

$(\Omega, \mathcal{F}, \mathbb{P})$ probability space

X random variable $X: \Omega \rightarrow \mathbb{R}$ measurable funct.

$$\forall A \in \mathcal{B}(\mathbb{R}) \quad X^{-1}(A) = \{\omega \in \Omega \mid X(\omega) \in A\} \in \mathcal{F}$$

X induces on \mathbb{R} a Borel measure \mathbb{P}_X (a Borel measure on \mathbb{R})

$$\mathbb{P}_X(A) = \mathbb{P}\{\omega \in \Omega \mid X(\omega) \in A\} = \mathbb{P}(X^{-1}(A))$$

\mathbb{P}_X is a FINITE BOREL MEASURE $(\mathbb{P}_X(\mathbb{R}) = \mathbb{P}(X^{-1}(\mathbb{R})) = \mathbb{P}(\Omega) = 1)$

it is associated to a cumulative distribution function

$$G_X(x) = \mathbb{P}_X(-\infty, x] = \mathbb{P}\{\omega \mid X(\omega) \leq x\}$$

right continuous $\lim_{x \rightarrow a^+} G(x) = G(a)$, NON DECREASING

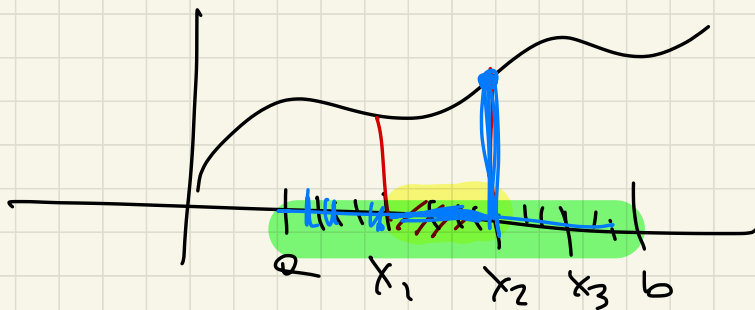
We may define the notion of integral with respect to any μ Borel σ -finite or finite measure

$f: \mathbb{R} \rightarrow \mathbb{R}$ continuous,

$$\int_{[a,b]} f(x) dP_x(x) = \sup \left[\sum_{n=1}^{k+1} f(x_n) [G_x(x_n) - G_x(x_{n-1})] \right], \left. \begin{array}{l} \text{for all possible} \\ \text{subdivision} \\ x_0 = a < x_1 < \dots < x_{k+1} = b \end{array} \right\}$$

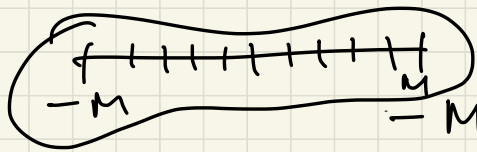
take a subdivision of (a, b) $x_0 = a < x_1 < x_2 < \dots < x_k < x_{k+1} = b$

$$\begin{aligned} & \sum_{n=1}^{k+1} f(x_n) \cdot P_x(x_{n-1}, x_n) \\ &= \sum_{n=1}^{k+1} f(x_n) \cdot [G_x(x_n) - G_x(x_{n-1})] \end{aligned}$$



$$\int_{\mathbb{R}} f(x) dP_X(x) = \sum_{(-M, M]} \sum_{k=1}^{n+1} f(x_k) [G_X(x_k) - G_X(x_{k-1})],$$

covering all possible



$$-M = x_0 < x_1 < x_2 < \dots < x_{k+1} = M$$

$$\forall i \neq j \quad |x_i - x_{i-1}| = |x_j - x_{j-1}|$$

Def X is an absolutely continuous variable

for a.e. x

$$G'_X(x) = g(x)$$

$g(x) \geq 0$ density

$$\int_{\mathbb{R}} f(x) dP_X(x) = \int_{\mathbb{R}} f(x) g(x) dx$$

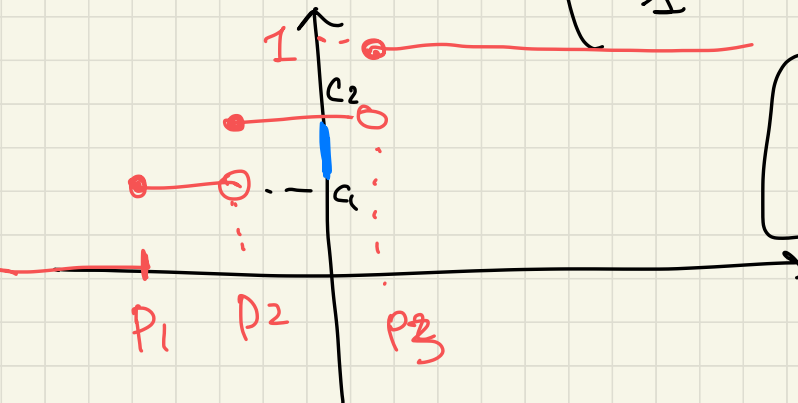
If X is a discrete random variable

\downarrow X takes values only on a discrete set

$$G_X(x) = \begin{cases} 0 = c_0 & x < p_1 \\ c_1 & p_1 \leq x < p_2 \\ c_2 & p_2 \leq x < p_3 \\ c_3 & \vdots \\ \vdots & \vdots \\ 1 & p_m \leq x < +\infty \end{cases} \quad \{p_1, \dots, p_m\}$$

$$0 < c_1 < c_2 < \dots < 1$$

$$P_X(x) = \sum_{i=1}^m (c_i - c_{i-1}) \delta_{p_i}(x)$$



$$\int_{\mathbb{R}} f(x) dP_X(x) = \sum_{i=1}^m f(p_i) \cdot (c_i - c_{i-1})$$

e.g. $f(x) = x^k$ $k \geq 1$ cont.

$$\int_{\mathbb{R}} x^k dP_x(x) = \mathbb{E}(X^k)$$

↑
if it exists

mean of X random variable

$$\mathbb{E}(X) := \int_{\mathbb{R}} x dP_x(x)$$

k -moment of X random variable is

$$\mathbb{E}(|X|^k) := \int_{\mathbb{R}} |x|^k dP_x(x)$$

$$\mathbb{E}(X^k) < +\infty \iff \mathbb{E}(|X|^k) < +\infty$$

It is not completely obvious but it is true

that

$$\text{if } \int_{\mathbb{R}} x^k dP_x(x) < +\infty$$

then

$$\int_{\mathbb{R}} |x|^k dP_x(x) < +\infty$$

and also the vice versa is true.

$M =$ space of all random variables on $(\Omega, \mathcal{F}, \mathbb{P})$

↓

$X: \Omega \rightarrow \mathbb{R}$ meas.

$$M_k = \left\{ X \in M, X \text{ random variable} \right. \\ \left. \text{such that } \int_{\mathbb{R}} |x|^k dP_x(x) < +\infty \right\}$$

$M_1 = \{ \text{random variables which have finite mean} \}$

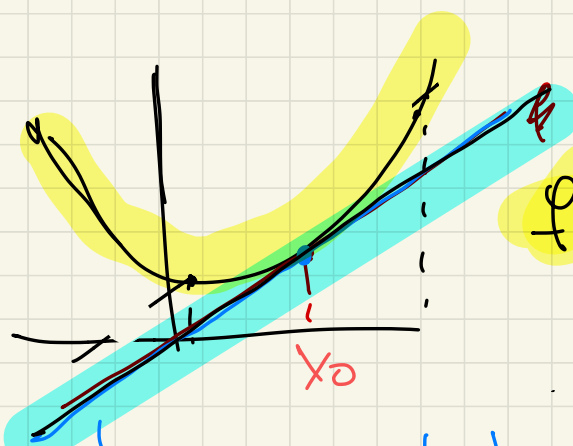
$M_2 = \{ \text{random variables which have finite } E(X^2) = \int_{\mathbb{R}} x^2 dP_X(x) < +\infty \}$

"Jensen inequality": if $f(x) : \mathbb{R} \rightarrow \mathbb{R}$ convex
(smooth and convex $f''(x) \geq 0 \forall x$)

$$\int_{\mathbb{R}} f(x) dP_X(x) \geq f \left[\int_{\mathbb{R}} x dP_X(x) \right]$$

$\forall X$ random variable

Proof, we have to use that $\int_{\mathbb{R}} dP_x(z) = 1$.



f convex and smooth.

$$f(x) \geq \underline{f(x_0)} + \underline{f'(x_0)} \cdot (x - x_0)$$

↓ line tangent to the graph of f passing in the point $(x_0, f(x_0))$ has equation

$$y = f(x_0) + f'(x_0)(x - x_0)$$

$$x_0 = \int_{\mathbb{R}} z dP_x(z) \quad \forall x$$

$$f(x) \geq f\left(\int_{\mathbb{R}} z dP_x(z)\right) + f'\left(\int_{\mathbb{R}} z dP_x(z)\right) \left[x - \int_{\mathbb{R}} z dP_x(z)\right]$$

$$\int_{\mathbb{R}} f(x) dP_x(x) \geq \underbrace{\int_{\mathbb{R}} f \left(\int_{\mathbb{R}} z dP_x(z) \right)}_{\text{CONST}} + \underbrace{\int_{\mathbb{R}} f' \left(\int_{\mathbb{R}} z dP_x(z) \right) \left(x - \int_{\mathbb{R}} z dP_x(z) \right)}_{dP_x(x)}$$

↓

$$\int_{\mathbb{R}} f(x) dP_x(x) \geq f \left(\int_{\mathbb{R}} z dP_x(z) \right) \int_{\mathbb{R}} dP_x(x) +$$

$$+ f' \left(\int_{\mathbb{R}} z dP_x(z) \right) \left[\int_{\mathbb{R}} z dP_x(x) - \int_{\mathbb{R}} 1 dP_x(x) \cdot \int_{\mathbb{R}} z dP_x(z) \right] = 0$$

Take $f(x) = |x|^k$ $k > 1$ f convex

$$\int_{\mathbb{R}} f(x) dP_x(x) \geq f\left(\int_{\mathbb{R}} x dP_x(x)\right)$$

$$+\infty > \int_{\mathbb{R}} |x|^k dP_x(x) \geq \left| \int_{\mathbb{R}} x dP_x(x) \right|^k$$

$$\text{if } \int_{\mathbb{R}} |x|^k dP_x(x) < +\infty \implies \int_{\mathbb{R}} x dP_x(x) < +\infty$$

$M^k \subset M^1$ (if X has finite k -moment then it has finite mean) -

CONVEXITY OF f

Jensen

$$f(x) \geq f(x_0) + f'(x_0)(x - x_0)$$

$\forall x$

↓ apply the convexity inequality to

$$f(x^n) \geq f(x_0) + f'(x_0)(x^n - x_0)$$

$(n \geq 1)$

$$x_0 = \int_{\mathbb{R}} z^n dP_x(z) \text{ constant}$$

$$\int_{\mathbb{R}} f(x^n) dP_x(x) \geq f\left(\int_{\mathbb{R}} z^n dP_x(z)\right) +$$

$$+ f'\left(\int_{\mathbb{R}} z^n dP_x(z)\right) \cdot \left[\int_{\mathbb{R}} x^n dP_x(x) - \int_{\mathbb{R}} z^n dP_x(z)\right]$$

$$\int_{\mathbb{R}} f(x^n) dP_x(x) \geq f\left(\int_{\mathbb{R}} z^n dP_x(z)\right)$$

(same argument gives that $\forall \varphi: \mathbb{R} \rightarrow \mathbb{R}$

$$\int_{\mathbb{R}} \underline{\underline{f}}(\varphi(x)) dP_x(x) \geq \underline{\underline{f}} \left[\int_{\mathbb{R}} \varphi(z) dP_x(z) \right]$$

f convex if $\int_{\mathbb{R}} \varphi(z) dP_x(z) < +\infty$

(in the convexity inequality I substitute x with $\varphi(x)$ and x_0 with $\int_{\mathbb{R}} \varphi(z) dP_x(z)$)

$1 < \underline{\underline{m}} < k$ $\varphi(x) = |x|^{\frac{k}{m}}$ convex $\varphi(x) = x^n$

$$+\infty > \int_{\mathbb{R}} |x|^k dP_x(x) = \int_{\mathbb{R}} \underline{\underline{f}}(\varphi(x)) dP_x(x) \geq \left[\int_{\mathbb{R}} x^n dP_x(x) \right]^{\frac{k}{n}}$$

$$E|X|^k \geq |E(X)^n|^{\frac{k}{n}} \geq \underline{\underline{[E|X|^n]}}^{\frac{k}{n}}$$

$\forall k > n$

(DIS. DI JENSEN)

$$\left[\int_{\mathbb{R}} |x|^k dP_x(x) \right]^{\frac{1}{k}} \geq \left[\int_{\mathbb{R}} |x|^n dP_x(x) \right]^{\frac{1}{n}}$$

$$[E|X|^k]^{\frac{1}{k}} \geq [E|X|^n]^{\frac{1}{n}}$$

$1 \leq n < k$

We want to put some "differential" structure on these spaces (which are infinite dimensional \rightarrow the "points" are the seedone variables.)

X be an (infinite dimensional) VECTORIAL SPACE on \mathbb{R} ($x_1, x_2 \in X$ $\underline{\lambda}x_1 + \underline{\mu}x_2 \in X$ $\forall \lambda, \mu \in \mathbb{R}$).

X is a METRIC SPACE if \exists a function

d (distance) : $X \times X \rightarrow [0, +\infty)$

- $d(x_1, x_2) \geq 0 \quad \forall x_1, x_2 \in X$
 - $d(x_1, x_2) = 0 \iff x_1 = x_2$
 - $d(x_1, x_2) = d(x_2, x_1) \mapsto$ SYMMETRY COND.
 - $d(x_1, x_2) \leq d(x_1, x_3) + d(x_3, x_2)$
(TRIANGULAR INEQUALITY)
-

We say that the distance is associated to a

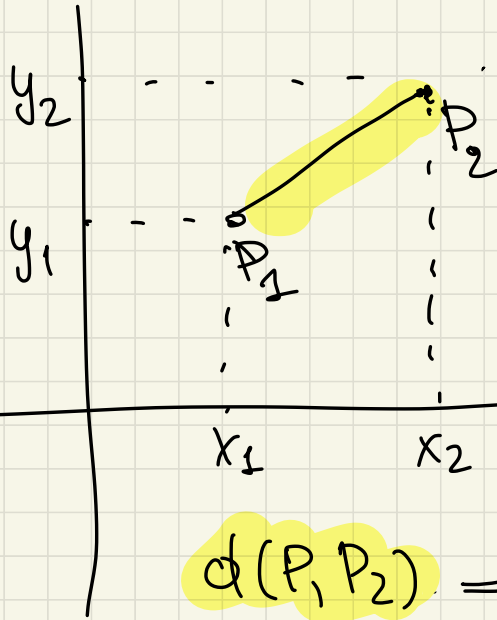
NORM iff $d(x_1, x_2) = \|x_1 - x_2\|$

$\|\cdot\| : X \rightarrow \mathbb{R}$

NORM :

- 1) $\|x\| \geq 0 \quad \|x\| = 0 \iff x = 0$
- 2) $\|\lambda x\| = |\lambda| \|x\| \quad \forall \lambda \in \mathbb{R}$.
- 3) $\|x_1 + x_2\| \leq \|x_1\| + \|x_2\|$.

$$P_1 = (x_1, y_1)$$
$$P_2 = (x_2, y_2)$$



\mathbb{R}^2
distance = length of
the segment (P_1, P_2)

$$d(P_1, P_2) = \sqrt{|x_1 - x_2|^2 + |y_1 - y_2|^2}$$

$$|P_1 - P_2| = \rightarrow$$

Def X vectorial space on \mathbb{R}

is a NORMED SPACE if there exists

or norm $\|\cdot\| : X \rightarrow [0, +\infty)$

1) $\|x\| = 0 \Leftrightarrow x = 0 \quad \|x\| \geq 0 \quad \forall x$

2) $\|\lambda x\| = |\lambda| \|x\| \quad \lambda \in \mathbb{R}$

$\|x_1 - x_2\| = \|x_2 - x_1\|$

3) $\|x_1 + x_2\| \leq \|x_1\| + \|x_2\|$

the norm is associated to a distance

$d(x_1, x_2) = \|x_1 - x_2\|$

$y_1 - y_3 = x_1 \quad x_2 = y_3 - y_2$

$\|x_1 + x_2\| = \|y_1 - y_3 + y_3 - y_2\| = \|y_1 - y_2\| \leq \|y_1 - y_3\| + \|y_3 - y_2\|$

(Ex) $X = \{ f : \underline{[0, 1]} \rightarrow \mathbb{R} \text{ CONTINUOUS} \}$

$f_1, f_2 \in X \Rightarrow \lambda f_1 + \mu f_2$ is continuous
 $: [0, 1] \rightarrow \mathbb{R}$

$\lambda, \mu \in \mathbb{R}$

NORM (L^∞ NORM)

$\|f\|_\infty = \max_{x \in [0, 1]} |f(x)|$

$\|f\|_\infty \geq 0 \quad \forall f$
 $\|f\|_\infty = 0 \Leftrightarrow \max_{x \in [0, 1]} |f(x)| = 0$
 $\|\lambda f\|_\infty = |\lambda| \|f\|_\infty$

$d_\infty(f, g) = \max_{x \in [0, 1]} |f(x) - g(x)|$

$\|f + g\|_\infty = \max_x |f(x) + g(x)| \leq$
 $\leq \max_x [|f(x)| + |g(x)|]$
 $\leq \max_x |f(x)| + \max_x |g(x)|$

$$\underline{\|f\|_1} = \int_0^1 |f(x)| dx \quad \text{it is also a } \underline{\text{NORM on } X}$$

$$\|f\|_1 \geq 0 \quad \int_0^1 |f(x)| dx = 0 \Leftrightarrow f(x) = 0 \quad \forall x$$

$$\|\lambda f\|_1 = \int_0^1 |\lambda f(x)| dx = |\lambda| \int_0^1 |f(x)| dx = |\lambda| \|f\|_1$$

$$\|f+g\|_1 = \int_0^1 |f(x)+g(x)| dx \leq \int_0^1 |f(x)| + |g(x)| dx \leq \int_0^1 |f(x)| dx + \int_0^1 |g(x)| dx$$

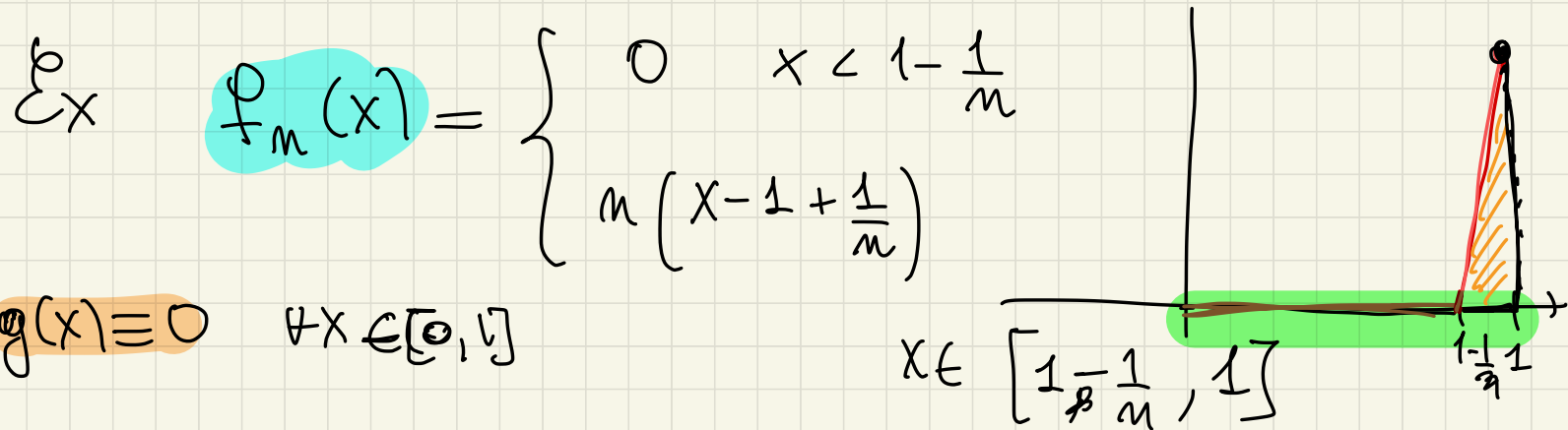
$$d_1(f, g) = \int_0^1 |f(x) - g(x)| dx$$

$$\underline{\underline{d_\infty(f, g) \leq \delta}} \iff |f(x) - g(x)| \leq \delta \quad \forall x$$

$$g(x) - \delta \leq f(x) \leq g(x) + \delta \quad \forall x \in [0, 1]$$

$$d_1(f, g) \leq \delta \iff \int_0^1 |f(x) - g(x)| dx \leq \delta$$

$$\int_0^1 g(x) dx - \delta \leq \int_0^1 f(x) dx \leq \int_0^1 g(x) dx + \delta$$



$$\begin{aligned}
 d_1(f_m, 0) &= \int_0^1 |f_m(x) - 0| dx = \\
 &= \int_0^1 m(x - 1 + \frac{1}{m}) dx = \frac{1}{2m} < 1
 \end{aligned}$$

$$d_\infty(f_m, 0) = \max_{x \in [0, 1]} |f_m(x) - 0| = 1 \quad \forall m$$

$$\lim_{n \rightarrow +\infty} d_1(f_n, 0) = 0$$

$$\lim_{n \rightarrow +\infty} d_\infty(f_n, 0) = 1$$

f_n is "going" to the constant function 0 in the sense of the d_1 distance, but not in the sense of the d_∞ distance.

If X is a NORMED SPACE $X, \|\cdot\|$ and $x_n \in X \quad \forall n$

and $x \in X$ we say that

$$\lim_{n \rightarrow +\infty} x_n = x$$

\iff

$$\lim_{n \rightarrow +\infty} \|x_n - x\| = 0$$

$x_n \rightarrow x$ iff DISTANCE BETWEEN x_n, x GOES TO 0.

Definition, Let X be a vectorial space on \mathbb{R}
and a norm $\|\cdot\|$ on it.

We say that the normed space $(X, \|\cdot\|)$

IS COMPLETE if

$\forall (x_n)$ sequence in X such that

$$\lim_{n, m \rightarrow +\infty} \|x_n - x_m\| = 0 \quad (\text{such sequences are called CAUCHY SEQUENCES})$$

it holds that there exists $x \in X$ such that

$$\lim_n \|x_n - x\| = 0$$

(so Cauchy sequences are converging to some element in the space X).

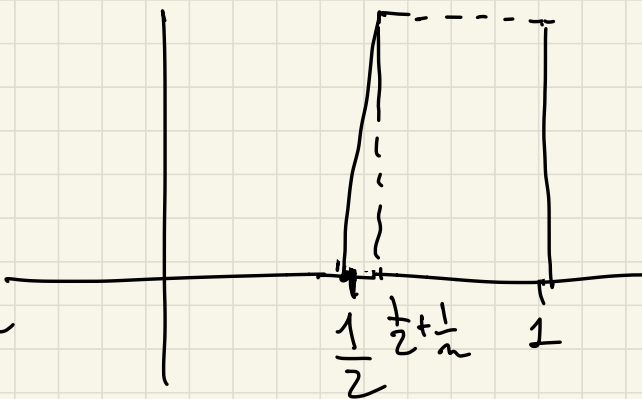
A normed complete space is called a BANACH SPACE

^{obs} Not all the normed spaces are complete.

$$X = \{ f : [0, 1] \rightarrow \mathbb{R} \text{ continuous} \}$$

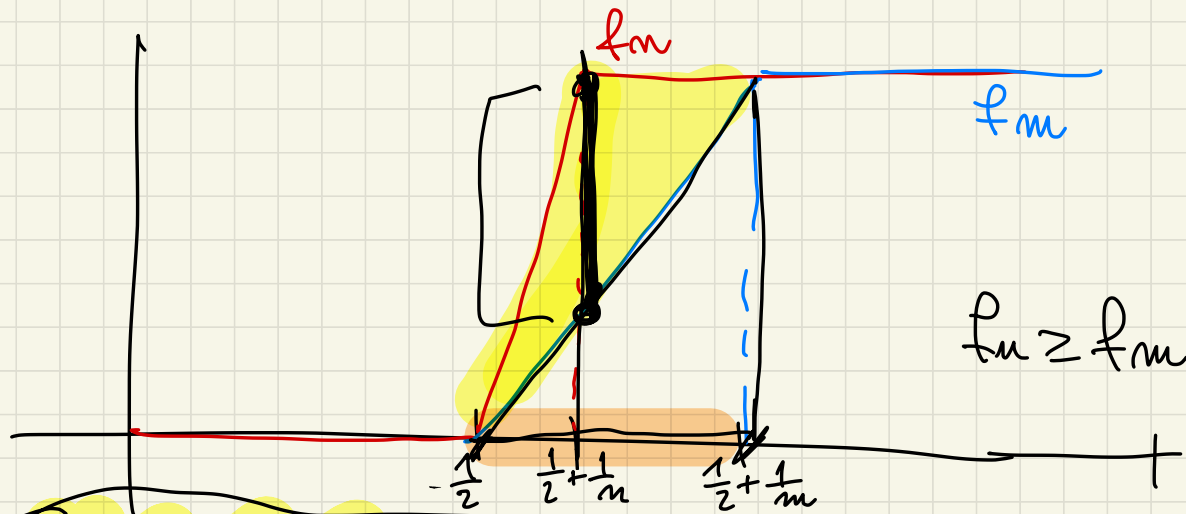
$$\|f\|_1 = \int_0^1 |f(x)| dx$$

$$f_n(x) = \begin{cases} 0 & x \leq \frac{1}{2} \\ n(x - \frac{1}{2}) & \frac{1}{2} < x < \frac{1}{2} + \frac{1}{n} \\ 1 & x \geq \frac{1}{2} + \frac{1}{n} \end{cases}$$



is a Cauchy sequence

$$\int_0^1 |f_n(x) - f_m(x)| dx \rightarrow 0$$



$$m > n$$

$$\frac{1}{2} + \frac{1}{n} < \frac{1}{2} + \frac{1}{m}$$

$$f_n \geq f_m$$

$$\int_0^1 |f_n(x) - f_m(x)| dx = \int_{\frac{1}{2}}^{\frac{1}{2} + \frac{1}{n}} m(x - \frac{1}{2}) - n(x - \frac{1}{2}) dx +$$

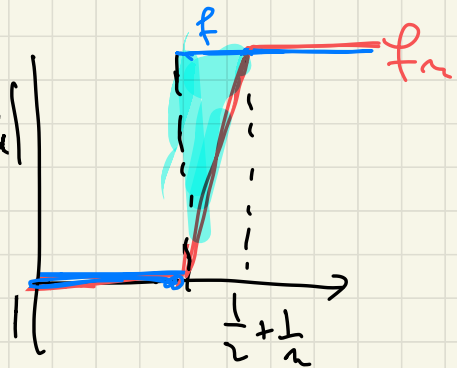
$$+ \int_{\frac{1}{2} + \frac{1}{n}}^1 1 - m(x - \frac{1}{2}) dx = \left[\frac{1}{2n} + \frac{1}{n} - \frac{1}{n} \right] - \frac{1}{2n} = \frac{1}{2n} - \frac{1}{2n} =$$

$$= \frac{1}{2} \left[\frac{1}{m} - \frac{1}{n} \right] \rightarrow 0$$

as $m, n \rightarrow +\infty$.

f_n is Cauchy $\|f_n - f_m\|_1 = \frac{1}{2} \left| \frac{1}{n} - \frac{1}{m} \right|$

Nevertheless f_n is NOT CONVERGING in $\|\cdot\|_1$ to a function f in X .



We claim $f_n \rightarrow f = \begin{cases} 0 & x < \frac{1}{2} \\ 1 & x \geq \frac{1}{2} \end{cases}$ in $\|\cdot\|_1$

$$d(f_n, f) = \|f_n - f\|_1 = \int_0^1 |f_n(x) - f(x)| dx = \int_{\frac{1}{2}}^{\frac{1}{2} + \frac{1}{n}} 1 - n(x - \frac{1}{2}) dx = \frac{1}{2n} \rightarrow 0 \text{ as } n \rightarrow \infty$$

$f \notin X$ SINCE f IS NOT CONTINUOUS!

$(X, \|\cdot\|_1)$ IS NOT COMPLETE

Observe that f_n is NOT CAUCHY WITH RESPECT TO $\|\cdot\|_\infty$!

$n > m$

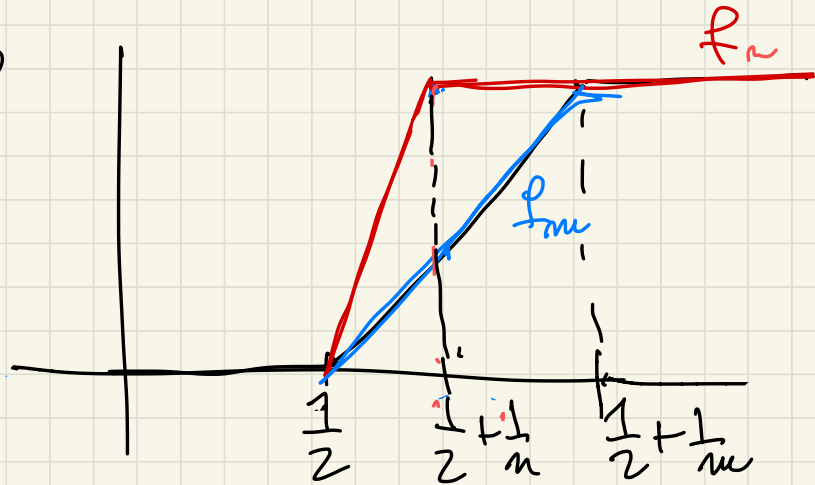
$$\|f_n - f_m\|_\infty = \sup_{[0,1]} |f_n(x) - f_m(x)| \geq f_n\left(\frac{1}{2} + \frac{1}{n}\right) - f_m\left(\frac{1}{2} + \frac{1}{n}\right)$$

$$= 1 - \frac{m}{n} = \frac{n-m}{n} \not\rightarrow 0$$

$m < n$

$$\lim_{n, m \rightarrow +\infty} \|f_n - f_m\|_\infty =$$

$$= \lim_{n, m \rightarrow +\infty} \frac{n-m}{n} \not\rightarrow 0$$



(take $m = \frac{n}{2}$ $\|f_n - f_{\frac{n}{2}}\|_\infty = \frac{1}{2} \quad \forall n$

$m = \frac{n}{3}$ $\|f_n - f_{\frac{n}{3}}\|_\infty = \frac{2}{3} \quad \forall n$

One can prove that

$(X = \{f: [0,1] \rightarrow \mathbb{R} \text{ continuously}, \| \cdot \|_\infty\})$

is COMPLETE.

$$f_n \text{ Cauchy } \|f_n - f_m\|_\infty = \sup_{x \in [0,1]} |f_n(x) - f_m(x)| \rightarrow 0$$

$$\text{so } \forall x \in [0,1] \quad |f_n(x) - f_m(x)| \rightarrow 0 \Rightarrow \exists \lim_n f_n(x) = f(x)$$

One can also prove that f is continuous.