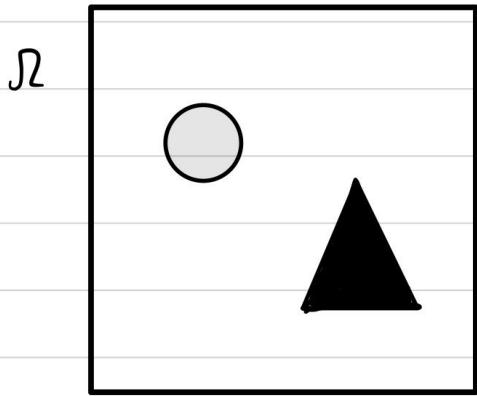
- PLATEAU'S PROBLEM : Given a collection of curves in  $\mathbb{R}^3$  find

$$\min \left\{ \text{Area}(S) \mid S \subseteq \mathbb{R}^3 \text{ surface}, \partial S = \gamma \right\}$$



Again this requires  
GMT

- 2) IMAGING FUNCTIONALS: used for tasks such as Denoising, Segmentation, reconstruction of medical data. Usually



$$u: \Omega \rightarrow \mathbb{R} , \quad \Omega \subseteq \mathbb{R}^2, \quad \mathbb{R}^3$$

$\mu$  encodes gray-scale value of pixels  
of a picture in the frame  $S^2$

Example : To segment the image at the left , i.e. find contours of shapes within it , one could minimize the MUMFORD - SHAH functional .

$$F(u, k) = \int_{\Omega \setminus k} |\nabla u|^2 dx + \int_{\Omega} |u - f|^2 dx + \text{Length}(k)$$

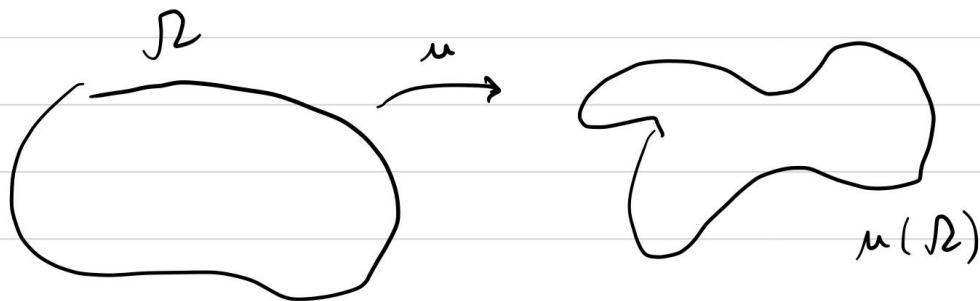
- $f$  is the original picture, generally noisy. Want to clean  $f$  and detect edges  $K$  within it
  - (I): enforces smoothness of  $u$  outside of  $K$  (we don't want to pay energy for the natural transitions)

- (II) : Enforces the clean image  $u$  to be close to the original  $f$
- (III) : Forces short contours

A solution is then

$$(u, k) \in \arg\min \{ F(u, k) \mid k \subseteq \bar{\Omega} \text{ compact}, u \in C^1(\Omega \setminus k) \}$$

3. VECTORIAL PROBLEMS : For example in materials science  
 $\Omega \subseteq \mathbb{R}^3$  represents the reference configuration  
of an elastic body,  $u: \Omega \rightarrow \mathbb{R}^3$  is  
a deformation



The elastic energy of the deformed configuration is

$$E(u) = \int_{\Omega} W(\nabla u) dx, \quad W: \mathbb{R}^{3 \times 3} \rightarrow [0, +\infty]$$

In this case the problem of minimizing  $E$  is vectorial, and the analysis requires advanced tools (quasi-convexity, etc)

An equilibrium configuration given boundary data  $g: \partial\Omega \rightarrow \mathbb{R}^3$  is

$$\min \{ E(u) \mid u \in C^1(\Omega; \mathbb{R}^3), u = g \text{ on } \partial\Omega \}$$

# BASIC FUNCTIONAL ANALYSIS (Revision)

REFERENCE: J. B. CONWAY  
"A COURSE IN FUNCTIONAL ANALYSIS"  
SECOND EDITION, SPRINGER, 1997

## METRIC SPACE

$X$  set,  $d: X \times X \rightarrow [0, +\infty)$ . We say that  $d$  is a METRIC over  $X$  if

- $d(x, y) = 0 \iff x = y$
- $d(x, y) = d(y, x)$ ,  $\forall x, y \in X$  (symmetric)
- $d(x, y) \leq d(x, z) + d(y, z)$ ,  $\forall x, y, z \in X$  (triangle inequality)

The pair  $(X, d)$  is called a Metric Space

## CONVERGENCE

For  $\{x_n\} \subseteq X$  we say that  $x_n \rightarrow x_0$  if

$$\forall \varepsilon > 0, \exists N_\varepsilon \in \mathbb{N} \text{ s.t. } d(x_n, x_0) < \varepsilon \text{ if } n \geq N_\varepsilon$$

## CAUCHY SEQUENCE

$\{x_n\} \subseteq X$  is a Cauchy sequence if

$$\forall \varepsilon > 0, \exists N_\varepsilon \in \mathbb{N} \text{ s.t. } d(x_n, x_m) < \varepsilon \text{ if } n, m \geq N_\varepsilon$$

## COMPLETENESS

A metric space  $(X, d)$  is complete if every Cauchy sequence  $\{x_n\} \subseteq X$  converges to some  $x_0 \in X$ .

## Topology generated by $d$

$(X, d)$  metric space. Define

$$\tau := \{ A \subseteq X \mid \forall x \in A, \exists \varepsilon > 0 \text{ s.t. } B_\varepsilon(x) \subseteq A \}$$

with  $B_\varepsilon(x) := \{ y \in X \mid d(x, y) < \varepsilon \}$ . Then  $\tau$  is a TOPOLOGY over  $X$ . The sets in  $\tau$  are called OPEN.  $A \subseteq X$  is closed if  $A^c := X - A$  is open.

## NOTATION

$(X, \tau)$  topological space,  $A \subseteq X$ . We denote by

- $\overset{\circ}{A}$  the INTERIOR of  $A$ :  $\overset{\circ}{A} = \bigcup \{ O \mid O \subseteq A, O \text{ open} \}$
- $\overline{A}$  the CLOSURE of  $A$ :  $\overline{A} = \bigcap \{ C \mid A \subseteq C, C \text{ closed} \}$

In other words :

- $\overset{\circ}{A}$  is the largest open set contained in  $A$
- $\bar{A}$  is the smallest closed set which contains  $A$

DENSITY  $(X, d)$  metric space.  $D \subseteq X$  is DENSE in  $X$  if  $\overline{D} = X$

SEPARABILITY  $(X, d)$  metric space is SEPARABLE if  $\exists$  a COUNTABLE set  $D \subseteq X$  which is dense, i.e.,  $\overline{D} = X$

LIMITS  $(X, d_X), (Y, d_Y)$  metric spaces,  $U \subseteq X$  open,  $F: U \rightarrow Y$ ,  $x_0 \in U$ . We say that  $F(x) \rightarrow L$  as  $x \rightarrow x_0$ , in symbols

$$\lim_{x \rightarrow x_0} F(x) = L ,$$

if  $\forall \varepsilon > 0$ ,  $\exists \delta > 0$  s.t.  $d_Y(F(x), L) < \varepsilon$  if  $d_X(x, x_0) < \delta$

CONTINUITY  $(X, d_X), (Y, d_Y)$  metric spaces,  $U \subseteq X$  open,  $F: U \rightarrow Y$ . We say that  $F$  is continuous at  $x_0 \in U$  if  $F(x) \rightarrow F(x_0)$  as  $x \rightarrow x_0$ .  $F$  is continuous in  $U$  if it is continuous  $\forall x_0 \in U$ .

NORMED SPACE  $X$  vector space over  $\mathbb{R}$ ,  $\| \cdot \|: X \rightarrow [0, +\infty)$ .

We say that  $\| \cdot \|$  is a norm over  $X$  if

- $\|x\| = 0 \iff x = 0$
- $\|\lambda x\| = |\lambda| \|x\|$ ,  $\forall \lambda \in \mathbb{R}, x \in X$  (1-homogeneous)
- $\|x+y\| \leq \|x\| + \|y\|$ ,  $\forall x, y \in X$  (Subadditive)

The pair  $(X, \| \cdot \|)$  is called normed space

REMARK  $(X, \| \cdot \|)$  normed space. Then  $d(x, y) = \|x - y\|$  is a metric over  $X$ . In particular  $X$  is a topological space with  $\tau$  induced by  $d$ . By convention all the topological notions in  $X$  are given WRT  $\tau$ .

BANACH SPACE  $(X, \|\cdot\|)$  normed space is BANACH if  $(X, d)$  with  $d(x, y) = \|x - y\|$  is complete.

LINEAR OPERATORS

$X, Y$  normed spaces,  $T: X \rightarrow Y$ . We say that

- $T$  is LINEAR if  $T(\lambda x + y) = \lambda T_x + T_y$ ,  $\forall \lambda \in \mathbb{R}, x, y \in X$
- $T$  is BOUNDED if

$$\sup_{\|x\|_X \leq 1} \|Tx\|_Y < +\infty$$

FACT Let  $T: X \rightarrow Y$  be linear. Then

$$T \text{ is continuous} \iff T \text{ is bounded}$$

NOTATION

$$\mathcal{L}(X, Y) := \{ T: X \rightarrow Y \mid T \text{ linear bounded} \}$$

$$X^* := \mathcal{L}(X, \mathbb{R}) \quad \text{DUAL SPACE of } X$$

REMARK

1)  $\mathcal{L}(X, Y)$  is a vector space over  $\mathbb{R}$ , with operations

$$(\alpha T_1 + T_2)(x) := \alpha T_1 x + T_2 x, \quad \forall \alpha \in \mathbb{R}, T_1, T_2 \in \mathcal{L}(X, Y)$$

2)  $\mathcal{L}(X, Y)$  is a normed space with norm

$$\|T\|_{\mathcal{L}(X, Y)} := \sup_{\|x\|_X \leq 1} \|Tx\|_Y$$

3) If  $Y$  is Banach then  $\mathcal{L}(X, Y)$  is Banach

4)  $X$  normed space  $\Rightarrow X^*$  Banach space

CONVERGENCES  $(X, \|\cdot\|)$  normed space  $\{x_n\} \subseteq X, x_0 \in X, \{\varphi_n\} \subseteq X^*$   
 $\varphi_0 \in X^*$

- 1)  $x_n \rightarrow x_0$  **STRONGLY** if  $\|x_n - x_0\|_X \rightarrow 0$  as  $n \rightarrow +\infty$
- 2)  $x_n \rightharpoonup x_0$  **WEAKLY** if  $\varphi(x_n) \rightarrow \varphi(x_0)$ ,  $\forall \varphi \in X^*$
- 3)  $\varphi_n \xrightarrow{*} \varphi_0$  **WEAKLY\*** if  $\varphi_n(x) \rightarrow \varphi_0(x)$ ,  $\forall x \in X$
- 4)  $\varphi_n \rightarrow \varphi_0$  **STRONGLY** if  $\|\varphi_n - \varphi_0\|_{X^*} \rightarrow 0$

NOTE •  $x_n \rightarrow x_0 \Rightarrow x_n \rightharpoonup x_0$   
• The reverse implication is not true. For example let

$$X = \ell^p := \left\{ (x_1, x_2, \dots, x_n, \dots) \mid \sum_{j=1}^{+\infty} |x_j|^p < +\infty \right\}$$

with  $1 < p < +\infty$ . Recall that  $(X, \|\cdot\|)$  is a normed space

with

$$\|x\| := \left( \sum_{j=1}^{+\infty} |x_j|^p \right)^{1/p}$$

Let  $e_j := (0, \dots, 0, \underset{j\text{-th position}}{1}, 0, \dots)$ . Then  $e_j \rightarrow 0$  but

$$\|e_j\| = 1 \neq 0.$$

DEFINITION  $(X, \|\cdot\|)$  normed space,  $K \subseteq X$ ,  $\tilde{K} \subseteq X^*$

- 1)  $K$  is **COMPACT** if  $\{x_n\} \subseteq K$ ,  $\exists x_0 \in K$  s.t.  $x_{n_k} \rightarrow x_0$  along some subsequence
- 2)  $K$  is **SEQUENTIALLY WEAKLY COMPACT** if  $\{x_n\} \subseteq K$ ,  $\exists x_0 \in K$  s.t.  $x_{n_k} \rightharpoonup x_0$  along some subsequence
- 3)  $\tilde{K}$  is **SEQUENTIALLY WEAKLY\* COMPACT** if  $\{\varphi_n\} \subseteq \tilde{K}$ ,  $\exists \varphi_0 \in \tilde{K}$  s.t.  $\varphi_{n_k} \xrightarrow{*} \varphi_0$  along some subsequence.

## WARNING

If  $(X, \tau)$  is a topological space then  $K \subseteq X$  is compact if any OPEN COVER of  $K$  admits a FINITE SUBCOVER.

If the topology  $\tau$  is metrizable (e.g. metric or normed spaces) then SEQUENTIAL COMPACTNESS is equivalent to COMPACTNESS.

However, if  $X$  is normed space, then the weak topology on  $X$  and weak\* topology on  $X^*$  are NOT metrizable in general. Thus, in general WEAK (WEAK\*) COMPACTNESS and WEAK (WEAK\*) SEQUENTIAL COMPACTNESS are not equivalent. With additional assumptions, however, they are the same:

- 1) If  $X$  is Banach then WEAK SEQUENTIAL COMPACTNESS and WEAK COMPACTNESS are equivalent  
( EBERLEIN - SMULIAN THEOREM )
- 2) If  $X$  is a SEPARABLE BANACH space then WEAK\* SEQUENTIAL COMPACTNESS and WEAK\* COMPACTNESS are equivalent

THEOREM (BANACH - ALAOGLU)  $(X, \|\cdot\|)$  normed space. Denote

by  $B := \{ \varphi \in X^* \mid \|\varphi\| \leq 1 \}$  the closed unit ball of  $X^*$ :

- 1) Then  $B$  is WEAKLY\* COMPACT
- 2) If in addition  $X$  is BANACH and SEPARABLE then  $B$  is also SEQUENTIALLY WEAKLY\* COMPACT

There is a corollary of Banach-Alaoglu concerning the weak compactness of the unit ball of  $X$ . For that we need the following definition

## REFLEXIVITY

$(X, \|\cdot\|)$  normed space. Consider  $X^*$  and its dual w.r.t. to the strong norm of  $X^*$ , i.e.,  $X^{**} := (X^*, \|\cdot\|_{X^*})^*$

Define the **CANONICAL EMBEDDING**

$$J: X \rightarrow X^{**} \text{ s.t. } J(x)(\varphi) := \varphi(x), \quad x \in X, \varphi \in X^*$$

We have  $\|J(x)\|_{X^{**}} = \|x\|_X$ . We say that  $X$  is **REFLEXIVE** if  $J$  is surjective, i.e., if

$$X^{**} = \{J(x), x \in X\}$$

## COROLLARY (of BANACH-ALAOGLU)

$(X, \|\cdot\|)$  normed space. Define  $B := \{x \in X \mid \|x\| \leq 1\}$ .

- 1) If  $X$  is reflexive then  $B$  is WEAKLY COMPACT
- 2) If  $X$  is reflexive and Banach then  $B$  is WEAKLY SEQUENTIALLY COMPACT

As a consequence of the PRINCIPLE OF UNIFORM BOUNDEDNESS (PUB)  
(See book of Conway), we have.

## PROPOSITION

$(X, \|\cdot\|)$  Banach space

- 1) If  $\{x_n\} \subseteq X$  is s.t.  $x_n \rightharpoonup x_0$  then  $\sup_n \|x_n\| < +\infty$
- 2) If  $\{\varphi_n\} \subseteq X^*$  is s.t.  $\varphi_n \not\rightharpoonup \varphi_0$  then  $\sup_n \|\varphi_n\|_{X^*} < +\infty$

Another important notion needed throughout the course is the one of lower semicontinuity.

## DEFINITION

$(X, d)$  metric space,  $F: X \rightarrow \mathbb{R}$ . We say that  $F$  is LOWER SEMICONTINUOUS at  $x_0 \in X$  if

$$F(x_0) \leq \liminf_{n \rightarrow +\infty} F(x_n),$$

for all  $\{x_n\} \subseteq X$  s.t.  $x_n \rightarrow x_0$ .

## DEFINITION $(X, \|\cdot\|)$ normed space, $F: X \rightarrow \mathbb{R}$ , $G: X^* \rightarrow \mathbb{R}$ .

1)  $F$  is (SEQUENTIALLY) WEAKLY LOWER SEMICONTINUOUS at  $x_0 \in X$  if

$$F(x_0) \leq \liminf_{n \rightarrow +\infty} F(x_n)$$

for all  $\{x_n\} \subseteq X$  s.t.  $x_n \rightharpoonup x_0$ .

2)  $G$  is (SEQUENTIALLY) WEAKLY\* LOWER SEMICONTINUOUS at  $p_0 \in X^*$  if

$$G(p_0) \leq \liminf_{n \rightarrow +\infty} G(p_n)$$

for all  $\{p_n\} \subseteq X^*$  s.t.  $p_n \rightharpoonup p_0$ .

## PROPOSITION

$(X, \|\cdot\|)$  normed space. Then

1) The norm  $\|\cdot\|$  is WEAKLY SEQUENTIALLY LOWER SEMICONTINUOUS, i.e.,

$$x_n \rightarrow x_0 \Rightarrow \|x_0\| \leq \liminf_{n \rightarrow +\infty} \|x_n\|$$

2) The norm  $\|\cdot\|_{X^*}$  is WEAKLY\* SEQUENTIALLY LOWER SEMICONTINUOUS, i.e.,

$$p_n \rightharpoonup p_0 \Rightarrow \|p_0\|_{X^*} \leq \liminf_{n \rightarrow +\infty} \|p_n\|_{X^*}$$