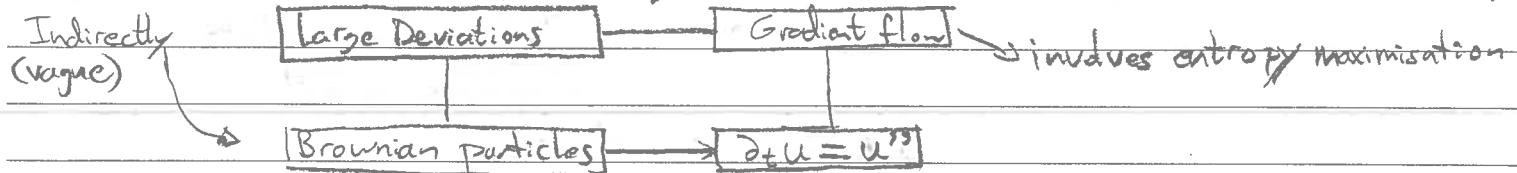


1

A particle systems method to derive Wasserstein gradient flow structures

Not about connection particle systems and differential equations... not directly!



1) discuss connections

2) exploit connection to derive gradient flow structures

Gradient flow - diff. eq.

(time-discrete schemes)

1²] sequence $(\rho^j)_{j=1}^\infty$: Given ρ^{j-1} , let ρ^j be minimiser of :

$$\rho \mapsto \frac{1}{2} \int_{\mathbb{R}} |\rho'|^2 dx + \frac{1}{2\tau} \int_{\mathbb{R}} \|\rho - \rho^{j-1}\|_2^2 dx \quad (\text{in } H^1(\mathbb{R}))$$

(in $H^1(\mathbb{R})$)

energy small time step metric

$$\text{Euler-Lagrange: } -\rho_t^{j+1} + \frac{\rho^j - \rho^{j-1}}{\tau} = 0$$

$$\downarrow \tau \rightarrow 0 \quad \partial_t u = u'' = -\text{grad}_{L^2} E(u)$$

Wasserstein

$$S(\rho) := \begin{cases} \int p(x) \log p(x) dx, & \rho(dx) = p(x)dx \\ \infty & \text{otherwise} \end{cases}$$

& Wasserstein distance on $\mathcal{P}_2(\mathbb{R}^d)$

Given ρ^{j-1} , let ρ^j be minimiser of:

$$\boxed{\rho \mapsto \frac{1}{2} S(\rho) - \frac{1}{2} S(\rho^{j-1}) + \frac{1}{4\tau} d(\rho, \rho^{j-1})^2} \quad (\mathcal{P}_2)$$

$$\text{Euler Lagrange: } \text{grad}_\rho S(\rho^j) + \frac{\rho^j - \rho^{j-1}}{\tau} = 0$$

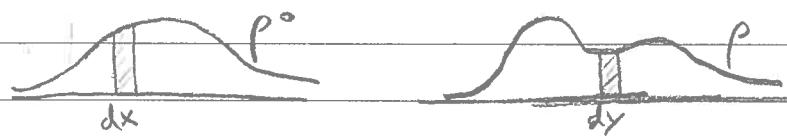
$$\downarrow \tau \rightarrow 0 \quad (\star) \text{ Jordan, Kinderlehrer, Otto 1998} \quad \partial_t u = \text{grad}_\rho S(u) = u''$$

• Two competing forces.

• $\rho^{j-1} \rightarrow \rho^0$ (only intended in first transition from ρ^0 to ρ^1)

12:00

12 Wasserstein distance



$\gamma \in \mathcal{P}(\mathbb{R} \times \mathbb{R})$ transports p^o to p if:

$$\gamma(B \times \mathbb{R}) = p^o(B)$$

$$\gamma(\mathbb{R} \times B) = p(B)$$

$\Gamma(p^o, p)$ set of transport plans γ from p^o to p

$\gamma(dx dy)$ measures how much mass is transported from dx to dy

If "cost" to transport dx to dy is $|y - x|^2$, what is the minimal cost to transport p^o to p ?

$$d(p^o, p)^2 := \inf_{\gamma \in \Gamma(p^o, p)} \iint |y - x|^2 \gamma(dx dy)$$

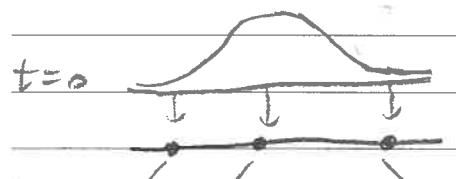
$$= \inf \left\{ \int_0^1 \| \partial_t \nu^t \|_{L^2}^2 dt : \text{curves } \nu^t \text{ s.t. } \nu^0 = p^o \text{ and } \nu^1 = p \right\}$$

$(\nu^t \in AC((0,1); P_2(\mathbb{R})))$

15:00

$$\| \partial_t \nu^t \|_{L^2}^2 = \int_{\mathbb{R}} |v^t|^2 dx, \quad \partial_t v^t + \operatorname{div}(v^t v^t) = 0$$

13 Brownian Particle system - diff. eq.



$p^o \in \mathcal{P}(\mathbb{R})$ initial distribution

$X_1(0), X_2(0), \dots$ iid with law p^o

$$\theta^t(x, y) = \frac{1}{\sqrt{4\pi t}} \exp\left(-\frac{(y-x)^2}{4t}\right) \text{ Brownian transition prob.}$$



$$L_n(t) := \frac{1}{n} \sum_{i=1}^n \delta_{X_i(t)}$$



$$L_n(t) \xrightarrow{n \rightarrow \infty} \mathbb{E}[L_n(t)] = p^o * \theta^t \quad \text{as } n \rightarrow \infty$$

solution of diff. eq

Prob($L_n(t) \in C$) \rightarrow $\begin{cases} 1, & p^o * \theta^t \in C \\ 0, & \text{otherwise} \end{cases}$ for all $C \subseteq \mathcal{P}(\mathbb{R})$

meas. $\xrightarrow{\text{cont. sets}}$

\rightarrow two topologies

120:00

14) Brownian particles - Large Deviations

How fast converge to 0 or 1?

$$\text{Prob}(L_n(\tau) \approx p | L_0(\tau) \approx p^0) \sim \exp(-n J^\tau(p|p^0)) \quad \text{as } n \rightarrow \infty$$

Th Léonard 2007, Peletier-R 2011

$$J^\tau(p|p^0) = \inf_{\gamma \in \Gamma(p^0, p)} H(\gamma | p^0 \otimes \delta^\tau) \quad \text{with} \quad H(\nu | \mu) = \begin{cases} \int \int \log \left(\frac{d\nu}{d\mu} \right) d\mu & \mu \ll 2 \\ \infty & \text{else} \end{cases}$$

Th Dawson-Gärtner 1987, Kipnis-Olla 1996

$$J^\tau(p|p^0) = \inf \left\{ \frac{1}{4} \int_0^\tau \| \partial_t \gamma^\tau - \nu^\tau \|^2 ds \mid \gamma^\tau \text{ curves with } \gamma^0 = p^0, \gamma^\tau = p \right\}$$

5) Large Deviations - Gradient flows

→ only has to work for small τ .

Th Adams, Dirr, Peletier, Zimmer 2010; Duong, Laschos, R. 2012

$$J^\tau(p|p^0) \approx \frac{1}{2} S(p) - \frac{1}{2} S(p^0) + \frac{1}{4\tau} d(p^0, p)^2 \quad \text{as } \tau \rightarrow 0$$

(Development in terms of Γ -convergence)

Summary

Conclusion: entropy-WS minimisation = likelihood maximisation

30.09

6) New gradient flow structures

How to find GF-structure for different PDE?

- ① Describe particle system by transition prob.
- ② Calculate conditional LDP $J^\tau(p|p^0)$
- ③ Asymptotic development of J^τ for small τ
- ④ Prove GF-structure

(A) Fokker-Planck

$$\partial_t u = u'' + \operatorname{div}(u \Psi') \text{ for some potential } \Psi$$

- transition prob: fundamental solution

- Asympt. development LDP:

Th (Pelletier-R 2011, Duong, Laschos-R 2012)

$$J^t(p|p^0) \approx \frac{1}{2} S(p) + \frac{1}{2} \int \Psi dp - \frac{1}{2} S(p^0) - \frac{1}{2} \int \Psi dp^0 + \frac{1}{4t} d(p^0, p)^2$$

- prove GF structure:

(Jordan, Kinderlehrer, Otto 1994)

(B) Diffusion with bulk decay

$$\partial_t u = u'' - \lambda u, \text{ for some } \lambda > 0$$

Not mass-conserving \Rightarrow add 'decayed matter' and assume it continues its Brownian motion. State space = $\mathbb{R} \times \{N, D\}$

- transition prob: Brownian motion - σt

indep. (Decay): $\operatorname{Prob}(N \rightarrow N) = e^{-\lambda t}$

$$\operatorname{Prob}(N \rightarrow D) = 1 - e^{-\lambda t}$$

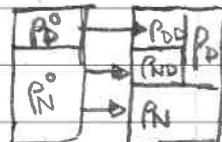
$$\operatorname{Prob}(D \rightarrow N) = 0$$

$$\operatorname{Prob}(D \rightarrow D) = 1$$

normal \uparrow dark \uparrow

- Asympt. development LDP. Only works if you know

Th Pelletier-R 2011



$$J^t(p_N, p_{ND}, p_D | p_N^0, p_D^0) \approx \frac{1}{2} S(p_N + p_D) - \frac{1}{2} S(p_N^0) + \frac{1}{4t} d(p_N^0, p_N + p_D)^2$$

$$+ \frac{1}{2} S(p_{ND}) - \frac{1}{2} S(p_D^0) + \frac{1}{4t} d(p_D^0, p_D)^2$$

$$- S(p_N + p_D + S(p_N) + S(p_{ND})) - \|p_N\| \log(e^{-\lambda t}) - \|p_{ND}\| \log(1 - e^{-\lambda t})$$

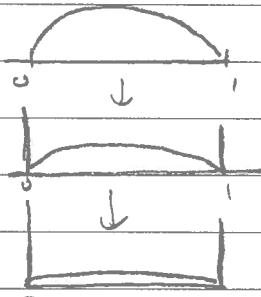
drift diffusion

drives decay; minimal if $p_N = e^{-\lambda t} (p_N^0 + p_{ND})$

C) Diffusion with decay at boundary

$$\begin{cases} \partial_t u = u'' \text{ in } (0,1) \\ u(0) = u(1) = 0 \end{cases}$$

Not mass-conserving \rightarrow add 'decayed' mass at boundary:



- transition prob: Brownian motion with "sticking boundary"

- work in progress... it seems that

$$J^*(\rho|\rho^*) \approx \inf_{\substack{\rho^*_{\text{rel}} + \rho^*_{\text{int}} + \rho^*_{\text{far}} \\ ||\rho^*_{\text{null}}|| = ||\rho||}} \left\{ \frac{1}{2} S(\rho) - \frac{1}{2} S(\rho^*) + S(\rho_{\text{rel}}) + S(\rho_{\text{int}}) + S(\rho_{\text{far}}) - S(\rho^*) \right\}$$

$$- \inf \left\{ \iint_{[0,1]^2} |y-x|^2 \gamma(dx dy) : \gamma \in \mathcal{P}([0,1] \times [0,1]) : \gamma(B \times [0,1]) = \rho^*(B) \right. \\ \left. \gamma([0,1] \times B) = \rho(B), B \subset [0,1] \right. \\ \left. \gamma([0,1] \times [0,1]) = 0 \right\}$$

- similar to metric proposed in Figalli-Gigli 2011.

- only this one is asymmetric!

Conclusion

- This method can provide new Wasserstein gradient flow structures.
- Thus found gradient flows are meaningful as a measure for unlikelihood.