

Let $P(\mathbb{R}) = \{\mu \text{ Borel measure } \mu(\mathbb{R}) = 1\}$

Let $\mu, \nu \in P(\mathbb{R}) \quad p \in [1, +\infty)$

Move OPTIMAL TRANSPORT PROBLEM:

Among all maps $\psi: \mathbb{R} \rightarrow \mathbb{R}$ which transport μ to ν

that is $\psi \# \mu = \nu$, i.e. $\nu(A) = \mu\{\psi^{-1}(A)\}$

find the one which has minimal cost

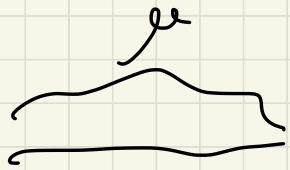
$$\inf_{\psi: \mathbb{R} \rightarrow \mathbb{R}} \int_{\mathbb{R}} \underbrace{|x - \psi(x)|^p}_{\text{cost}} d\mu(x)$$

$$\psi \# \mu = \nu$$

cost of moving x to $\psi(x)$ is
given by $|x - \psi(x)|^p$

$$\mathbb{R} \xrightarrow{\psi} \mathbb{R}$$

$$x \mapsto \psi(x)$$



$$\nu = \psi \# \mu$$

$$\nu(A) = \mu \{ x \in \mathbb{R} \mid \psi(x) \in A \}$$

$$\boxed{\nu(A) = \mu(\psi^{-1}(A))}$$

$$\nu(\psi(B)) = \nu(B)$$

$$\begin{array}{ccc} B & \xrightarrow{\psi} & \psi(B) \\ \mu(B) & \xrightarrow{\quad} & \nu(\psi(B)) \end{array}$$

We may also restate the problem as follows

Let $\mu \in P(\mathbb{R})$ and $\bar{\psi} : \mathbb{R} \rightarrow \mathbb{R}$ GIVEN

It is possible to find $\psi : \mathbb{R} \rightarrow \mathbb{R}$ such that

$$\textcircled{1} \quad \psi \# \mu = \bar{\psi} \# \mu$$

$$\textcircled{2} \quad \int_{\mathbb{R}} |x - \psi(x)|^p dx \leq \int_{\mathbb{R}} |x - \bar{\psi}(x)|^p dx$$

(Among all possible transport maps ψ such that

$\psi \# \mu$ is the same ($= \bar{\psi} \# \mu = \nu$)

there is one which is associated to minimal cost?)

① First case in which this problem can be solved

$$\mu = \frac{1}{N} \delta_{x_1} + \frac{1}{N} \delta_{x_2} + \dots + \frac{1}{N} \delta_{x_N} \quad N \in \mathbb{N}.$$

↓

$$\nu = \frac{1}{N} \delta_{y_1} + \frac{1}{N} \delta_{y_2} + \dots + \frac{1}{N} \delta_{y_N}$$

All possible $\psi: \mathbb{R} \rightarrow \mathbb{R}$ which satisfy $\psi \# \mu = \nu$

are

$$\psi(x_i) = y_j \quad \begin{matrix} i = 1 \dots N \\ j = 1 \dots N \end{matrix}$$

(BIJECTIVE)

$$\psi: \{x_1, \dots, x_N\} \rightarrow \{y_1, \dots, y_N\}$$

Monge probleee

$$\mu = \frac{1}{N} \delta_{x_1} + \frac{1}{N} \delta_{x_2} + \dots + \frac{1}{N} \delta_{x_N}$$

inf

$$\psi: \{x_1 \dots x_N\} \rightarrow \{y_1 \dots y_N\}$$

Bijection

$$\int_{\mathbb{R}} |x - \psi(x)|^p d\mu(x)$$

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$$\left[(x_1 - \psi(x_1))^p + (x_2 - \psi(x_2))^p + \dots + (x_N - \psi(x_N))^p \right]$$

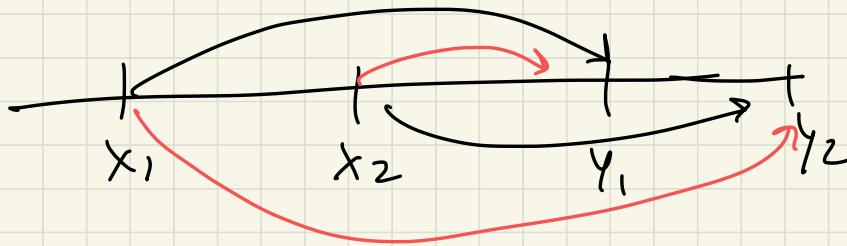
$$\psi: \{x_1 \dots x_N\} \rightarrow \{y_1 \dots y_N\}$$

Bijection

the solution exists since I am
minimizing on a finite set
(the set of all bijections $\{x_1 \dots x_N\} \rightarrow \{y_1 \dots y_N\}$ is finite)

UNIQUENESS.

example $\{x_1, x_2\} \rightarrow \{y_1, y_2\}$



$\psi_1: x_1 \rightarrow y_1$
 $\psi_1: x_2 \rightarrow y_2$
 $\cancel{\psi_2}: x_1 \rightarrow \cancel{y_2}$
 $x_2 \rightarrow y_1$

Cost of ψ_1 $|x_1 - y_1|^p + |x_2 - y_2|^p$

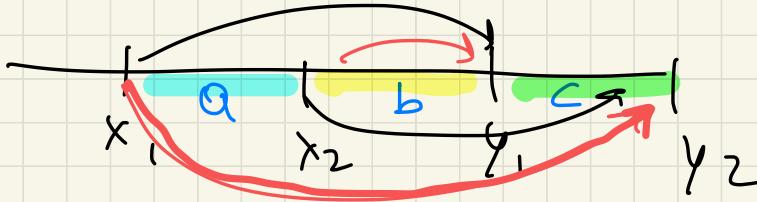
cost of ψ_2 $|x_1 - y_2|^p + |x_2 - y_1|^p$

$p=1$

Cost of $\psi_1 =$ Cost of ψ_2 -

$$|x_1 - y_1| + |x_2 - y_2| = |x_1 - y_2| + |x_2 - y_1|$$

$$P=2$$



$$\begin{aligned}\text{Cost of } \psi_1 &= |x_1 - y_1|^2 + |x_2 - y_2|^2 = \\ &= \underbrace{(a+b)^2}_{\text{blue}} + \underbrace{(b+c)^2}_{\text{yellow}} = a^2 + 2b^2 + c^2 + 2ab + 2bc\end{aligned}$$

$$\begin{aligned}\text{Cost of } \psi_2 &= |x_1 - y_2|^2 + |x_2 - y_1|^2 = \\ &= (a+b+c)^2 + b^2 = a^2 + 2b^2 + c^2 + 2ab + 2bc + \underline{\underline{2ac}}\end{aligned}$$

Cost of $\psi_1 < \text{Cost of } \psi_2$
 ψ_1 is the MONOTONE TRANSPORT MAP

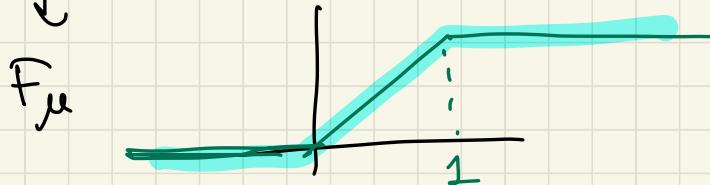
$$x_1 < x_2 \quad \psi(x_1) < \psi(x_2)$$

Ex (BOOK SHIFT)

$$\gamma_1 \# \mu = \gamma \quad \gamma_2 \# \mu = \gamma$$

$$\mu = \mathcal{L}_{[0,1]}$$

$$\downarrow$$

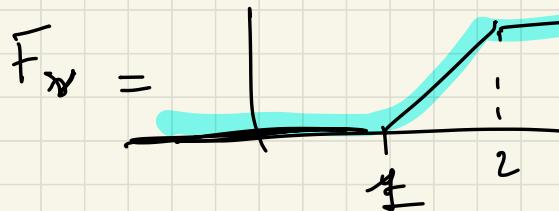


(Lebesgue measure restricted to
[0,1])

$$F_\mu(x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x \leq 1 \\ 1 & x > 1 \end{cases}$$

$$\gamma = \mathcal{L}_{[1,2]}$$

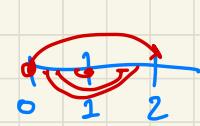
Lebesgue measure restricted to [1,2]



$$\psi_1: x \rightarrow 1+x$$



$$\psi_2: x \rightarrow 2-x$$



Cost of ψ_1

$$p=1$$

$$\mu = \begin{cases} 1 & x \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$

$$\mu = \int_{[0, 1]} dx$$

$$\int_{\mathbb{R}} |x - \psi_1(x)|^p d\mu(x) =$$

$$= \int_0^1 |x - \psi_1(x)|^p dx = \psi_1(x) = x + 1$$

$$= \int_0^1 |x - x - 1|^p dx = 1$$

Cost of ψ_2

$$\psi_2(x) = 2 - x$$

$$\int_{\mathbb{R}} |x - \psi_2(x)|^p d\mu = \int_0^1 |x - \psi_2(x)|^p dx =$$

$$= \int_0^1 |x - 2 + x|^p dx = \int_0^1 \underbrace{2 - 2x}_{2 - 2x} dx$$

$$= 2 - \left[x^2 \right]_0^1 = 2 - 1 = 1$$

$$p=1 \quad \text{Cost } \psi_2 = \text{Cost } \psi_1$$

$$\phi = 2 \quad \text{cost } \psi_1 = 1$$

$$\text{cost } \psi_2 = \int_0^1 (2-2x)^2 dx =$$

$$= \int_0^1 4 - 4x + 4x^2 dx = 4 - 2[x^2]_0^1 + \frac{4}{3}[x^3]_0^1 =$$
$$= 4 - 2 + \frac{4}{3} \geq 1$$

~ . ~

So a solution to the Monge problem may not be unique

Moreover if μ, ν are given

$\{ \psi : \mathbb{R} \rightarrow \mathbb{R} \mid \psi \# \mu = \nu \}$ COULD ALSO BE EMPTY

$$\text{ex } \mu = \frac{1}{2} \delta_0 + \frac{1}{2} \delta_1, \quad \nu = \mathbb{1}_{[0,1]}$$

WE RESTATE THE PROBLEM IN A MORE GENERAL SETTING

$$\psi \# \mu = \nu \quad \Rightarrow \quad (\text{id}, \psi) \# \mu = \pi$$

$$\nu(B) = \mu \{ x \in \mathbb{R} \mid \psi(x) \in B \}$$

π is a probability measure
on $\mathbb{R} \times \mathbb{R}$

$$(\text{id}, \psi) : \mathbb{R} \xrightarrow{\mu} \mathbb{R} \times \mathbb{R}$$
$$x \mapsto (x, \psi(x))$$

$$\pi(A \times B) = (\text{id}, \psi) \# \mu(A \times B) = \mu \{ x \in \mathbb{R} \mid (x, \psi(x)) \in A \times B \}$$

$$= \mu \{ x, \quad x \in A, \quad \psi(x) \in B \}$$

$$\pi(A \times \mathbb{R}) = \mu \{ x, \quad x \in A, \quad \psi(x) \in \mathbb{R} \} = \mu(A)$$

$$\pi(\mathbb{R} \times B) = \mu \{ x, \quad x \in \mathbb{R}, \quad \psi(x) \in B \} = \nu(B)$$

So $\psi: \mathbb{R} \rightarrow \mathbb{R}$ such that $\psi \# \mu = \nu$ is associated to a measure π on $\mathbb{R} \times \mathbb{R}$ such that $\pi(A \times \mathbb{R}) = \mu(A)$
 $\pi(\mathbb{R} \times B) = \nu(B)$

Instead of considering $\{\psi: \mathbb{R} \rightarrow \mathbb{R} \mid \psi \# \mu = \nu\}$
 (set which can be empty) I consider the set
 $\{\pi: \mathbb{B}(\mathbb{R} \times \mathbb{R}) \rightarrow \mathbb{R}_0 + \infty \mid \pi(A \times \mathbb{R}) = \mu(A) \quad \pi(\mathbb{R} \times B) = \nu(B)\}$

COUPLING BETWEEN μ, ν

$\{\psi: \mathbb{R} \rightarrow \mathbb{R} \mid \psi \# \mu = \nu\} \subseteq \{\pi \text{ coupling between } \mu, \nu\}$

↑
 (indeed $\pi = (\text{id}, \psi) \# \mu$ is a coupling)

The coupling between μ and ν is also called a
 TRANSPORT PLAN (generalization of the transport map $\psi: \mathbb{R} \rightarrow \mathbb{R}$)

$$\pi(A \times B) = \int_A \pi(\{x\} \times B) d\mu(x) = \int_B \pi(A \times \{y\}) d\nu(y)$$

$$A \times B = \bigcup_{x \in A} (\{x\} \times B)$$

("DISINTEGRATION OF MEASURES")

~ ~

MONGE PROBLEM

$$\inf_{\Psi: \mathbb{R} \rightarrow \mathbb{R}} \int_{\mathbb{R}} |x - \Psi(x)|^p d\mu$$

$\Psi \# \mu = \gamma$

$\pi = (\text{id}, \Psi) \# \mu$

KANTOROVICH PROBLEM

$$\geq \inf_{\pi} \int_{\mathbb{R}^2} |x - y|^p d\pi(x, y)$$

π COUPLING

$$\pi(A \times \mathbb{R}) = \mu(A)$$

$$\pi(\mathbb{R} \times B) = \nu(B)$$

→

the set of all couplings is always NOT EMPTY

$\pi(A \times B) = \mu(A) \times (B)$ is always a possible coupling

$\inf_{\pi \text{ coupling between } \mu, \nu}$

$$\int_{\mathbb{R} \times \mathbb{R}} |x-y|^p d\pi(x, y) = \inf_{\mathbb{R} \text{ prob. space}} \mathbb{E} |X-Y|^p$$

$$\pi(A \times B) = \mu(A)$$

$$\pi(\mathbb{R} \times B) = \nu(B)$$

$\mathbb{R} \text{ prob. space}$
 $\mathcal{L}X = \mu$
 $\mathcal{L}Y = \nu$
 $X, Y \text{ random variables}$

π coupling between $\mu, \nu \rightsquigarrow \pi$ joint law of some X, Y

$$\mathcal{L}X = \mu \quad \mathcal{L}Y = \nu$$

$$\psi: \mathbb{R} \rightarrow \mathbb{R} \text{ transport map } \psi \# \mu = \nu \rightarrow \exists X, Y \quad \mathcal{L}X = \mu \quad \mathcal{L}Y = \nu \quad Y = \psi(X)$$

DEFINITION

let $\mathcal{P}_1(\mathbb{R}) = \{\mu \in \mathcal{P}(\mathbb{R})$

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

let $\mu, \nu \in \mathcal{P}_1(\mathbb{R})$

$W_1(\mu, \nu) = \text{WASSERSTEIN DISTANCE} = \inf_{\substack{\pi \\ \text{COUPLING} \\ \text{BETWEEN } \mu, \nu}} \int_{\mathbb{R} \times \mathbb{R}} |x-y| d\pi(x, y)$

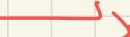
$(\mathcal{P}_1(\mathbb{R}), W_1)$ is a complete METRIC SPACE

Note that $W_1(\mu, \nu) = \inf_{\Omega} \mathbb{E} |X-Y|$

Ω

$\models x = \mu$

$\models y = \nu$

 distance in

$M^1 = \{X : \Omega \rightarrow \mathbb{R} \mid \mathbb{E}|X| < \infty\}$

(distance between μ, ν is the minimal distance in M^1 sense between random variables which have laws μ and ν).

Let $\mu, \nu \in \mathcal{P}_2(\mathbb{R})$

$$W_2(\mu, \nu) = \inf_{\substack{\pi \text{ coupling} \\ \pi(A \times \mathbb{R}) = \mu(A) \\ \pi(\mathbb{R} \times B) = \nu(B)}} \left[\int_{\mathbb{R} \times \mathbb{R}} |x-y|^2 d\pi(x, y) \right]^{1/2}$$

distance (\mathcal{P}_2, W_2) is a COMPLETE metric space

$$= \inf_{\substack{\Omega \\ \mathbb{P} X = \mu \\ \mathbb{P} Y = \nu}} \left[\mathbb{E} |X-Y|^2 \right]^{1/2}$$

distance between
X and Y in \mathbb{M}^2

$W_2(\mu, \nu) = \text{infimum of}$
all possible distances

in \mathbb{M}^2 sense of random

variables X, Y , $\mathbb{P} X = \mu$
 $\mathbb{P} Y = \nu$

Theorem (BRENTER)

1) $\forall \mu, \nu \in \mathcal{P}_p(\mathbb{R}) \quad \exists$ always at LEAST one $\bar{\pi}$ coupling such that $\int_{\mathbb{R} \times \mathbb{R}} |x-y|^p d\bar{\pi}(x,y) = \inf_{\substack{\pi \text{ coupling}}} \int_{\mathbb{R} \times \mathbb{R}} |x-y|^p d\pi(x,y)$ (for every $p \geq 1$)

($\exists \bar{x}, \bar{y}$ random variables

$$W_p(\mu, \nu) = \left[\mathbb{E}(|\bar{x} - \bar{y}|^p) \right]^{1/p}.$$

$$L_{(\bar{x}, \bar{y})} = \bar{\pi} \quad L_{\bar{x}} = \mu$$

$$L_{\bar{y}} = \nu$$

2) $\bar{\pi}$ is NOT UNIQUE for $p=1$

$\bar{\pi}$ is UNIQUE for $p=2$

($p > 1$)

either if $\mu \ll \mathcal{L}$

or if

μ, ν are both supported in some COMPACT

3) if $\mu \ll L$ the optimal $\bar{\pi}$ is actually associated to a Transport map

$$\bar{\psi}$$

$$(v = \bar{\psi} \# \mu) \quad \bar{\gamma} = \psi(x)$$

$$W_2(\mu, v) = \left(\int_{\mathbb{R}} |x - \bar{\psi}(x)|^2 d\mu(x) \right)^{1/2} = (\mathbb{E} |x - \psi(x)|^2)^{1/2}$$

where $\bar{\psi}$ is MONOTONE

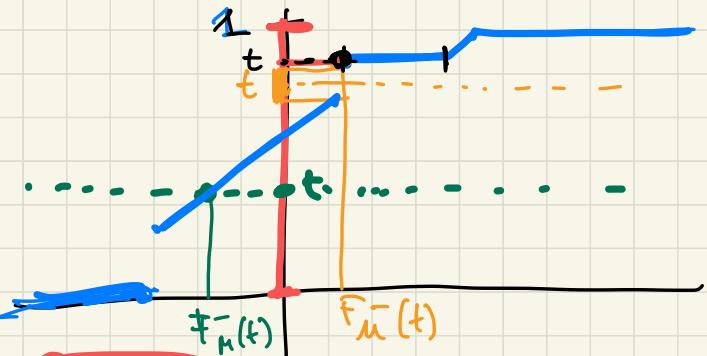
In dimension 1 everything can be computed EXPLICITLY.
by using cumulative DISTRIBUTION FUNCTIONS

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

$$F_\mu : \mathbb{R} \rightarrow [0, 1]$$

$$F_\mu(x) = \mu(-\infty, x]$$

Cumulative
DISTR. FUNCTION



$t \in [0, 1]$ (PSEUDO INVERSE)

$$F_\mu^-(t) = \inf \{ x \in \mathbb{R} \mid F_\mu(x) \geq t \}$$

F_μ is monotone, right cont.
 $0 \leq F_\mu(x) \leq 1$

$$F_\mu(-\infty) = 0$$

$$F_\mu^("+\infty") = 1$$

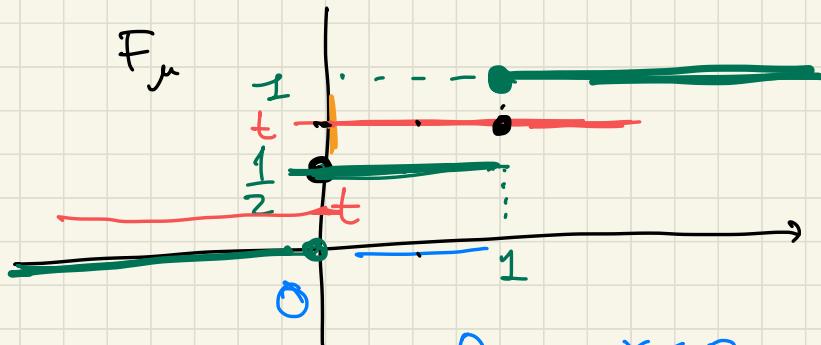
$$\mu(-\infty, +\infty) = 1$$

Ex 1

$$\mu = \frac{1}{2} \delta_0 + \frac{1}{2} \delta_1$$

$$\mu(A) = \begin{cases} 0 & 0, 1 \notin A \\ \frac{1}{2} & 0 \in A \text{ or } 1 \in A \\ 1 & 0, 1 \in A \end{cases}$$

$$F_\mu(x) = \mu(-\infty, x]$$



$$\mu(-\infty, x] = \begin{cases} 0 & x < 0 \\ \frac{1}{2} & 0 \leq x < 1 \\ 1 & x \geq 1 \end{cases}$$

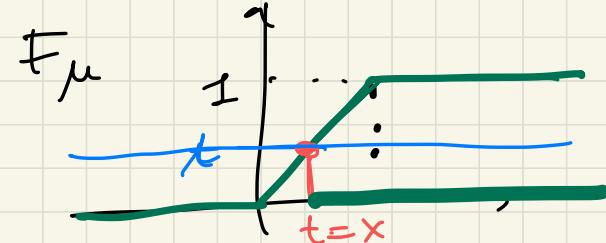
$$F_\mu^-(t) = \inf \{ x \mid F_\mu(x) \geq t \}$$

$$F_\mu^-(t) = \begin{cases} 0 & 0 < t \leq \frac{1}{2} \\ 1 & \frac{1}{2} < t < 1 \end{cases}$$

$F_\mu(x) \geq t > 0'$

$F_\mu(x) \geq t > \frac{1}{2}$

Ex 2 $\mu = \chi_{[0,1]} dx = \mathcal{L}_{[0,1]}$

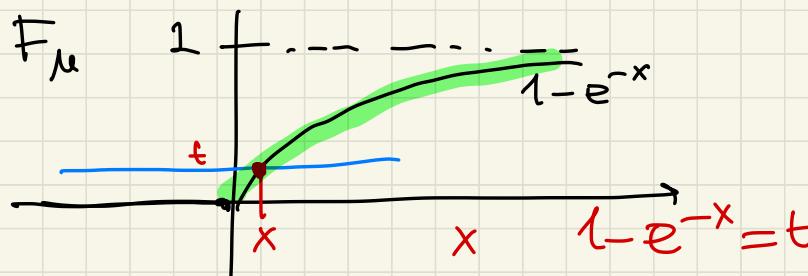


$$F_\mu^{-1}(t) = t = \inf \{x \mid F_\mu(x) \geq t\}$$

Ex 3 $\mu = e^{-x} \chi_{[0,+\infty)} dx$

$$F_\mu = \begin{cases} 0 & x < 0 \\ 1 - e^{-x} & x \geq 0 \end{cases}$$

$$\mu(-\infty, x] = \int_{-\infty}^x e^{-y} \chi_{[0,+\infty)}(y) dy = \int_0^x e^{-y} dy$$



$$F_\mu^{-1}(t) = \log\left(\frac{1}{1-t}\right) \quad t \in [0,1)$$

$$1 - t = e^{-x} \quad -x = \log(1-t)$$

Proposition $\mu = F_\mu^- \# \mathcal{L}_{[0,1]}$

(Observe $\mathcal{L}_{[0,1]} = \lambda_{[0,1]} dx$ Lebesgue measure restricted to $[0,1]$)

$$F_\mu^- : [0,1] \rightarrow \mathbb{R}^+$$

$$\mathcal{L}_{[0,1]} \rightsquigarrow F_\mu^- \# \mathcal{L}_{[0,1]}$$

Proof. First of all recall that

$$F_\mu^-(t) = \inf \{x \mid F_\mu(x) \geq t\}$$

then ① $F_\mu^-(t) \leq a \Leftrightarrow \inf \{x \mid F_\mu(x) \geq t\} \leq a \Leftrightarrow \forall x \ F_\mu(x) \geq t \Rightarrow x \leq a$
 $\Rightarrow F_\mu(a) \geq t$

② $F_\mu^-(t) > a \Leftrightarrow \inf \{x \mid F_\mu(x) \geq t\} > a \Leftrightarrow \forall x \ F_\mu(x) \leq t \text{ if hold } x > a$
 $\Rightarrow F_\mu(a) < t$

Let us fix $x \in \mathbb{R}$

$$\begin{aligned} F_\mu^- \# \mathcal{L}_{[0,1]} (-\infty, x] &= \mathcal{L} \{t \in [0,1] \mid F_\mu^-(t) \in (-\infty, x]\} \\ &= \mathcal{L} \{t \in [0,1] \mid F_\mu^-(t) \leq x\} = \\ &= \mathcal{L} \{t \in [0,1] \mid F_\mu(x) \geq t\} = \mathcal{L} [0, F_\mu(x)] \end{aligned}$$

$$F_{\mu} \# \mathcal{L}_{[0,1]} \left(-\infty, x \right] = \underbrace{\mathcal{L}([0, F_{\mu}(x)])}_{\text{LENGTH OF THE INTERVAL}} = F_{\mu}(x) = \mu \left(-\infty, x \right]$$

↓
LENGTH OF THE INTERVAL $\mathcal{L}[0, a] = a$

therefore $\forall x \in \mathbb{R}$ $F_{\mu} \# \mathcal{L}_{[0,1]} \left(-\infty, x \right] = \mu \left(-\infty, x \right]$

$$F_{\mu} \# \mathcal{L}_{[0,1]} = \mu$$

↓

Proposition Let $\mu, \nu \in P(\mathbb{R})$ such that F_{μ} is continuous
(e.g. $\mu \ll \mathcal{L}$) Then it holds

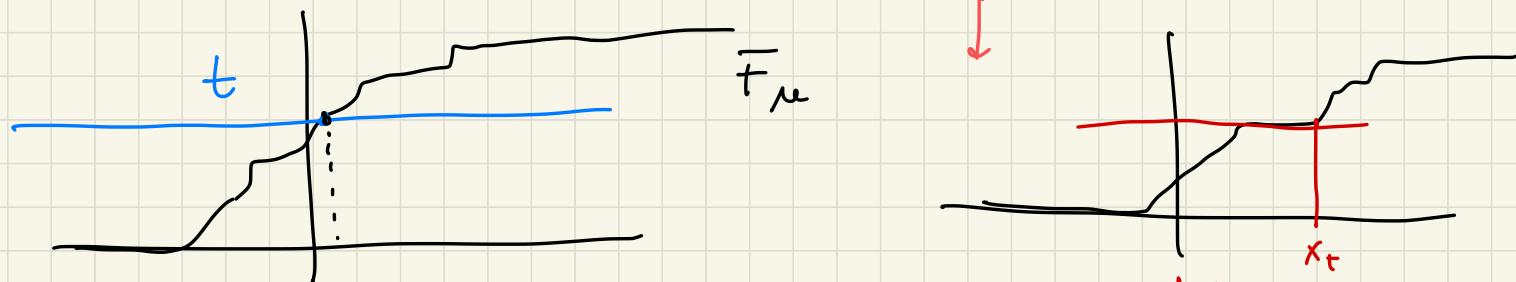
$$(F_{\nu} \circ F_{\mu}) \# \mu = \nu$$

need to show

$F_{\nu} \circ F_{\mu} : \mathbb{R} \rightarrow \mathbb{R}$
is MONOTONE.

Proof observation: if F_μ is continuous it holds

$$\forall t \in [0,1] \quad \{x, F_\mu(x) \leq t\} = (-\infty, x_t] \quad \text{where } F_\mu(x_t) = t$$



F_μ is continuous \rightarrow it has no jumps and by the
INTERMEDIATE VALUE THEOREM for continuous functions

$$\forall t \in [0,1] \quad \exists \text{ at least ONE } x_t \quad F_\mu(x_t) = t$$

(may be more than one!)

↓
in any case $\{x, F_\mu(x) \leq t\} = (-\infty, x_t]$ with $F_\mu(x_t) = t$

x_t maximal x

Fix $x \in \mathbb{R}$

$$F_V^{-1} \circ F_\mu: \mathbb{R} \xrightarrow{\mu} \mathbb{R}$$

$$(F_V^{-1} \circ F_\mu) \# \mu (-\infty, x] = \mu \{ y \in \mathbb{R} \mid F_V^{-1} \circ F_\mu(y) \in (-\infty, x] \}$$

$$= \mu \{ y \in \mathbb{R} \mid \underbrace{F_V^{-1}(F_\mu(y))}_{F_V^{-1}(F_\mu(y)) \leq x} \leq x \} = \left[\begin{array}{l} F_V^{-1}(a) \leq x \\ \uparrow \\ F_V(t) \geq a \end{array} \right]$$
$$= \mu \{ y \in \mathbb{R} \mid \underbrace{F_V(y) \geq F_\mu(y)}_{F_V(y) \geq F_\mu(y)} \}$$

by continuity of F_μ $\{ y \mid F_\mu(y) \leq t \} = (-\infty, y_+]$ $F_\mu(y_+) = t$

$$= \mu (-\infty, y_{F_V(x)}] \quad \text{where } F_\mu(y_{F_V(x)}) = F_V(x).$$

$$= F_\mu(y_{F_V(x)}) = F_V(x) = \nu(-\infty, x]$$

$$\forall x \quad (F_V^{-1} \circ F_\mu) \# \mu (-\infty, x] = \nu(-\infty, x] \Rightarrow \boxed{(F_V^{-1} \circ F_\mu) \# \mu = \nu}$$

COROLLARY of the BRENIER THEOREM.

If $\mu \ll \mathbb{P}$ then $\forall \nu \in \mathbb{P}(\mathbb{R})$, $\forall p \geq 1$

$$\inf_{\substack{\pi \text{ coupling} \\ \text{between } \mu, \nu}} \int_{\mathbb{R} \times \mathbb{R}} |x - y|^p d\pi(x, y) = \int_{\mathbb{R}} |x - F_{\nu}^{-1} \circ F_{\mu}(x)|^p d\nu(x)$$

(this means that the TRANSPORT MAP - which

is MONOTONE - $\psi = F_{\nu}^{-1} \circ F_{\mu} : \mathbb{R} \rightarrow \mathbb{R}$

has minimal cost among all possible transport plans between μ, ν (also among all possible couplings).

For $p=1$ it is not the UNIQUE one.

For $p > 1$ it is the UNIQUE ONE.

$$\mathcal{E} X \quad \mu = e^{-x} \int_{[0, +\infty)} x e^{-x} dx$$

(EXPONENTIAL DISTRIBUTION)

$$F_{\mu}(x) = \mu(-\infty, x] = \int_0^x e^{-t} dt = 1 - e^{-x}$$

$$F_{\mu}(x) = \begin{cases} 0 & x \leq 0 \\ 1 - e^{-x} & x > 0 \end{cases}$$

continuous!

$$\lambda \in [0, 1]$$

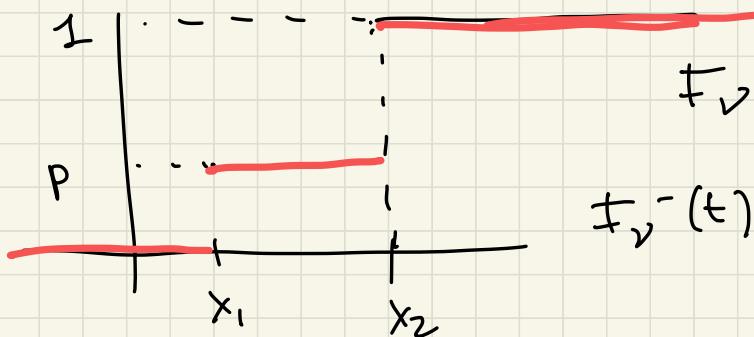
$$V = \lambda \delta_{x_1} + (1-\lambda) \delta_{x_2}$$

$$x_1 < x_2$$

$$V(A) = \begin{cases} 0 & x_1, x_2 \notin A \\ \lambda & x_1 \in A \quad x_2 \notin A \\ 1-\lambda & x_1 \notin A \quad x_2 \in A \\ 1 & x_1, x_2 \in A \end{cases}$$

(BERNOULLI DISTRIBUTION)

$$F_V(x) = \begin{cases} 0 & x < x_1 \\ \lambda & x_1 \leq x < x_2 \\ 1 & x_2 \leq x \end{cases}$$



$$F_V^-(t) = \inf \{x \mid F_\mu(x) \geq t\}$$

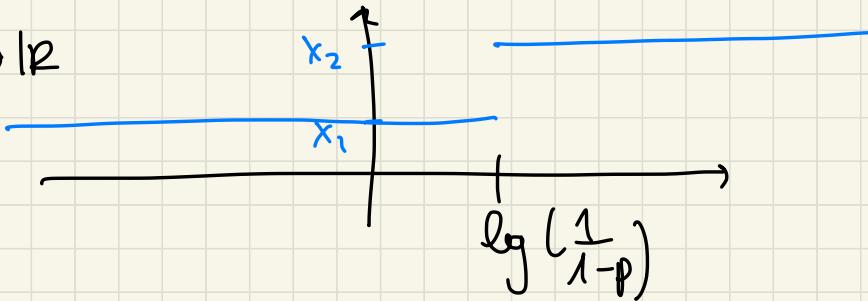
$$F_V^-(t) = \begin{cases} x_1 & 0 < t \leq p \\ x_2 & p < t \leq 1 \end{cases} \quad \begin{matrix} F_\mu(x) \geq t > 0 \\ F_\mu(x) \geq t > p \end{matrix}$$

$$F_V^- \circ F_\mu(x) = \begin{cases} x_1 & 0 < F_\mu(x) \leq p \implies x \leq \log\left(\frac{1}{1-p}\right) \\ x_2 & p < F_\mu(x) \implies x > \log\left(\frac{1}{1-p}\right) \end{cases}$$

$$F_\mu(x) = 1 - e^{-x} \leq p \implies 1 - p \leq e^{-x} \implies \log(1-p) \leq -x$$

$$\implies -\log\left(\frac{1}{1-p}\right) \leq -x \implies \log\left(\frac{1}{1-p}\right) \geq x$$

$$F_\nu^{-1} \circ F_\mu : \mathbb{R} \rightarrow \mathbb{R}$$



$$(F_\nu^{-1} \circ F_\mu) \# \mu = \nu$$

$$W_2(\mu, \nu)^2 = \inf_{\substack{\pi \\ \text{coupling}}} \int_{\mathbb{R} \times \mathbb{R}} |x - y|^2 d\pi(x, y) = \int_{\mathbb{R}} |x - F_\nu^{-1} \circ F_\mu(x)|^2 d\mu(x)$$

$$= \int_0^{+\infty} |x - F_\nu^{-1} \circ F_\mu(x)|^2 e^{-x} dx$$

$$= \int_0^{\lg(\frac{1}{1-p})} |x - x_1|^2 e^{-x} dx + \int_{\lg(\frac{1}{1-p})}^{+\infty} |x - x_2|^2 e^{-x} dx = \dots$$

$$\mu = \chi_{[0, +\infty)} e^{-x} dx$$

= (can be computed explicitly)

Note that $W_2(\mu, \nu) = W_2(\nu, \mu)$ since W_2 is a
distance!

(so it is also the optimal cost to transport ν to μ
- even if the transport is a transport plan
not a transport map!)