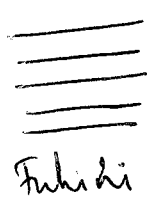
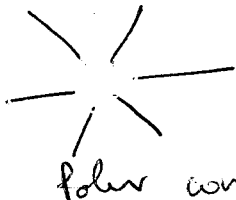


Goal: show to prove various analytic/geometric inequalities by reduction to 1-d analysis.



Fubini



Fubini words



(???)

Don't start up here.

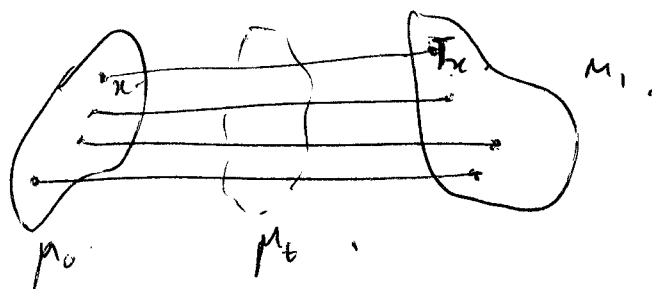
Example from OT: $\mu_1, \mu_2 \ll \text{Leb}$ on \mathbb{R}^n , $\int |x|^2 d\mu_i < \infty$ (prob measure).

$$W_2^2(\mu_1, \mu_2) = \min_{T: \mu_1 \rightarrow \mu_2} \int |Tx - x|^2 d\mu_1(x)$$

Brenier map T realises this min uniquely. (p.o.-a.e.)
 Characterised by $-T = \nabla \varphi$, $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}$ convex.

McCann '97: $\mu_t = \underbrace{\mathbb{R}[(1-t)\text{Id} + tT]}_{T_t} \# (\mu_0)$, $\dot{\mu}_t = \frac{d\mu_t}{dt}$

interpolation along 1-dim transport paths.



μ_t is, in fact, the geodesic b/w μ_0 and μ_1 in W_2 .

Defⁿ: $\mathbb{E}(\gamma) = \int \gamma(x) dx$ is called displacement convex if $[0,1] \ni t \mapsto \mathbb{E}(\mu_t)$ is convex $\mu_0, \mu_1 \ll \text{Leb}$.

McCann Characterisation: Enough to check displacement convexity when transform infinitesimal random elements which amounts to checking Jacobian of T_t . "1-d reduction"

↑
Take this as a principle.

T_t^μ (Brenier / Brenier comp-high' T_0). let $\mu = \mathcal{P}(n)$ dm in \mathbb{R}^n

$N \in (-\infty, 0) \cup [n, \infty]$, TFAE:

(I) $(N-n)t^{\frac{1}{N-n}}$ is concave on its convex set.

(II) $\forall A, B \in \text{CTR}^n \quad \mu((1-t)A + tB) \geq [(1-t)\mu(A)^{\frac{1}{N}} + t\mu(B)^{\frac{1}{N}}]^N$.

Rems: (I) $A+B = \{a+b : a \in A, b \in B\}$,
 (II) $N=n \Rightarrow \gamma \equiv 1$, $\mu = \text{leb}$ (Brenier-McCann),
 (III) $N=\infty$, Prékopa-herdler, $\mu((1-t)A + tB) \geq \mu(A)^{1-t} \mu(B)^t$.
 (IV) $N < 0$ ineq. are reversed.

Pf (II) \Rightarrow (I); use $A = x + B(a, \varepsilon)$, $B = y + B(b, \varepsilon)$, take $\varepsilon \rightarrow 0$ and optimize in $a, b > 0$.

(I) \Rightarrow (II): true for $N \geq n$. let $T = \nabla \varphi$ be the Brenier map from $\mu_0 = \gamma_0 \mu - t_0$ to $\mu_1 = \gamma_1 \mu$.

$$\mu_t = (T_t)_\#(\mu_0) \quad \varphi_t = \frac{d\mu_t}{d\mu}$$

Step 1: show that claim follows from displacement convexity of $t \mapsto \text{Ent}_N(\mu_t(M)) = - \int \varphi_t \frac{N-1}{t} d\mu$.

$$\left[\text{Hint: } \gamma_0 = \frac{\gamma_A}{M(A)}, \gamma_1 = \frac{\gamma_B}{M(B)} \right]$$

$(1-t)A + tB \supset \text{int}(M_t)$. Use Jensen's inequality to get

$$\mu(\text{int}(M_t))^{\frac{1}{N}} \geq -\text{Ent}(\mu_t/\mu).$$

Step 2: Show that $\int \varphi_t(y)^{\frac{N-1}{N}} d\mu_t(y) = \int J_t(x)^{\frac{1}{N}} \rho_0^{\frac{N-1}{N}} dx$.

$$y = T_t x.$$

Reduces to 1-d analysis, i.e., that

$t \mapsto J_t(x)^N$ is concave.

Step 3: $J_t(x) = J_t^G(x) J_t^W(x)$. J_t^G geometric jacobian.
 J_t^W weighted jacobian.

[$J_t(x)$ is the jacobian of $T_t(x) = (1-t)x + t \nabla \varphi_{x-r+\mu}$].

$$J_t^G(x) = \det((1-t)\text{Id} + t \nabla^2 \varphi)$$

$$J_t^W(x) = \frac{\varphi(T_t(x))}{\varphi(x)}$$

$(J_t^G(x))^{\frac{1}{N}}$ concave. b/c $\det^{\frac{1}{N}}$ is concave on pos. def. matrices.

$(J_t^W(x))^{\frac{1}{N-1}}$ concave. by assumption — $\varphi^{\frac{1}{N}}$ concave in interval.

and then by Hölder. $J_t(x)^{\frac{1}{N}}$ is concave.

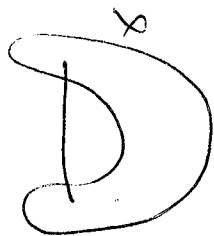
1-D localization

Setup (M, g) smooth, connected Riem. mfd.

induced geodesic distance

$$d(x, y) = \inf \left\{ \int_a^b |\dot{\gamma}(t)| dt ; \gamma(a) = x, \gamma(b) = y \right\}.$$

quodernally curves:



$\forall x, y \in M, \exists$
 dist minimizing
 geodesic connecting
 x and y inside
 set.

$$\mu(x) = \int_{\Omega} \mu_{\alpha} dv(\alpha)$$

← where.

$\gamma > 0$ not continuous, $\gamma_{\alpha} =$ where for $\sqrt{\det g_{ij}}$, dv_1, \dots, dv_n .

Def^v. $\mu = \int_{\Omega} \mu_{\alpha} dv(\alpha)$ is called a disintegration of μ .

- ① $(\Delta, \mathcal{F}, \nu)$ complete measure space.
- ② \forall sub-measure $A \in \mathcal{M}$
 - ⓐ $\Delta \ni \alpha \mapsto \mu_{\alpha}(A)$ is defined ν -a-e.
and is \mathcal{F} -measurable.
 - ⓑ $\mu(A) = \int_{\Omega} \mu_{\alpha}(A) dv(\alpha)$.

Def^v Δ -d. location of μ

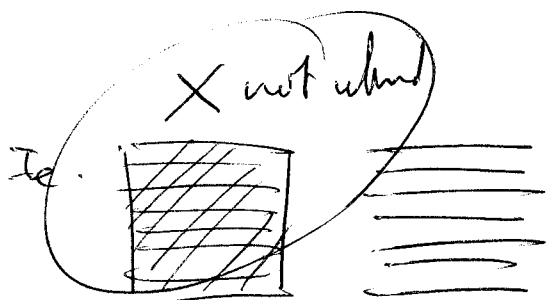
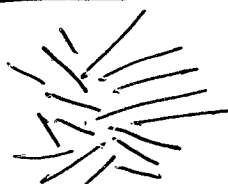
is a disintegration. $\mu = \int_{\Omega} \mu_{\alpha} dv(\alpha)$. s.t.

$\exists \Delta_0 \subset \Delta$. which $\forall (\Delta \setminus \Delta_0)$ s.t.

- ① $\forall \alpha \in \Delta_0$, μ_{α} is a " Δ -d" needle, is fixed on a
 distance minimizing geodesic $\gamma_{\alpha} = \underset{C \subset \mathbb{R}}{L_{\alpha}} \rightarrow M$.

$$\mu_{\alpha}(A \cap \gamma_{\alpha}) = \mu_{\alpha}(A).$$

- ② $\{\gamma_{\alpha}\}_{\alpha \in \Delta_0}$ are disjoint.



← almost ④

Defⁿ let $f \in L^1(\mu)$ $\int f d\mu = 0$.

A d -d localisation is called f -balanced if
 $\forall v = a e_1 + b e_2$ $\int f d\mu_v = 0$.

Such a thing is problem dependent, and extremely powerful!

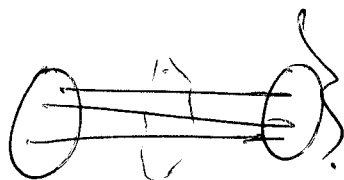
Ex^o

(I) Reduction of R_{μ} -type maps to 1 -d:

Want $\mu((1-t)A \overset{\sim}{\cup} tB) \geq F(\mu(A), \mu(B); t, \tilde{d}(A, B))$.

$F(v_1, v_2, t, \tilde{d})$ - homogenous in (v_1, v_2) and monotone in \tilde{d} .

$$(1-t)A + tB: \begin{cases} \exists z \in A: \exists x \in A, \exists y \in B. \\ d(x, y) = \frac{d(z, z)}{t} = \frac{d(z, y)}{t}. \end{cases}$$



Given $\dots A, B \subset M$, set $f = \frac{\chi_A}{\mu(A)} - \frac{\chi_B}{\mu(B)}$. $\int f d\mu = 0$.

let $\mu = \int \mu_\alpha d\nu(\alpha)$ f -balanced localisation.

If $\mu_\alpha((1-t)(A \cap \alpha) + t(B \cap \alpha)) \geq$

$F(\mu_\alpha(A \cap \alpha), \mu_\alpha(B \cap \alpha), t, \tilde{d}(A \cap \alpha, B \cap \alpha))$.

$\geq \frac{\mu_\alpha(A \cap \alpha)}{\mu(A)} F(\mu(A); \mu(B), t, d(A, B))$

$$\int f d\mu_\alpha = 0 \quad \frac{\mu_\alpha(A)}{\mu(A)} = \frac{\mu_\alpha(B)}{\mu(B)}$$

+ homogeneity.

Since $(1-t)A + tB \geq U_\alpha((1-t)(A \cap \sigma_\alpha) + t(B \cap \sigma_\alpha))$.

and sub additivity of measure.

$$\mu(\quad) \geq \mu(\quad).$$

$$\int \mu_\alpha((1-t)(A \cap \sigma_\alpha) + t(B \cap \sigma_\alpha)) d\nu(\alpha).$$

$$\geq F(\mu(A), \mu(B), t, \tilde{\alpha}(A, B)) \underbrace{\int \frac{\mu_\alpha(A)}{\mu(A)} d\nu(\alpha)}_{= 1}.$$

↗
Δ-d analysis