Math 547: Mathematical Foundations of Statistical Learning Theory Fall 2017

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27.1 Talagrand's Contraction Inequality

First, from last time: $P(|Z - \mathbb{E}Z| \ge t) \le 2e^{-2t^2/\sum c_j^2}$ by a proof a bit more careful than ours. Let ϕ be a convex functions which is nondecreasing on $[0, \infty)$ and let φ be a contraction, meaning it's Lipschitz with constant 1. Then $\varphi(0) = 0$.

Theorem 1. Let $T \subset \mathbb{R}^n$. Then

$$\phi\left(\frac{1}{2}\mathbb{E}\sup_{t\in T}\left|\sum \varepsilon_{j}\varphi(t_{j})\right|\right) \leq \mathbb{E}\phi\left(\sup_{t\in T}\left|\sum \varepsilon_{j}t_{j}\right|\right).$$

Corollary 1. Let $\mathcal{F} = \{f : S \to \mathbb{R}\}$ such that $f \in \mathcal{F} \Rightarrow |f| \leq U$ a.s. Also let $X_1, \ldots, X_n \in S$, $T = \{(f(X_1), \ldots, f(X_n)) : f \in \mathcal{F}\}$. If X_1, \ldots, X_n are iid from P, then

$$\mathbb{E}_{\varepsilon}||R_n||_{\omega \circ \mathcal{F}} \le 2\mathbb{E}_{\varepsilon}||R_n||_{\mathcal{F}}.$$

For example, if $\tilde{\varphi}(x) = x^2$ on the interval [-U, U], $\varphi(x) = x^2/2U$ is a contraction. Hence

$$\mathbb{E}\sup_{f\in\mathcal{F}}\left|\frac{1}{n}\sum \varepsilon_j f^2(X_j)\right| \leq 4U\mathbb{E}\sup_{f\in\mathcal{F}}\left|\frac{1}{n}\sum \varepsilon_j f(X_j)\right|.$$

Proof of theorem:

First, we show that $\forall A: T \to \mathbb{R}$ measurable,

$$\mathbb{E}\phi\left(\sup_{t\in T}[A(t)+\sum_{j=1}^n\varepsilon_j\varphi(t_j)]\right)\leq \mathbb{E}\phi\left(\sup_{t\in T}[A(t)+\sum_{j=1}^n\varepsilon_jt_j]\right).$$

Turns out that n=1 is sufficient to prove the above. Indeed, if

$$\mathbb{E}\phi\left(\sup_{t\in T}[A(t)+\varepsilon\varphi(t)]\right) \leq \phi\left(\sup_{t\in T}[A(t)+\varepsilon t]\right),$$

then

$$\mathbb{E}\phi\left(\sup_{t\in T}[A(t) + \sum_{j=1}^{n-1}\varepsilon_{j}\varphi(t_{j}) + \varepsilon_{n}\varphi(t_{n})]\right) = \mathbb{E}\phi\left(\sup_{t\in T}[\widehat{A(t)} + \varepsilon_{n}\varphi(t_{n})]\right)$$

$$\leq \mathbb{E}_{\varepsilon,\varepsilon_{n-1}}\phi\left(\sup_{t\in T}[A(t) + \sum_{1}^{n-1}\varepsilon_{j}\varphi_{j}(t_{j}) + \varepsilon_{n}t_{n}]\right)$$

$$\cdots \leq \mathbb{E}\phi\left(\sup_{t\in T}[A(t) + \sum_{1}^{n}\varepsilon_{j}t_{j}]\right).$$

Want:

$$\mathbb{E}\phi\left(\sup_{t\in T}[A(t)+\varepsilon\varphi(t)]\right) \leq \mathbb{E}\phi\left(\sup_{t\in T}[A(t)+\varepsilon t]\right) \Leftrightarrow \phi\left(\sup_{t}[A(t)-\varphi(t)]\right) + \varphi\left(\sup_{t}[A(t)+\varphi(t)]\right) \\ \leq \phi\left(\sup_{t}[A(t)-t]\right) + \phi\left(\sup_{t}[A(t)+t]\right).$$

Let $A(t) = t_1$, $t = t_2$; we've reduced to $t \in \mathbb{R}$. Then want

$$\phi\left(\sup_{t_1,t_2\in\tilde{T}} t_1 - \varphi(t_2)\right) + \phi\left(\sup_{t_1,t_2} (t_1 + \varphi(t_2))\right) \le \phi\left(\sup_{t_1,t_2} (t_1 - t_2)\right) + \phi\left(\sup_{t_1,t_2} (t_1 + t_2)\right).$$

We can assume WLOG that supremum is attained (if not, just do argument for supremum minus ϵ for small $\epsilon > 0$. Let $\overline{t_1}, \overline{t_2}; \overline{s_1}, \overline{s_2}$ be the argmaxes of the above LHS expressions.

Case (a): $\overline{t_2} \geq 0$, $\overline{s_2} \geq 0$, $\overline{t_2} \geq \overline{s_2}$. This case is similar to the same with $\overline{t_2} \leq \overline{s_2}$ after renaming the \overline{t} and \overline{s} and replacing φ with $-\varphi$. We will show

$$\phi(\overline{t_1} + \varphi(\overline{t_2})) + \phi(\overline{s_1} - \varphi(\overline{s_2})) \le \phi(\overline{t_1} + \overline{t_2}) + \phi(\overline{s_1} - \overline{s_2}).$$

Now some cumbersome arithmetic...define

$$a = \overline{t_1} + \varphi(\overline{t_2})$$
 $b = \overline{t_1} + \overline{t_2}$ $c = \overline{s_1} - \overline{s_2}$ $d = \overline{s_1} - \varphi(\overline{s_2})$.

Then $a \leq b$ since φ is a contraction, $c \leq d$. Also $b - a = \overline{t_2} - \varphi(\overline{t_2}) \geq \overline{s_2} - \varphi(\overline{s_2}) = d - c$, so that

$$\varphi(\overline{t_2}) - \varphi(\overline{s_2}) \le \overline{t_2} - \overline{s_2}.$$

Then $c \le a \le d \le b$; it remains to show $c \le a$.

$$a = \overline{t_1} + \varphi(\overline{t_2}) \ge \overline{s_1} + \varphi(\overline{s_2}) \ge \overline{s_1} - \overline{s_2} = c,$$

the first due to nature of $\overline{t_1}$ as argmax, and the second because φ is a contraction. Then by convexity

$$\phi(b) - \phi(a) \ge \phi(d) - \phi(c),$$

which evidently completes our proof.

Case (b): $\overline{s_2} \ge \overline{t_2}$ reduces to previous case.

Case (c): $\overline{t_2} \leq 0$, $\overline{s_2} \leq 0$; then rename $(\overline{t_1}, \overline{t_2}) \mapsto (\overline{t_1}, -\overline{t_2})$, same for s, $\varphi(x) = \varphi(-x)$. Reduces to previous case.

Case (d): $\overline{t_2} \ge 0$, $\overline{s_2} \le 0$. Then $\varphi(\overline{t_2}) \le \overline{t_2}$, $\varphi(\overline{s_2}) \le -\overline{s_2}$. Since ϕ is monotone,

$$\phi(\overline{t_1} + \varphi(\overline{t_2})) + \phi(\overline{s_1} - \varphi(\overline{s_2})) \le \phi(\overline{t_1} + \overline{t_2}) + \phi(\overline{s_1} - \overline{s_2}).$$