

The Fokker-Planck equation as a gradient-flow in the Wasserstein space :  
 estimates in the time-discrete scheme

GRAD. FLOWS

$$\begin{cases} x'(t) = -\nabla F(x(t)) \\ x(0) = x_0 \end{cases} \quad F: \mathbb{R}^n \rightarrow \mathbb{R}$$

TIME-DISCR. For  $\tau > 0$   $x_0, x_1, x_2, \dots, x_k \approx x(k\tau)$

DEFINED VIA

$$x_{k+1} \in \operatorname{argmin} F(x) + \frac{|x - x_k|^2}{2\tau}$$

$$\nabla F(x) + \frac{x - x_k}{\tau} = 0$$

THIS APPROACH CAN BE EXTENDED TO METRIC SPACES

$$\min F(x) + \frac{d^2(x, x_k)}{2\tau}$$

no case in

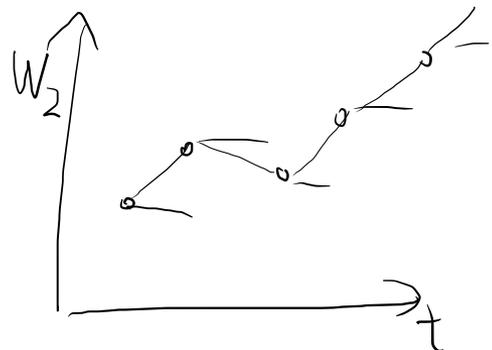
$$(X, d) = (B(\Omega), W_2)$$

$$t \mapsto p_t$$

$$(t, x) \mapsto p(t, x)$$

$$p_{\text{min}} \in \text{argmin } F(p) + \frac{W_2^2(p, p_2)}{2\tau}$$

$$F: B(\Omega) \rightarrow \bar{\mathbb{R}}$$



This scheme is called JKO scheme

FEW FACTS ABOUT O.T.

given  $\mu, \nu \in \mathcal{P}(\Omega)$   $\Omega \subset \mathbb{R}^d$

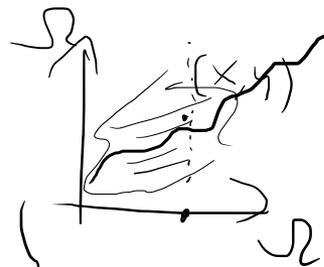
$$\min \left\{ \int_{\Omega} \frac{1}{2} |x - T(x)|^2 d\mu : T \# \mu = \nu \right\}$$

$$\min \left\{ \int_{\Omega \times \Omega} \frac{1}{2} |x - y|^2 d\gamma : \gamma \in \Pi(\mu, \nu) \right\}$$

$$\Pi(\mu, \nu) = \left\{ \gamma \in \mathcal{P}(\Omega \times \Omega) : \right.$$

$$\Pi_x \# \gamma = \mu$$

$$\Pi_y \# \gamma = \nu \left. \right\}$$



$$\max_{\gamma \in \Pi(\mu, \nu)} \left\{ \int \varphi d\mu + \int \psi d\nu : \varphi(x) + \psi(y) \leq \frac{1}{2} |x - y|^2 \right\}$$

FACTS  $\mathbb{R}^d \ll \mathbb{R}^d$

$$\exists! \gamma_{opt} \quad \gamma = (id, T) \# \mu$$

$$\exists \text{ optimal } \varphi \in Lip$$

$$T(x) = x - \nabla \varphi(x) = \nabla \mu \quad \mu \text{ concave}$$

$$\frac{|x|^2}{2} - \varphi \quad \text{convex}$$

$$I - D^2 \varphi \geq 0$$

With OT we can define a distance

$$W_2(\mu, \nu) = \sqrt{\min \left\{ \int |x-y|^2 d\gamma \cdot \gamma \in \Pi(\mu, \nu) \right\}}$$

$$\frac{1}{2} W_2^2(\mu, \nu) = \max \left\{ \int \varphi d\mu + \int \psi d\nu : \varphi \oplus \psi \leq \frac{1}{2} |x-y|^2 \right\}$$

It is a distance on  $\mathcal{P}(\Omega)$  (if  $\Omega$  is compact)  
which metrizes  $\star \rightarrow$

Optimality conditions of

$$\min F(\beta) + \frac{W_2^2}{2L}(\beta, \beta_u)$$

Notation:  $\frac{\delta F}{\delta \beta}$  = First variation of  $F$  defined via

Examples

$$F(\beta) = \int V d\beta \Rightarrow \frac{\delta F}{\delta \beta} = V$$

$$F(\beta) = \int f(\beta(x)) dx \Rightarrow \frac{\delta F}{\delta \beta} = f'(\beta)$$

$$\left. \frac{d}{d\varepsilon} F(\beta + \varepsilon \chi) \right|_{\varepsilon=0} = \int \frac{\delta F}{\delta \beta} d\chi$$

$\frac{\delta F}{\delta \beta}$  defined up to add. const.

$$\frac{\delta F}{\delta \beta} + \frac{1}{L} \frac{\delta}{\delta \beta} \left( \frac{1}{2} W_2^2(\cdot, \beta_u) \right) = C$$

$$\frac{\delta}{\delta g} \frac{1}{2} W_2^2(\cdot; v) = \varphi$$

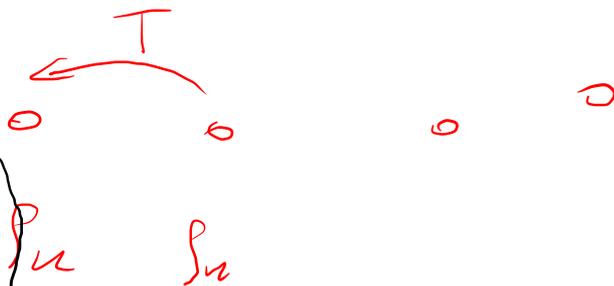
fixed

← optimal function  
(cont. potential)  
in this respects  
from  $g$  to  $v$

$$\Rightarrow \nabla \frac{\delta F}{\delta g} + \frac{\nabla \varphi}{T} = 0$$

||  
DIVERGENCE  
 $v$

$$T \# \rho_{k+1} = \rho_k$$



The PDE which I mention for  $v$

$$\mathcal{L}_t \rho + \nabla \cdot (\rho v) = 0$$

$$v = - \nabla \frac{\delta F}{\delta g}$$

$$\lambda \rho + \nabla \cdot (\rho v) = 0$$

$$v = -\nabla \frac{\delta F}{\delta \rho}$$

example

$$F(\rho) = \int f(\rho(x)) dx$$

$$\frac{\delta F}{\delta \rho} = f'(\rho)$$

$$\lambda \rho - \nabla \cdot (\rho \nabla f'(\rho)) = 0$$

$$\lambda \rho - \nabla \cdot (\rho f''(\rho) \nabla \rho) = 0$$

$$\text{if } f(\rho) = \rho \log \rho \quad f'(\rho) = \log \rho + 1 \quad f''(\rho) = \frac{1}{\rho}$$

$\Rightarrow$  heat equation

(now mult. heat eq. is also the equation of  $F(\rho) = \frac{1}{2} \int |\rho_p|^2$   
(if 1. Riesz  $W_2$  norm  $L^2$ )

Fokker-Planck

$$\lambda f - \Delta f - \nabla \cdot (f \nabla V) = 0 \quad V: \Omega \rightarrow \mathbb{R}$$

and then in  $W_2$  or  $F(f) = \int f \log f + fV$

$$\frac{\delta F}{\delta f} = \log f + V$$

$$\lambda f - \nabla \cdot (f \nabla (\log f + V)) = 0$$

(Theorem by JKO)

ESTIMATES on TAKING PLUCK via JKO

$$F(\rho) = \int f(\rho) + \rho V \quad \min F(\rho) + \frac{W_2^2(\rho, g)}{2\epsilon}$$

$V \in \text{Lip}$        $f$  convex

$$f'(\rho) + V + \frac{\rho}{\epsilon} = c$$

$$T_{\#} \rho = g \quad T(x) = x - \alpha \varphi(x) \quad \begin{matrix} I - D\varphi \\ \parallel \end{matrix}$$

$$\rho(x) = g(T(x)) \text{ dLTA } DT(x)$$

push-forward of  $g$ .

SUPPOSE (for a while)  $g \in C^{0,\alpha}$   $\varphi \in \text{Lip}$   
BDD BOUNDED NON INCR

$$\Rightarrow f'(\rho) \in \text{Lip}$$

$$\Rightarrow \rho \in \text{Lip} \quad | \text{log } \rho \in \text{Lip}$$

$$\Rightarrow \rho \in \text{Lip} \\ 0 < a \leq \rho \leq b < +\infty$$

CAMPANELLI'S REGULARITY FOR P.D.

$$g, \rho \in C^{0,\alpha} \text{ BDD BOUNDED} \Rightarrow \varphi \in C^{2,\alpha}$$

$\Omega$  convex, smooth

$L^\infty$  Bounds on  $f$

$$V=0$$

$$x_0 \in \text{argmax } f \Rightarrow \text{argmax } f'(p)$$

$$\Rightarrow \min \varphi \Rightarrow \begin{aligned} & \text{(if } x_0 \notin \Omega) \\ & \nabla \varphi = 0 \\ & D^2 \varphi \geq 0 \end{aligned}$$

$$f(x_0) = g(\underbrace{T(x_0)}_{x_0}) \det(\underbrace{I - D^2 \varphi(x_0)}_{1})$$

$$\Rightarrow \underbrace{f(x_0)}_{\|f\|_\infty} \leq \underbrace{g(x_0)}_{\|g\|_\infty} \leq \|g\|_\infty$$

$$\|f\|_\infty$$

$$\text{If } V \neq 0 \quad \text{Then } x_0 \in \text{argmax } f'(p) + V \Rightarrow \varphi \text{ min}$$

$$f(x_0) = g(x_0) \circ \text{Lte} \left( \underset{\substack{\uparrow \\ \uparrow}}{\Gamma - D\psi} \right)$$

$$\Rightarrow f'(f(x_0)) \in f'(g(x_0))$$

$$(f'(f) + v)(x_0) \in (f'(g) + v)(x_0) \in \|f'(g)\|_{L^\infty}$$

$$k \mapsto \|f'(f_k) + v\|_{L^\infty} \searrow$$

DIPPER:  $x_0 \in \partial\Omega$

if  $\min \varphi$  ATTAINS on  $\partial\Omega$



$$\tau(x_0) = x_0 - \nabla\varphi(x_0) \in \Omega \Rightarrow \nabla\varphi(x_0) \cdot n \geq 0$$

$\Rightarrow$  NOT POSSIBLE to have  $\min \varphi$  on  $\partial\Omega$

What about min  $\int$

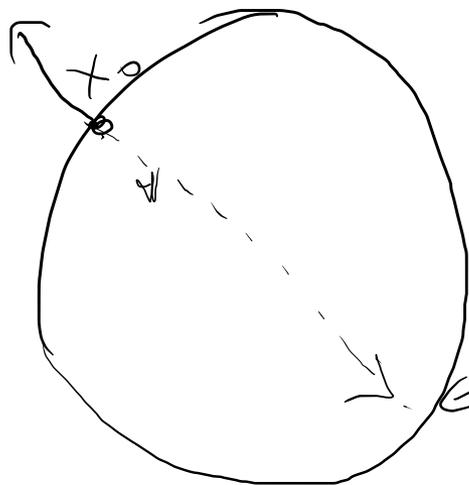
very same result, but we have to exclude

$$\exists \gamma \max x \varphi \in \partial \Omega$$

If  $\int, \gamma$  smooth  $T$  smooth  $T$  is a diffeomorphism:  $\Omega \rightarrow \Omega$

$$T(\partial \Omega) = \partial \Omega$$

$\nabla \varphi(x_0)$



$$T(x_0) = x_0 - \nabla \varphi(x_0) \in \partial \Omega$$

if  $T = \partial \Omega$  was nonempty

$L^\infty$  bounds on  $\nabla(\log S + V)$  via F-P

$$\nabla(\log S + V) + \frac{\gamma \varphi}{L} = c$$

$$\log S = \log S_0^T + \log(\det(I - D^t \varphi))$$

$$\frac{\partial}{\partial x_i} (\log S)_i = (\log S)_j T_j^i + B^{jk} \varphi_{jki} \quad \leftarrow \sum_{j,k}$$

$$B = (I - D^t \varphi)^{-1}$$

$$\max \frac{|\gamma \varphi|^2}{2} \Rightarrow D^t \varphi \varphi = 0$$

$$\varphi_{i,jk} \varphi_i + \varphi_{ij} \varphi_{i,k} \leq 0 \quad (D^t \varphi)^2$$

$$(\log g)_i = (\log g)_j T_j^i + B^{jk} \varphi_{jki}$$

Thus  $\varphi_i$ :

$$\begin{aligned} \nabla \varphi = \nabla \log g &= \nabla \log g - (I - D^2 \varphi) \nabla \varphi \\ &\quad - \underbrace{B^{ja} \varphi_{ia} \varphi_i}_{\leq 0} \end{aligned}$$

$$\nabla \varphi = \nabla \log g \geq \nabla \log g^{(1)} \cdot \nabla \varphi$$

$V = 0$

$$\nabla \log g = -\frac{\nabla \varphi}{L} - \frac{10\varphi^2}{L} \geq \nabla \log g = \nabla \varphi$$

$$|\operatorname{div} g| = \frac{|\nabla \varphi|^2}{\Gamma} \leq |\nabla \log g| + |\nabla \varphi|$$

$$\| \nabla \log g \|_{L^\infty} \quad \kappa \longmapsto \| \nabla \log g_\kappa \|_{L^\infty}$$

(IF  $V = 0$  IF not add  $\nabla U$   
 (mod the  $\mathcal{D}^1 U \geq \lambda I$ )

OF course one needs to exclude the  
 case  $x_0 \in \partial \Omega$

Prop IF  $\varphi, g$  smooth  $\varphi$  is a pot  $\Omega = B(o, R)$   
 $\max |\nabla \varphi|^2$  is never attained on  $\partial \Omega$

$$|x - \nabla \varphi(x)|^2 = R^2 \quad \text{for } |x| = R \quad \Rightarrow \quad x \cdot \nabla \varphi = \frac{1}{2} |\nabla \varphi|^2$$

Thm

• IF  $a \leq \log f_0 + V \leq b \Rightarrow \forall u$

$$a \leq \log f_u + V \leq b$$

(IF  $\Omega$  is convex, and  $V$ )

• IF  $\text{LIP}(\log f_0 + V) \leq M$

IF  $\Omega$  is convex  $D^2 V \geq \lambda I$

$\Rightarrow \text{LIP}(\log f_u + V) \leq M(1 - \lambda t)^k$

$$\text{LIP}(\log f_t + V) \leq \text{LIP}(\log f_0 + V) e^{-\lambda t}$$