

Lecture 21, Friday, 5/13/2022

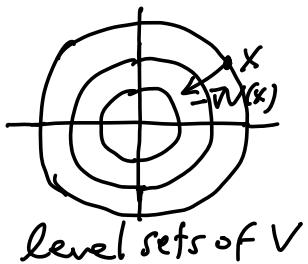
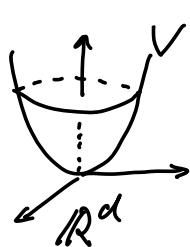
① Finish the proof of convergence theorem

② The TKO scheme

The steepest descent / gradient descent

$$x = x(t) \in \mathbb{R}^d : \dot{x} = -\nabla V(x). \text{ i.e. } \frac{dx(t)}{dt} = -\nabla V(x(t))$$

$V: \mathbb{R}^d \rightarrow [0, \infty)$  given, smooth, convex, ...



$$\frac{d}{dt}[V(x(t))] = \nabla V(x(t)) \cdot \dot{x}(t) \\ = -\|\nabla V(x(t))\|^2 \leq 0.$$

$\forall x \in \mathbb{R}^d : \nabla V(x) \perp \{y \in \mathbb{R}^d : V(y) = V(x)\}$ .

$\|\nabla V(x), \vec{v}\|$  ( $\vec{v}$ : unit vector) is maximized at  $\pm \nabla V(x) / \|\nabla V(x)\|$ .

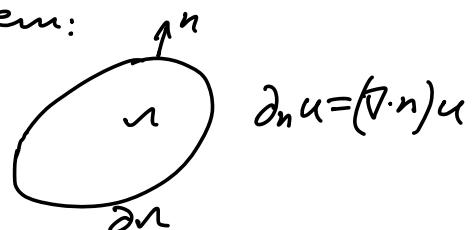
$\Rightarrow$  steepest descent of  $V(x(t))$ .

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

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

The initial-boundary-value problem:

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



③ Conservation of mass:

$$\frac{d}{dt} \int_{\Omega} u dx = \int_{\Omega} \partial_t u dx = \int_{\Omega} \Delta u dx = \int_{\partial \Omega} \partial_n u ds = 0$$

$$\Rightarrow \int_{\Omega} u dx = \text{const.} = \int_{\Omega} u_0 dx \quad \forall t \geq 0.$$

④ This is the gradient flow.  $\partial_t u = -\delta I[u]$ .

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

$$\delta I[v][\varphi] = \left. \frac{d}{dt} I[v + t\varphi] \right|_{t=0} = \left. \frac{d}{dt} \right|_{t=0} \int_{\Omega} \frac{1}{2} |\nabla v + t \nabla \varphi|^2 dx$$

$$= \int_{\Omega} \nabla v \cdot \nabla \varphi dx = - \int_{\Omega} \Delta v \varphi dx = \langle -\Delta v, \varphi \rangle_{L^2(\Omega)}, \quad \forall \varphi \in C^1_c(\bar{\Omega}).$$

Identify  $\delta I[v] = -\Delta v$ .

Note: This is the gradient flow w.r.t.  $L^2(\Omega)$ .

The implicit Euler scheme Let  $\tau > 0$  be small (time step). Let  $t_k = k\tau$  ( $k=0, 1, 2, \dots$ ). Approximate  $u(x, t_k)$  by  $u_k(x) = u_k^\tau(x)$  ( $x \in \Omega$ ,  $k=1, 2, \dots$ ).

$$\begin{cases} u_0 \text{ given,} \\ \frac{u_{k+1} - u_k}{\tau} = \Delta u_{k+1} \quad (k=0, 1, \dots). \end{cases}$$

Ideas:  $\partial_t u(x, t+\tau) \approx \frac{u(x, t+\tau) - u(x, t)}{\tau}$

$$\partial_t u(x, t+\tau) = \Delta u(x, t+\tau). \quad \text{r.h.s} = \text{r.h.s.}$$

Set  $u^\tau(x, t) = u_k^\tau(x)$  if  $t_k \leq t < t_{k+1}$ .

Convergence:

$$u^\tau \rightarrow u \text{ as } \tau \rightarrow 0.$$

Claim

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

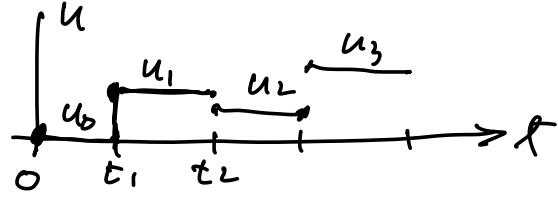
Proof  $I_k[u] := \frac{1}{2\tau} \|u - u_k\|_{L^2(\Omega)}^2 + I[u]$

$$= \int \left( \frac{1}{2\tau} (u - u_k)^2 + \frac{1}{2} (\nabla u)^2 \right) dx$$

$$\delta I_k[u][\varphi] := \frac{d}{dt} (I_k[u+t\varphi]) \Big|_{t=0}$$

$$= \int [\nabla u \cdot \nabla \varphi + \frac{1}{\tau} (u - u_k) \varphi] dx$$

$$= \int [-\Delta u + \frac{1}{\tau} (u - u_k)] \varphi dx.$$



So,  $\delta I_k[u] = -\Delta u + \frac{1}{\tau} (u - u_k)$ . Now,  $u_{k+1}$  solves the the Euler-Lagrange eq.  $\delta I_k[u] = 0$ . Since  $I_k$  is convex,  $u_{k+1}$  is the unique minimizer. QED

### The JKO scheme (Jordan-Kinderlehrer-Otto, 1998)

Define

$$F(\rho) = \int_{\mathbb{R}} \rho \log \rho dx, \quad \rho = \rho(x) \geq 0, \quad x \in \mathbb{R}.$$

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

Given:  $\rho_0 \in \mathcal{P}(\mathbb{R})$ ,  $\tau > 0$ ,  $t_k = k\tau$ ,  $k = 0, 1, 2, \dots$ .

$$\rho_{k+1} = \arg \min_{\rho} \left( \frac{1}{2\tau} W_2(\rho, \rho_k)^2 + F[\rho] \right).$$

Note: for  $\xi, \eta \in \mathcal{P}(\mathbb{R})$ ,  $W_2(\xi, \eta) = W_2(\xi dx, \eta dx)$ .  
 $x \in \mathbb{R}$ : compact, Euclidean distance.

Convergence:  $\rho^{\tau} \rightarrow \rho$  weakly 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:  $\rho dx$  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.

Lemma Let  $\rho_0 \in \mathcal{P}(n)$  be such that  $F[\rho_0] < \infty$ .

There exists a unique  $\hat{\rho} \in \mathcal{P}(n)$  s.t.

$$\hat{\rho} = \underset{\rho \in \mathcal{P}(n)}{\operatorname{argmin}} J[\rho]$$

where  $J[\rho] = W_2^2(\rho, \rho_0) + F[\rho]$ ,

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

Proof Let  $\alpha := \inf_{\rho \in \mathcal{P}(n)} J[\rho] < \infty$ . ( $F[\rho_0] < \infty$ )

Let  $\rho_n \in \mathcal{P}(n)$  ( $n=1, 2, \dots$ ) be such that  $J[\rho_n] \rightarrow \alpha$ .

Since  $\{F[\rho_n]\}$  is bounded and  $s \mapsto s \log s$  is superlinear, there exists a subseq. of  $\{\rho_n\}$ , not relabeled, such that  $\rho_n \rightharpoonup \hat{\rho}$  weakly in  $L'(n)$ .

(by de la Vallée Poussin's criterion of weakly sequential compactness of  $L'(n)$ .) Since all  $\rho_n \in \mathcal{P}(n)$ , we have  $\hat{\rho} \in \mathcal{P}(n)$ .

Since  $s \mapsto s \log s$  is convex, the functional  $F$  is sequentially weakly semi continuous. Thus,

$$\liminf_{n \rightarrow \infty} F[\rho_n] \geq F[\hat{\rho}].$$

Let  $\gamma_n \in \mathcal{A}(\rho_n, \hat{\rho})$  ( $= \mathcal{A}(\rho_n dx, \hat{\rho} dx)$ ) be such that  $W_2^2(\rho_n, \hat{\rho}) = \int_{\mathbb{R}^n \times \mathbb{R}^n} |x - y|^2 d\gamma_n(x, y)$ . ( $n=1, 2, \dots$ ).

Since  $\rho_n \rightharpoonup \hat{\rho}$  weakly in  $L'(n)$ ,  $\rho_n dx \rightarrow \hat{\rho} dx$

narrowly. Hence,  $\{\rho_n\}$  is tight by Prokhorov's Theorem. Hence  $\{\gamma_n\}$  is also tight, and again by Prokhorov's Theorem,  $\{\gamma_n\}$  is sequentially compact. Hence, it has a further subseq., not relabeled, such that  $\gamma_n \rightarrow \hat{\gamma} \in \mathcal{A}(\hat{P}, P_0)$  narrowly. Consequently,

$$W_2^L(P_n, P_0) = \int_{X \times X} |x - y|^2 d\gamma_n(x, y)$$

$$\rightarrow \int_{X \times X} |x - y|^2 d\hat{\gamma}(x, y) \geq W_2^L(\hat{P}, P_0).$$

Therefore,  $\alpha = \liminf_{n \rightarrow \infty} J[\rho_n] \geq J[\hat{P}] \geq \alpha$ , and  $\hat{P} \in \mathcal{P}(n)$  is a minimizer of  $J$  over  $\mathcal{P}(n)$ . The uniqueness from the convexity of  $J[P]$ , and  $W_2^L(\cdot; P_0)$ ; Cf. a result below. QED

Proposition Let  $(X, d)$  be a Polish space. Let  $\mu_0, \mu_1, \nu \in \mathcal{P}_2(X)$ . Then for each  $\lambda \in (0, 1)$ ,  $\mu_\lambda := (1-\lambda)\mu_0 + \lambda\mu_1 \in \mathcal{P}_2(X)$ . Moreover,

$$W_2^L(\mu_\lambda, \nu) \leq (1-\lambda) W_2^L(\mu_0, \nu) + \lambda W_2^L(\mu_1, \nu).$$

Proof Clearly  $\mu_\lambda \in \mathcal{P}_2(X)$ . For  $i=1, 2$ , let  $\gamma_i \in \mathcal{A}(\mu_i, \nu)$  be optimal:  $W_2^L(\mu_i, \nu) = \int_{X \times X} d^2(x, y) d\gamma_i(x, y)$ . Then,  $\gamma_\lambda = (1-\lambda)\gamma_0 + \lambda\gamma_1 \in \mathcal{A}(\mu_\lambda, \nu)$ .

Moreover,  
 $W_2^L(\mu_\lambda, \nu) \leq \int_{X \times X} d^2(x, y) d\gamma_\lambda(x, y) = (1-\lambda) W_2^L(\mu_0, \nu) + \lambda W_2^L(\mu_1, \nu)$ . 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  $P_0 \in \mathcal{P}(\Omega)$  and  $F[P_0] = \int_{\Omega} P_0 \log P_0 dx < \infty$ .

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

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

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

Remark The convergence can be improved to be the strong convergence of the entire original sequence  $\rho^\tau$ .