

In general, if we have a ~~non~~ fully
nonlinear ~~etc.~~ eqn:

$$F(D^2u) = f. \quad F(r) \in \mathbb{R}^n \rightarrow$$

F is elliptic if $\frac{\partial F}{\partial r_{ij}} > 0$.

uniformly elliptic if $\exists \lambda > 0, \Lambda > 0$
s.t.

$$\lambda I \leq \frac{\partial F}{\partial r_{ij}} \leq \Lambda I$$

Now, if $F(r) = \det r$, then

$$\frac{\partial F}{\partial r_{ij}} = (\text{cof } r)_{ij} \quad \text{Hence } \nabla^2 u$$

is elliptic iff $Du > 0$ i.e. u convex
BUT $\nabla^2 u$ is not uniformly elliptic
(unless you have a bound
for Du)

Sources of MA eqn & Applications:

① ~~prescribing curvatures of Gauss curvature~~

$$u: \Omega \xrightarrow{cR} \mathbb{R} \quad (\text{with } x^2)$$

Maue - Ampere equation (elliptic)

real version: $u: \Omega \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$

$\det D^2 u = f \rightarrow$ given function

A natural generalization of Poisson's equation:

$$\Delta u = f.$$

Connection between MA eqn & Poisson's equation:

$\lambda_1, \dots, \lambda_n$ be the eigenvalues of $D^2 u$.

then Poisson: $\sum_{i=1}^n \lambda_i = f.$

MA: $\prod_{i=1}^n \lambda_i = f.$

Eqn in between: m -Hessian eqn.
 $1 \leq m \leq n.$

$$\sigma_m(\lambda) = f.$$

$\sigma_m = m$ -th symmetric polynomial.

Q: When is MA eqn elliptic?

Ans (Non-elliptic eqn is very difficult...)

Good news - The later problem always

have a solution (possibly ~~not unique~~ ^(direct min))

Such a soln $\gamma = (\text{id} \times T) \# \mu$

if $\forall (x, y) \in \text{supp}(\gamma)$, $y =$
 $f(x)$ for μ -a.e. x . if $(x, y_1), (x, y_2)$

$$y_1 = y_2 \quad \in \text{supp}(\gamma)$$

"split mass of μ "
 "piece γ doesn't"

Example: $\nu = \delta_a$

$$\Pi(\mu, \nu) = \mu \times \delta_a$$

$$T(X) = \{a\}$$

$$\gamma(x) = \delta_{x, a}$$

Example: $X = \{x_1, \dots, x_n\}$, $Y = \{y_1, \dots, y_m\}$

$$\mu = \frac{1}{n} \sum_{i=1}^n \delta_{x_i}, \quad \nu = \frac{1}{m} \sum_{j=1}^m \delta_{y_j}$$

$$\Pi(\mu, \nu) = \left\{ \gamma_{ij} \right\}, \quad \gamma_{ij} = \gamma \left(\frac{1}{n} \delta_{x_i} \times \frac{1}{m} \delta_{y_j} \right)$$

$$\sum_{j=1}^m \gamma_{ij} = \frac{1}{n} \quad \sum_{i=1}^n \gamma_{ij} = \frac{1}{m} \quad \forall i, j$$

$$\int c(x, y) d\gamma = \sum_{i,j} c(x_i, y_j) \gamma_{ij}$$

① Prescribing the Gauss curvature
of a graph (positive)

$$K(x) = \frac{1}{(1 + |Du|^2)^{\frac{n}{2} + 1}} \det D^2 u$$

$\Omega \subseteq \mathbb{R}^n$ a convex domain, $u|_{\partial\Omega} = \varphi$
& Gauss curvature = $K(x)$.

② Optimal transport
(in ~~the~~ very special setting)

$\Omega_1, \Omega_2 \subseteq \mathbb{R}^n$ be two domains

$f > 0$ in Ω_1 , $g > 0$ in Ω_2

$$\int_{\Omega_1} f dx = \int_{\Omega_2} g dx$$

Look for a Borel map $T: \Omega_1 \rightarrow \Omega_2$

which minimizes $\int_{\Omega_1} |x - T(x)| f(x) dx$

with $T \# f dx = g dy$, which
minimizes $\int_{\Omega_1} |x - T(x)| f(x) dx$

Theorem (1) The optimal map is unique
given by the gradient of a convex function

$$(2) \text{det } D^2\varphi(x) = \frac{f(x)}{g(\varphi(x))} \quad (x)$$

(3) ~~For a convex function φ on S .~~

Let ψ be the Legendre transform of φ . Show that $S \ni y$ minimizes

$$\int_{\Omega_2} (y - S(y))^2 g(y) dy \quad \text{over all maps } S \# (\rho dy) = f dx$$

So the regularity of optimal maps is equivalent to studying the regularity of $f(x)$.

General set-up of optimal transport problem

Formulation of the Problem (Monge)

$(X, \mu), (Y, \nu)$ \Leftarrow two measure spaces,
 $\sum_i \mu(X) = \sum_i \nu(Y) = 1$

Consider all measurable maps

$$T: X \rightarrow Y \text{ s.t. } T\# \mu = \nu.$$

Wish $C: X \times Y \rightarrow \mathbb{R} \cup \{+\infty\}$ be
a measurable function.

wish to minimize

$$\int_X C(x, T(x)) d\mu(x).$$

$$\forall g \in L^1(Y, d\nu)$$

$$\int g(y) d\nu(y) = \int g(T(x)) d\mu(x)$$

Very In general very difficult to tell if
there is such a map

(Kantorovich formulation)

Just Idea: Relax the problem to allow for
"transport plans" i.e. ~~consider~~ we replace

~~optimal map~~ with

$$\text{the " } T\# \mu = \nu \text{ " to " } \gamma \in \Pi(\mu, \nu).$$

$$\text{s.t. } (\text{proj}_1)_\# \gamma = \mu \quad (\text{proj}_2)_\# \gamma = \nu$$

& minimize $\int_{X \times Y} C(x, y) d\gamma(x, y)$