

Lecture 22, Monday, 5/16/2022

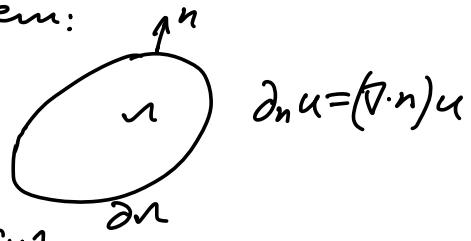
Today: Convergence of the JKO scheme.

The heat equation $\partial_t u = \Delta u$ $u=u(x,t)$

Given: $\mathcal{V} \subseteq \mathbb{R}^d$, bounded and smooth domain
 $u_0 = u_0(x)$, bounded and smooth function.

The initial-boundary-value problem:

$$\begin{cases} \partial_t u = \Delta u & \mathcal{V} \times (0, \infty) \\ \partial_n u = 0 & \partial \mathcal{V} \times (0, \infty) \\ u(\cdot, 0) = u_0 & \end{cases}$$



The gradient flow. $\partial_t u = -\delta I[u]$.

$$I[v] := \int_{\mathcal{V}} \frac{1}{2} |\nabla v|^2 dx, \quad \delta I[v] = -\Delta v.$$

The implicit Euler scheme Let $\tau > 0$, $t_k = k\tau$.

$$u_k = u_k^\tau \approx u(\cdot, t_k) : \begin{cases} u_0 \text{ given} \\ \frac{u_{k+1} - u_k}{\tau} = \Delta u_{k+1} \quad (k=0, 1, \dots) \end{cases}$$

$$u_{k+1} = \arg \min_u \left(\frac{1}{2\tau} \|u - u_k\|_{L^2(\mathcal{V})}^2 + I[u] \right), \text{ unique}$$

The JKO scheme Define

$$F(p) = \int_{\mathcal{V}} p \log p dx, \quad p=p(x) \geq 0, \quad x \in \mathcal{V}.$$

$$\mathcal{P}(\mathcal{V}) = \left\{ p \in L^1(\mathcal{V}) : p \geq 0, \int_{\mathcal{V}} p dx = 1 \right\}. \Rightarrow \int_{\mathcal{V}} |x|^2 p(x) dx < \infty$$

Given: $p_0 \in \mathcal{P}(\mathcal{V})$, $\tau > 0$, $t_k = k\tau$, $k=0, 1, 2, \dots$.

$$p_{k+1} = \arg \min_p \left(\frac{1}{2\tau} W_2^2(p, p_k) + F[p] \right)$$

Note: for $\xi, \eta \in \mathcal{P}(\mathcal{V})$, $W_2(\xi, \eta) = \sqrt{\int_{\mathcal{V}} (\xi - \eta)^2 dx}$.

$X = \bar{\mathbb{R}}$: compact, Euclidean distance.

Convergence: up to a subseq. $\rho^T \xrightarrow{T} \rho$ in W_2 and $\partial_t \rho = \Delta \rho$.

The Fokker-Planck eq. (FPE):

$$\partial_t \rho = \beta^{-1} \Delta \rho + \nabla \cdot (\rho \nabla V)$$

$V: \mathbb{R}^d \rightarrow \mathbb{R}$ given, bounded below.

Known: ρdx is the probability distribution for the stochastic process $X_t \in \mathbb{R}^d$:

$$dX_t = -\nabla V(X_t) dt + \sqrt{2\beta^{-1}} dW_t$$

Here: W_t is the standard d -dimensional Brownian motion.

To day: Proof of convergence $\rho^T \xrightarrow{W_2} \rho$.

We have proved (cf. last lecture): Given ρ_κ

$$\exists ! \rho_{\kappa+1} = \arg \min_{\rho \in \mathcal{P}(\mathbb{R}^d)} \left(\frac{1}{2T} W_2^2(\rho, \rho_\kappa) + F[\rho] \right),$$

where $F[\rho] = \int \rho \log \rho dx$.

Brenier's Theorem Let $X = \mathbb{R}^d$ and $\mu, \nu \in \mathcal{P}_2(X)$ with $d\mu = g dx$ for some $g \in L^1(\mathbb{R}^d)$. Then, $\exists ! \tilde{f} \in \mathcal{E}(\mu, \nu)$

$$\text{s.t. } W_2^2(\mu, \nu) = \int_{X \times X} \frac{1}{2} |x-y|^2 d\tilde{f}(x, y). \text{ Moreover,}$$

$$\tilde{f} = (Id \times T)^\# \mu \text{ and } T = \nabla \varphi \text{ for some convex } \varphi.$$

We will prove this theorem when we study the duality.

The minimizer \hat{P} satisfies some minimality condition.

Lemma Let P_0, \hat{P} be the same as above. Assume:

(1) $\xi \in C^\infty(\mathcal{N}, \mathbb{R}^d)$ is tangent to $\partial\mathcal{N}$;

(2) $T: \mathcal{N} \rightarrow \mathcal{N}$ is the optimal map from P_0 to \hat{P} .

Then $\int_{\mathcal{N}} \hat{P} \nabla \cdot \xi \, dx = \frac{1}{2\pi} \int_{\mathcal{N}} \langle \xi \circ T, T - Id \rangle P_0 \, dx$.

The idea is to perturb \hat{P} and use the minimizer property of \hat{P} . This is similar to deriving the Euler-Lagrange eq. But the perturbation is done quite differently. Sometimes, the method is called the method of domain variations.

Proof of Lemma Consider $\begin{cases} \dot{\Phi}(t, x) = \xi(\Phi(t, x)), \\ \Phi(0, x) = x. \end{cases}$

Since ξ is tangential to $\partial\mathcal{N}$, $\underline{\Phi}(t): \mathcal{N} \rightarrow \mathcal{N}$ is a diffeomorphism. Define $P_\varepsilon = \underline{\Phi}(\varepsilon) \# \hat{P} \in \mathcal{P}(\mathcal{N})$.

[Note that $\varepsilon \in \mathbb{R}$ can be positive, negative, or 0. Note also that $\underline{\Phi}(\varepsilon, x) = x + \varepsilon \xi(x) + O(\varepsilon^2)$ if $|\varepsilon| \ll 1$.] We have

$$\hat{P}(x) = P_\varepsilon(\underline{\Phi}(\varepsilon, x)) \det \nabla \underline{\Phi}(\varepsilon, x). \quad (*)$$

$$\begin{aligned}
\text{So, } \int_{\mathbb{R}} p_{\varepsilon}(y) \log p_{\varepsilon}(y) dy &= \int_{\mathbb{R}} \log p_{\varepsilon}(y) p_{\varepsilon}(y) dy \\
p_{\varepsilon} = \Phi(\varepsilon) \# \hat{P} &\stackrel{*}{=} \int_{\mathbb{R}} \log p_{\varepsilon}(\Phi(\varepsilon, x)) \hat{P}(x) dx \\
&\stackrel{(*)}{=} \int_{\mathbb{R}} \hat{P}(x) \log \frac{\hat{P}(x)}{\det \nabla \Phi(\varepsilon, x)} \\
&= \int_{\mathbb{R}} \hat{P}(x) \log \hat{P}(x) dx - \int_{\mathbb{R}} \hat{P}(x) \log \det \nabla \Phi(\varepsilon, x) dx \\
&= \int_{\mathbb{R}} \hat{P}(x) \log \hat{P}(x) dx - \int_{\mathbb{R}} \hat{P}(x) \log (1 + \varepsilon \nabla \cdot \xi + o(\varepsilon)) dx \\
&= \int_{\mathbb{R}} \hat{P}(x) \log \hat{P}(x) dx - \varepsilon \int_{\mathbb{R}} \hat{P}(x) \nabla \cdot \xi dx + o(\varepsilon).
\end{aligned}$$

Now, let $\gamma \in \mathcal{A}(\hat{P}, P_0)$ s.t. $W_2^2(\hat{P}, P_0) = \int_{\mathbb{R} \times \mathbb{R}} |x-y|^2 d\gamma(x, y)$.

Set $\gamma_{\varepsilon} = (\Phi(\varepsilon, \cdot) \times \text{Id}) \# \gamma \in \mathcal{A}(P_{\varepsilon}, P_0)$. Since $\Phi(\varepsilon, x) = x + \varepsilon \xi(x) + o(\varepsilon)$, we get

$$\begin{aligned}
W_2^2(P_{\varepsilon}, P_0) &\leq \int_{\mathbb{R} \times \mathbb{R}} |x-y|^2 d\gamma_{\varepsilon}(x, y) = \int_{\mathbb{R} \times \mathbb{R}} |\Phi(\varepsilon, x) - y|^2 d\gamma(x, y) \\
&= \int_{\mathbb{R} \times \mathbb{R}} [|x-y|^2 + 2\varepsilon \langle \xi(x), x-y \rangle + o(\varepsilon)] d\gamma(x, y) \\
&= W_2^2(\hat{P}, P_0) + 2\varepsilon \int_{\mathbb{R} \times \mathbb{R}} \langle \xi(x), x-y \rangle d\gamma(x, y) + o(\varepsilon).
\end{aligned}$$

Therefore, since \hat{P} is a minimizer,

$$\begin{aligned}
\frac{W_2^2(\hat{P}, P_0)}{2\varepsilon} + \int_{\mathbb{R}} \hat{P} \log \hat{P} dx &\leq \frac{W_2^2(P_{\varepsilon}, P_0)}{2\varepsilon} + \int_{\mathbb{R}} P_{\varepsilon} \log P_{\varepsilon} dx \\
&\leq \frac{W_2^2(\hat{P}, P_0)}{2\varepsilon} + \frac{\varepsilon}{\tau} \int_{\mathbb{R} \times \mathbb{R}} \langle \xi(x), x-y \rangle d\gamma(x, y) \\
&\quad + \int_{\mathbb{R}} \hat{P} \log \hat{P} dx - \varepsilon \int_{\mathbb{R}} \hat{P} \nabla \cdot \xi dx + o(\varepsilon),
\end{aligned}$$

leading to

$$\varepsilon \left[\frac{1}{\tau} \int_{X \times X} \langle \xi(x), x-y \rangle d\pi(x,y) - \int_X \tilde{\rho}^* \nabla \cdot \xi dx \right] + o(\varepsilon) \geq 0.$$

But $\Sigma > 0$, $\Sigma < 0$ possible. So,

$$\int_X \tilde{\rho}^* \nabla \cdot \xi dx = \frac{1}{\tau} \int_{X \times X} \langle \xi(x), x-y \rangle d\pi(x,y).$$

If T is the optimal transport map from ρ_0 to $\tilde{\rho}^*$, with $\gamma = (T \times \text{Id}) \# \rho_0$, then

$$\int_{X \times X} \langle \xi(x), x-y \rangle d\gamma(x,y) = \int \langle (\xi \circ T)(x), T(x)-x \rangle \rho_0(x) dx.$$

Hence, $\int_X \tilde{\rho}^* \nabla \cdot \xi dx = \frac{1}{\tau} \int \langle \xi \circ T, T - \text{Id} \rangle \rho_0 dx$. QED

Theorem Let $\Omega \subseteq \mathbb{R}^d$ be a bounded and smooth domain. Let $\tau > 0$ and $t_k = k\tau$ ($k=0, 1, \dots$) Assume $\rho_0 \in \mathcal{P}(\Omega)$ and $F[\rho_0] = \int_{\Omega} \rho_0 \log \rho_0 dx < \infty$.

Let $\rho_{k+1}^{\tau} \in \mathcal{P}(\Omega)$ be the unique minimizer of

$$J_k[\rho] = \frac{1}{2\tau} W_2^2(\rho, \rho_k) + F[\rho]$$

among all $\rho \in \mathcal{P}(\Omega)$, $k=0, 1, \dots$. Let $\rho^{\tau}(t) = \rho_k^{\tau}$ if $t_k \leq t < t$. Then there exist $\rho \in L^1_{loc}([0, \infty) \times \Omega)$ and a subsequence of $\{\rho^{\tau}\}_{\tau > 0}$, not relabeled, such that $\rho^{\tau} \rightarrow \rho$ weakly in $L^1_{loc}([0, \infty) \times \Omega)$. Further, ρ is the distributional solution to the initial-boundary-value problem of the heat equation.

Sketch of Proof Need to show that

$$\int \varphi \frac{\rho^\tau - \rho_0}{\tau} dx = \int \Delta \varphi \rho^\tau + o(\tau) \quad \forall \varphi \in C_c^\infty(\Omega).$$

By the above lemma, with $\zeta = \nabla \varphi$.

$$\int \rho^\tau \Delta \varphi dx = \frac{1}{\tau} \int \langle T - Id, \nabla \varphi \rangle \rho^\tau dx. \quad (**)$$

Let $T \# \rho^\tau = \rho_0$. Then

$$\begin{aligned} \int \varphi \frac{\rho^\tau - \rho_0}{\tau} dx &= \frac{1}{\tau} \int \varphi \rho^\tau dx - \frac{1}{\tau} \int \varphi \rho_0 dx \\ &= \frac{1}{\tau} \int \varphi \rho^\tau dx - \frac{1}{\tau} \int (\varphi \circ T) \rho^\tau dx \\ &= -\frac{1}{\tau} \int (\varphi \circ T - \varphi) \rho^\tau dx \\ &= -\frac{1}{\tau} \int \int_0^1 \langle \nabla \varphi((1-t)x + tT(x)), T(x) - x \rangle dt \rho^\tau(x) dx \\ &= -\frac{1}{\tau} \int \int_0^1 \langle \nabla \varphi(x), T(x) - x \rangle \rho^\tau(x) dx + R(\tau) \\ &\stackrel{(**)}{=} \int \Delta \varphi \rho^\tau + R(\tau). \end{aligned}$$

Moreover,

$$\begin{aligned} |R(\tau)| &\leq \frac{L_{\gamma^1}(\nabla \varphi)}{\tau} \int \int_0^1 t |T(x) - x|^2 dt \rho^\tau(x) dx \\ &= \frac{L_{\gamma^1}(\nabla \varphi)}{2\tau} W_2^2(\rho_0, \rho_\tau) = o(\tau). \quad \underline{\text{QED}} \end{aligned}$$