

Limit

911.5834 \cup $\frac{1}{\sqrt{2}}$ / ITSL / $\frac{1}{\sqrt{2}}$ \cup $\frac{1}{\sqrt{2}}$

(1)

$$\begin{aligned} \chi(P_n \| Q) &= E_Q\left(\frac{P_n}{Q}\right) - 1 = E_Q\left(\frac{1}{Q}\right) \int p_\theta \pi(d\theta) \int p_{\theta'} \pi(d\theta') - 1 \\ &= \int E_Q\left(\frac{p_\theta p_{\theta'}}{Q}\right) \pi(d\theta) \pi(d\theta') - 1 \\ &= E_{\theta, \theta'}(G(\theta, \theta')) - 1 \checkmark \end{aligned}$$

$$D_{KL}(P_n \| Q_n) < \chi(P_n \| Q) \Rightarrow D_{KL}(P_n \| Q_n) < O(1) \rightarrow \int P_n \log \frac{P_n}{Q_n} < O(1)$$

$$\forall q_n = 0 \Rightarrow P_n = 0 \Rightarrow P \Delta Q \checkmark$$

—————

$$\begin{aligned} P_\sigma(A_{ij}) &= \begin{cases} \sigma_i \sigma_j = 1 : P(A_{ij}) \\ \sigma_i \sigma_j = -1 : q(A_{ij}) \end{cases} \rightarrow P_\sigma(A) = \prod_{\sigma_i \sigma_j = 1} P(A_{ij}) \prod_{\sigma_i \sigma_j = -1} q(A_{ij}) \\ &= \prod \left(\frac{1 + \sigma_i \sigma_j}{2} P(A_{ij}) + \frac{1 - \sigma_i \sigma_j}{2} q(A_{ij}) \right) \\ &= \prod \left(\frac{1}{2} (P(A_{ij}) + q(A_{ij}) + \sigma_i \sigma_j (P(A_{ij}) - q(A_{ij}))) \right) \\ &= \prod \frac{1}{2} (P(A_{ij}) + q(A_{ij})) \end{aligned}$$

$$\begin{aligned} \frac{P_\sigma(A) P_{\tilde{\sigma}}(A)}{P_0} &= \prod_{i,j} \frac{1}{2} \times \frac{1}{2} \times (P(A_{ij}) + q(A_{ij}) - \sigma_i \sigma_j (P(A_{ij}) - q(A_{ij}))) \times (P(A_{ij}) + q(A_{ij}) - \tilde{\sigma}_i \tilde{\sigma}_j (P(A_{ij}) - q(A_{ij}))) \\ &\stackrel{\text{Simplify}}{\rightarrow} \prod_{i,j} \frac{1}{2} \beta \rightarrow \frac{P_\sigma(A) P_{\tilde{\sigma}}(A)}{P_0} = \prod_{i,j} \frac{1}{2} \beta \Rightarrow \int \frac{1}{2} \beta = \frac{1}{2} \int P(A) P(A) + \sigma_i \sigma_j \tilde{\sigma}_i \tilde{\sigma}_j \int P(A) - q(A) \\ &+ \sigma_i \sigma_j \tilde{\sigma}_i \tilde{\sigma}_j \int \frac{(P(A) - q(A))^2}{P(A) + q(A)} = \frac{1}{2} (1 + 0 + 2 \rho \sigma_i \sigma_j \tilde{\sigma}_i \tilde{\sigma}_j) \leq e^{\rho \sigma_i \sigma_j \tilde{\sigma}_i \tilde{\sigma}_j} \rightarrow P_\sigma < n e^{\rho \sigma_i \sigma_j \tilde{\sigma}_i \tilde{\sigma}_j} \end{aligned}$$

(b)

$$P = \frac{1}{r} \sum \frac{(p-q)^r}{(p+q)^r} = \frac{1}{r} \sum \frac{(p-q)^r}{(p+q)^r} + \frac{(1-p)(1-q)^r}{(1-p+1-q)^r} =$$

$$\frac{1}{r} \frac{(a_n - b_n)^r}{\left(\frac{a+b}{n}\right)^r} + \frac{1}{r} \frac{(b_n - a_n)^r}{r - \frac{a+b}{n}} = \frac{(a-b)^r}{r(a+b)} \cdot \frac{1}{n} + \frac{(b-a)^r}{(r-a-b)n}$$

$$= \frac{I}{n} + \frac{O(1)}{n} = \frac{I+O(1)}{n} \checkmark$$

(a)

$$a) \hat{\theta} = \arg \min \sum |y_i - \theta_i|^2 + \lambda \mathbb{1}(\theta_i \neq 0)$$

$$\Rightarrow \hat{\theta}_i = \arg \min |y_i - \theta_i|^2 + \lambda \mathbb{1}(\theta_i \neq 0)$$

$$\begin{pmatrix} \theta_1 \\ \vdots \\ \theta_p \end{pmatrix} \in \mathbb{R}^p \quad \begin{matrix} \theta_i = 0 \Rightarrow y_i^2 \\ \theta_i \neq 0 = y_i \Rightarrow \lambda \end{matrix} \Rightarrow y_i^2 = \lambda \Rightarrow |y_i| < \sqrt{\lambda}$$

$$\Rightarrow \theta_i = y_i \mathbb{1}(|y_i| > \sqrt{\lambda}) \Rightarrow \vec{\theta} = \vec{y} \mathbb{1}(|y| > \sqrt{\lambda}) \Rightarrow \tau \geq \sqrt{\lambda}$$

(b)

$$\hat{\theta} = \arg \min_{\theta \in \mathbb{R}^p} \sum_{i=1}^p |y_i - \theta_i|^2 + \lambda |\theta_i|$$

$$\Rightarrow \begin{cases} \theta_i < 0 \\ \theta_i > 0 \end{cases} \Rightarrow \partial_{\theta_i} := \begin{cases} -2(y_i - \theta_i) - \lambda = 0 \\ -2(y_i - \theta_i) + \lambda = 0 \end{cases} \Rightarrow \begin{cases} \theta_i = y_i + \lambda/2 \\ \theta_i = y_i - \lambda/2 \end{cases}$$

$$\Rightarrow \theta_i = 0 \Rightarrow -2(y_i - \theta_i) \in [-\lambda, \lambda] \Rightarrow \theta_i = 0$$

$$\rightarrow \hat{\theta}_i^{ST} = \begin{cases} y_i - \lambda/2 & y_i > \lambda/2 \\ 0 & |y_i| \leq \lambda/2 \\ y_i + \lambda/2 & y_i < -\lambda/2 \end{cases}$$

(c) چون $\|\theta\|_1$ استاندارد غیر منفرد است پس باید به این فکر کنیم که $\|y\|_F \leq \tau$ چه معنی دارد؟

if $\|y\|_F > \tau \rightarrow$ در صورتی که $\|y\|_F > \tau$ باشد، θ باید به صورت $[y; 0]$ باشد. جواب است.

(d) اگر $\|y\|_F \leq \tau$ باشد، کمترین مقدار ممکن $\|\theta\|_1$ با $\|y\|_F = 0$ است. $\hat{\theta}_{ST}$ است.

$$\|y\|_F \leq \tau \Rightarrow \vec{\theta} = 0$$

$$\|y\|_F > \tau \Rightarrow \theta_i = \min\{|y_i| - \tau, |y_i| + \tau\}$$

WU/ST:

(9)

$$\hat{G}(x) = \begin{cases} \hat{G}_1(x) = x_1 \\ \hat{G}_1(x) = 0 \end{cases} \rightarrow E_{\theta}(\|\hat{\theta} - \theta\|_r^r) = E_{\theta}(\underbrace{(x_1 - \theta_1)^r}_{N(0,1)} + u_{\theta_1}^r) \quad \text{w/} \\ = 1 + r^r = 2$$

$$\inf_{\hat{\theta}} \sup_{\theta \in \Theta} E_{\theta}(\|\hat{\theta} - \theta\|_r^r) \leq 1 \quad \checkmark$$

1/2

a)

(V)

Optimal: $\begin{cases} u_0 := \theta_0 = 0 \\ u_1 := \theta \sim \text{uni}(\{\tau \vec{e}_1, \dots, \tau \vec{e}_p\}) \rightarrow e_1, \dots, e_p \text{ are vectors in } \mathbb{R}^n \\ \text{and } \tau > 0, \text{ separation condition is } \Delta \geq \tau^r_{\ell} \end{cases}$

$$\chi^r(p_0 \| p_1) = E(e^{\theta^T \hat{\theta}} - 1) = \frac{\exp(\tau^r) - 1}{p}$$

$$\xrightarrow{u_1} \hat{\theta} \text{ iid of } \theta: \quad \tau = \sqrt{(1-\varepsilon) \log p} \xrightarrow{\varepsilon \in \mathcal{O}(1)} \chi^r(p_1 \| p_0) \rightarrow 0$$

$$\Rightarrow \text{TV}(\|p_1, p_0\|) \leq \sqrt{\frac{D_{KL}(p_1 \| p_0)}{r}} \leq \sqrt{\frac{\chi^r(p_1 \| p_0)}{r}} \Rightarrow p \rightarrow \infty \Rightarrow \text{TV}(p_1 - p_0) = 0$$

$$\rightarrow \inf_{\hat{\theta}} \sup_{\theta} E_{\theta}(\|\hat{\theta} - \theta\|_{\text{max}}^2) \leq \frac{(1-\varepsilon) \log p}{p} = \mathcal{O}(\log p)$$

b)

$$T = \max x_i \rightarrow E_{\theta}(\|T - \theta_{\max}\|_r^r) \leq E_{\theta}(\|x - \theta\|_r^r) = E(\|z\|_r^r) \leq (1 + \theta_p \|1\|) \log p \quad \checkmark$$

a)

$$L(f(x), Y) = 1 \{ f(x)Y < 0 \}$$

$$\phi(f(x), y) = \phi(f(x)y) \Rightarrow R_\phi(f) = E(\phi(f(x)y))$$

$$= \int P(Y=1, x=x) \phi(f(x)) + P(Y=-1, x=x) \phi(-f(x)) dx^n$$

$$= E_x(\underbrace{\eta(x) \phi(f(x)) + (1-\eta(x)) \phi(-f(x))}_{\ell_\phi(f(x), \eta(x))})$$

$$\ell_\phi(f(x), \eta(x)) \checkmark$$

$$b) B_{\phi, \pi} = R_\phi^* : R_\phi^* = B_{\phi, \pi}(p_+, p_-) = \inf_x \int \eta \phi(x(x)) p_+(x) + \phi(-x(x)) p_-(x) dx$$

$$p_{\pm 1} = P(X = \pm 1) \quad \bullet \quad = \int \inf_x (\phi(x) p_+(x) + \phi(-x) p_-(x) (1-\pi)) dx$$

$$B_{\phi, \pi} = B_{\phi, \pi}(p_+, p_-) = \int (\ell_\phi^*(\pi) - \ell_\phi^*(\eta(x)) p(x)) dx$$

$$= \int \sup_x \left(\ell_\phi^*(\pi) - \phi(x) \frac{p_+(x)}{p_-(x)} \pi - \phi(-x) \frac{p_-(x)}{p_+(x)} (1-\pi) \right) p(x) dx$$

$$= \int \sup_x \left(\ell_\phi^*(\pi) - \frac{\phi(x) \frac{p_+(x)}{p_-(x)} \pi + \phi(-x) (1-\pi)}{1-\pi + \pi \frac{p_+(x)}{p_-(x)}} \right) \left(\frac{p_+(x)}{p_-(x)} \pi + 1-\pi \right) p_-(x) dx$$

$$\rightarrow \frac{f}{\pi} = \sup_x \left(\ell_\phi^*(\pi) - \frac{\pi \phi(x) + (1-\pi) \phi(-x)}{\pi + (1-\pi)} \right) (\pi + 1-\pi)$$

$$\rightarrow R_{f, \phi}^* - R_\phi^* = D_\phi(p_+, p_-) \checkmark$$

$$\ell_{\phi}^*(\pi) = \inf_a \pi \phi(a) + (1-\pi) \phi(-a)$$

$$= \inf_a \pi \log(1 + e^a) + (1-\pi) \log(1 + e^{-a})$$

$$\Rightarrow \frac{-\pi e^a}{1+e^a} + \frac{(1-\pi) e^a}{1+e^a} = 0$$

$$\hookrightarrow e^a = \frac{\pi}{1-\pi} \rightarrow \ell_{\phi}^*(\pi) = \pi \log\left(1 + \frac{1-\pi}{\pi}\right) + (1-\pi) \log\left(1 + \frac{\pi}{1-\pi}\right)$$

$$= H(\pi)$$

$$\frac{-\pi a e^a}{1+e^a} + \frac{(1-\pi) e^a}{1+e^a} = 0 \rightarrow e^a = \frac{\pi a}{1-\pi}$$

$$\hookrightarrow f(a) = \frac{\ell_{\phi}^*(\pi) - \pi a \log(\pi a) - (1-\pi) \log(1-\pi)}{\pi a + 1 - \pi} - \log \frac{1}{\pi a + 1 - \pi}$$

$$x(\pi a + 1 - \pi)$$

$$H(\pi | x=x) = P(Y=1|x) \log \frac{1}{P(Y=1|x)} + P(Y=2|x) \log \frac{1}{P(Y=2|x)}$$

$$\frac{\pi p_1(x)}{\pi p_1(x) + p_2(x)(1-\pi)}$$

$$\Rightarrow D_f(p, q_{\pi-1}) = \int \ell_{\phi}^*(\pi) - H(\pi | x=x) / p(x) dx$$

$$= (H(\pi) - H(\pi | x=x)) p(x) dx$$

$$= I(x, \pi) \checkmark$$