

μ Borel σ -finite measure on $\mathbb{R}$

$$\mu = \mu_{ac} + \mu_s$$

$$\mu_{ac} \ll \mathcal{L}$$

$$\mu_s \perp \mathcal{L}$$

$\mu_{ac} \rightarrow f_\mu$ density $f_\mu \geq 0$

$$\forall B \in \mathcal{B} \quad \mu_{ac}(B) = \int_B f_\mu(x) dx = \int_{B \setminus N} f_\mu(x) dx$$

$\mu \leftrightarrow F$ cumulative distribution function
 $f_\mu(x) = F'(x)$ on

(since F is monotone $\rightarrow F$ is differentiable for almost every $x \in \mathbb{R}$ that is F is differentiable

$$\forall x \in \mathbb{R} \setminus N \quad |N| = 0$$

F cumulative distribution function
↓
associated to μ

Jump set of $F = \{a \in \mathbb{R} \mid F(a) \neq \lim_{y \rightarrow a^-} F(y)\} = J$
(theorem: $|J| = 0$ J is at most COUNTABLE)

$$F_j(x) = \sum_{\substack{a \in J \\ a \leq x}} (F(a) - F(a^-)) = \sum_{\substack{a \in J \\ a \leq x}} \mu(\{a\}) \quad \left| \begin{array}{l} \text{PIECEWISE} \\ \text{CONSTANT} \end{array} \right.$$

$F - F_j$ is continuous

$$F_j' = 0 \text{ a.e.}$$

$$f_\mu = (F - F_j)' = F' - \cancel{F_j'}$$

$\tilde{\mu}_s$ (the jump singular part of μ) is associated with the cumulative distribution function F_j

μ_{ac} has density given by $f_\mu = (F - F_j)'$
 $= F'$

$$\begin{aligned} F_{ac}(x) &= \int_{-\infty}^x f_\mu(y) dy \quad (\mu \text{ is finite}) \\ &= \mu_{ac}(-\infty, x] \end{aligned}$$

In general $F_{ac}(x) \leq F(x) - F_j(x)$

It is not true in general that
 F continuous

$$F(x) = \int_{-\infty}^x F'(y) dy$$

$$f_{\mu} = F'(x) \quad F_{ac}(x) = \int_{-\infty}^x f_{\mu}(y) dy \leq \underline{\underline{F(x) - F_j(x)}}$$

(there are continuous functions which are differentiable e. g. such that they are not the integral of their derivative).

Devil staircase (Cantor function)

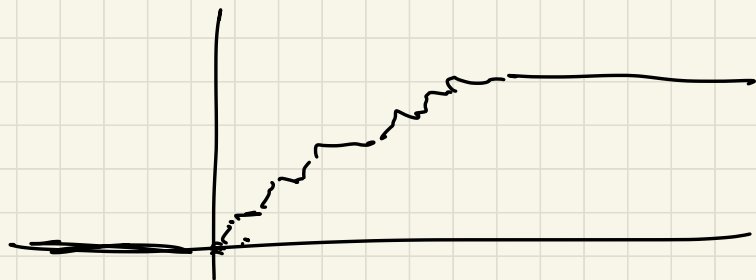
$$F: \mathbb{R} \rightarrow \mathbb{R}$$

$$F(x) = 0 \quad x \leq 0$$

$$F(x) = 1 \quad x \geq 1$$

F monotone non decreasing, continuous
(NO JUMPS)

$$F'(x) = 0 \quad \text{for a.e. } x$$



Cumulative distribution function of a measure μ
SINGULAR w.r.t. $\mathbb{L}$.

$$\begin{array}{c} \mathbb{R}^2 \\ // \\ \mathbb{R} \times \mathbb{R} \end{array}$$

σ -algebra of Borel sets

$$\mathcal{B}(\mathbb{R}^2) = \mathcal{B}(\mathbb{R}) \times \mathcal{B}(\mathbb{R})$$

$$A \in \mathcal{B}(\mathbb{R}^2)$$

$$\begin{array}{c} \swarrow \\ A = \underline{B \times C} \\ B \in \mathcal{B}(\mathbb{R}) \\ C \in \mathcal{B}(\mathbb{R}) \end{array}$$

μ Borel measure on $\mathbb{R}^2$
 $(\mathbb{R}^2, \mathcal{B}(\mathbb{R}^2), \mu)$

$$\mu : \mathcal{B}(\mathbb{R}^2) \rightarrow [0, +\infty]$$

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

X, Y 2 random variables

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

$$Y : \Omega \rightarrow \mathbb{R}$$

measurable

$$\boxed{(X, Y) : \Omega \rightarrow \mathbb{R}^2 \quad \omega \mapsto (X(\omega), Y(\omega))}$$

$(X, Y) : \Omega \rightarrow \mathbb{R}^2$ is measurable

$$A \in \mathcal{B}(\mathbb{R}^2) \quad A = B \times C$$

$$(X, Y)^{-1}(B \times C) = \{\omega \mid X(\omega) \in B, Y(\omega) \in C\}$$

$$\underline{(X, Y)}(\omega) \rightarrow (\underline{X(\omega)}, \underline{Y(\omega)}) \in \underline{B} \times \underline{C}$$

$$= \underbrace{\{\omega \mid X(\omega) \in B\}}_{\substack{\uparrow \\ \mathcal{F}}} \cap \underbrace{\{\omega \mid Y(\omega) \in C\}}_{\substack{\uparrow \\ \mathcal{G}}} \in \mathcal{G}$$

X is measurable

Y is meas.

$$\forall (t, s) \in \mathbb{R}^2$$

$$\{\omega \mid X(\omega) \leq t, Y(\omega) \leq s\} = \{\omega \mid X(\omega) \leq t\} \cap \{\omega \mid Y(\omega) \leq s\}$$

joint law $\mathbb{P}_{(X,Y)} = (X,Y) \# \mathbb{P}$ is a Borel
meas. on $\mathbb{R}^2$

$$\mathbb{P}_{(X,Y)}(A \times B) = \mathbb{P} \{ \omega, X(\omega) \in A, Y(\omega) \in B \}$$

$$\mathbb{P}_{(X,Y)}(\mathbb{R} \times \mathbb{R}) = \mathbb{P}_{(X,Y)}(\mathbb{R}^2) = 1$$

cumulative distribution function $(x,y) \in \mathbb{R}^2$

$$\begin{aligned} F_{(X,Y)}(x,y) &= \mathbb{P}_{(X,Y)}([-\infty, x] \times [-\infty, y]) = \\ &= \mathbb{P}(\omega \mid X(\omega) \leq x, Y(\omega) \leq y) \end{aligned}$$

$$\begin{aligned} L_{(X,Y)}(A \times \mathbb{R}) &= \mathbb{P} \{ \omega \mid X(\omega) \in A, \underline{Y(\omega) \in \mathbb{R}} \} = \\ &= \mathbb{P}(\omega \mid X(\omega) \in A) = L_X(A) \end{aligned}$$

$$L_{(X,Y)}(\mathbb{R} \times B) = L_Y(B)$$

the first marginal of $L_{(X,Y)}$ is L_X

the second marginal of $L_{(X,Y)}$ is L_Y

$\mu: \mathcal{B}(\mathbb{R}^2) \rightarrow [0, \infty]$ measure

first marginal of μ

second marginal of μ

$\mu(\cdot \times \mathbb{R}): \mathcal{B}(\mathbb{R}) \rightarrow [0, \infty]$

$\mu(\mathbb{R} \times \cdot): \mathcal{B}(\mathbb{R}) \rightarrow [0, \infty]$

$\mathcal{P}(\mathbb{R})$ = probability measure on $\mathbb{R}$ =

= { μ Borel measures on $\mathbb{R}$ $\mu(\mathbb{R})=1$ }

$p \geq 1$
= laws of all possible random variables taking values in $\mathbb{R}$.

$$\mathcal{P}_p(\mathbb{R}) \subseteq \mathcal{P}(\mathbb{R})$$

$\mathcal{P}_p(\mathbb{R})$ = { probability measures on $\mathbb{R}$ with finite p -moment }

= { μ Borel measure, $\mu(\mathbb{R})=1$

and $\int_{\mathbb{R}} |x|^p d\mu < +\infty$ }

$\mathcal{P}_1(\mathbb{R})$ = laws of all random variables X $\mathbb{E}(X) < +\infty$

$\mathcal{P}_2(\mathbb{R})$ = " " " " " X $\mathbb{E}(X^2) < +\infty$

BACK TO ABSTRACT DEFINITIONS

Def of METRIC SPACE

A a set : distance on A (metric on A)

$$d: A \times A \rightarrow [0, +\infty)$$

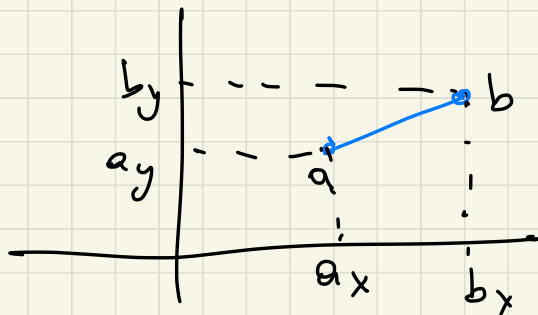
$$| \quad a, b \rightarrow d(a, b) = \text{distance between } a, b.$$

[d is a distance if it satisfies :

1) $d(a, a) = 0$

2) $d(a, b) = d(b, a)$

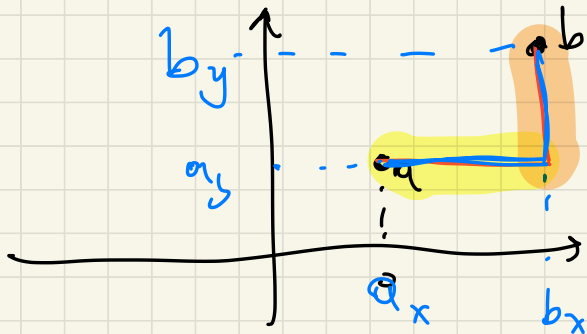
3) $d(a, b) \leq d(a, c) + d(c, b)$

$\mathbb{R}^2$ 

$$d_1(a, b) = |b - a|$$

length of the interval a, b

$$= \sqrt{(b_x - a_x)^2 + (b_y - a_y)^2}$$

 $\mathbb{R}^2$ 

$$d_2(a, b) = |b_x - a_x| + |b_y - a_y|$$

$\mathcal{P}(\mathbb{R})$

$\mu, \nu \in \mathcal{P}(\mathbb{R})$

$$d_{TV}(\mu, \nu) = 2 \sup_{A \in \mathcal{B}(\mathbb{R})} |\mu(A) - \nu(A)|$$

total variation
distance

$$d_{TV}(\mu, \nu) \leq d_{TV}(\mu, \rho) + d_{TV}(\rho, \nu)$$

$$2 \sup_A |\mu(A) - \nu(A)| = 2 \sup_A \underbrace{|\mu(A) - \rho(A) + \rho(A) - \nu(A)|}$$

$$\leq 2 \sup_A |\mu(A) - \rho(A)| + |\rho(A) - \nu(A)| \leq$$

$$\leq 2 \sup_{B \in \mathcal{B}(\mathbb{R})} |\mu(B) - \rho(B)| + 2 \sup_{C \in \mathcal{B}(\mathbb{R})} |\rho(C) - \nu(C)|$$

(A, d) $d: A \times A \rightarrow [0, +\infty)$ distance
is a metric space

(A, d) is COMPLETE if every Cauchy
sequence in A has a limit in A .

GIVEN A DISTANCE, I HAVE A NOTION OF CONVERGENCE

① $(a_n) \in A$ $a_n \rightarrow a$ in A if $\lim_{n \rightarrow +\infty} d(a_n, a) = 0$
 $\underbrace{a_1 \in A \ a_2 \in A \dots}_{n}$

(ex $\mu_n \in \mathcal{B}(\mathbb{R})$ $\mu_n \xrightarrow{TV} \mu$ if

$\lim_{n \rightarrow +\infty} \sup_{A \in \mathcal{B}(\mathbb{R})} |\mu_n(A) - \mu(A)| = 0 \Rightarrow$

$\forall A \in \mathcal{B}(\mathbb{R})$
 $\mu_n(A) \rightarrow \mu(A)$

Def of CAUCHY SEQUENCE

$(a_n)_{n \in \mathbb{N}}$ $\forall n, a_n \in A$ is a Cauchy sequence if

$$\lim_{\substack{m \rightarrow +\infty \\ n \rightarrow +\infty}} d(a_m, a_n) = 0$$

Obs if $a_n \rightarrow a$ then $(a_n)_n$ is a Cauchy sequence

$$d(a_m, a_n) \leq d(a_m, a) + d(a, a_n)$$

$$0 \leq \lim_{\substack{m \rightarrow +\infty \\ n \rightarrow +\infty}} d(a_m, a_n) \leq \lim_{m \rightarrow +\infty} d(a_m, a) + \lim_{n \rightarrow +\infty} d(a, a_n) = 0$$

Ex $\mathbb{Q}$ = rational numbers

$$d(a, b) = |b - a|$$



$(\mathbb{Q}, d)$ is NOT complete

$$a_0 = 1 \quad a_1 = 1, 4 \quad a_2 = 1, 41 \quad a_3 = 1, 414$$

$$\underline{a_4 = 1, 4142}$$

$$\underline{a_5 = 1, 41423 \dots}$$

$$\sqrt{2} = \dots$$

a_n = the first n -elements
of the number $\sqrt{2}$

B be a linear space on $\mathbb{R}$ (vectorial space)

B is a set such that

$$\forall a, b \in B \quad a + b \in B \quad (0 \in B \quad 0 + a = a \quad \forall a \in B)$$

$$\forall a \in B \quad \forall \lambda \in \mathbb{R} \quad \lambda \cdot a \in B$$

A NORM on B is a function $B \rightarrow [0, +\infty)$

such that

$$\textcircled{1} \quad \|0\| = 0 \quad \|b\| = 0 \Leftrightarrow b = 0$$

$$\textcircled{2} \quad \forall \lambda \in \mathbb{R} \quad \|\lambda b\| = |\lambda| \|b\| \quad \forall b \in B$$

$$\textcircled{3} \quad \forall a, b \in B \quad \|a + b\| \leq \|a\| + \|b\|$$

a norm INDUCES a DISTANCE on B $d(a, b) = \|b - a\|$

Not all the distances comes from norms

Def $(B, \|\cdot\|)$ is a Banach space if it is complete as a metric space

$(B, \|\cdot\|) \rightarrow$ is also a metric space
 $d(a, b) = \|b - a\| = \|a - b\|$

$$b_n \in B \quad b_n \rightarrow b \iff d(b_n, b) \rightarrow 0$$
$$\lim_{n \rightarrow \infty} \|b_n - b\| = 0$$

$$(\mathbb{R}, |\cdot|)$$

is a Banach space

$$(\mathbb{R}^2, \underline{|\cdot|})$$

$$\|(x, y)\| = \sqrt{x^2 + y^2}$$

$$d(a, b) = \|b - a\| = \sqrt{(b_x - a_x)^2 + (b_y - a_y)^2}$$

$\mathbb{R}^2, \|\cdot\|_1$

$$\|(x, y)\|_1 = |x| + |y| \quad (x, y) \in \mathbb{R}^2$$

$$\begin{aligned} d_1(a, b) &= \|(a_x, a_y) - (b_x, b_y)\|_1 = \\ &= |a_x - b_x| + |a_y - b_y| \end{aligned}$$



$$\mathbb{E}_X (\Omega, \mathcal{F}, \mathbb{P})$$

$$X: \Omega \rightarrow \mathbb{R} \quad \text{random variables}$$

$$\mathbb{E}(X) = \int_{\mathbb{R}} x \, d\mathbb{P}_X(x) < +\infty \quad \left(\mathbb{P}_X \in \mathcal{P}_1(\mathbb{R}) \right)$$

$$M^1 = \left\{ X: \Omega \rightarrow \mathbb{R} \quad \text{random variable on } (\Omega, \mathcal{F}, \mathbb{P}) \right. \\ \left. \text{with } \mathbb{E}(|X|) < +\infty \right\} \quad \mathbb{E}(X) < +\infty \Leftrightarrow \mathbb{E}(|X|) < +\infty$$

$$M^1 \text{ is a linear space} \quad X, Y \in M^1 \quad X+Y \in M^1$$

$$\lambda \in \mathbb{R} \quad \lambda X \in M^1 \quad \mathbb{E}(\lambda X) = \lambda \mathbb{E}(X) < +\infty \quad \mathbb{E}(X+Y) = \mathbb{E}(X) + \mathbb{E}(Y)$$

on $\mathcal{H}^1$ $\|X\| = \mathbb{E}(|X|) = \int_{\mathbb{R}} \underbrace{|x|}_{\text{blue}} \underbrace{d\mathcal{P}_X(x)}_{\text{red}}$

↓

$\|\cdot\|$ is a norm

1) $\|X\| = 0 = \mathbb{E}(|X|) \Leftrightarrow X = 0$ $X(\omega) = 0$ ~~for~~ with probability 1

$X(\omega) = 0$ for $\omega \in \Omega'$
 $\mathbb{P}(\Omega') = 1$

↓

$X = 0$

2) $\|\lambda X\| = \mathbb{E}(|\lambda X|) = |\lambda| \mathbb{E}(|X|) = |\lambda| \|X\|$

3) $\|X+Y\| = \mathbb{E}(|X+Y|) \leq \mathbb{E}(|X| + |Y|) = \|X\| + \|Y\|$

$$d(X, Y) = \|X - Y\| = \mathbb{E}(|X - Y|)$$

is a distance on M^1

$X_n \rightarrow X$ in this sense if

$$\lim_{n \rightarrow \infty} \mathbb{E}(|X_n - X|) = 0 \quad (X_n \rightarrow X \text{ in } \underline{\text{mean}})$$

$(M^1, \|\cdot\|)$ is Banach.