Theorem 1

Suppose we have sub-gauss errors.

Suppose

$$\int_{0}^{R} H^{1/2}(u, \mathcal{G}(T), \|\cdot\|_{V}) du \le \psi_{T}(R)$$

Suppose

$$\frac{\psi_T\left(a+u\right)}{u^2}$$

are nonincreasing wrt to u for all u > 0, a > 0.

Then for all $\delta > 0$ s.t.

$$\sqrt{n_V}\delta^2 \ge c \left[\psi_T \left(2 \left\| \hat{g}_{\tilde{\lambda}} - g^* \right\|_V + 2\delta \right) \vee \left(2 \left\| \hat{g}_{\tilde{\lambda}} - g^* \right\|_V + 2\delta \right) \right]$$

With high probability

$$\|\hat{g}_{\hat{\lambda}} - g^*\|_V \le \|\hat{g}_{\tilde{\lambda}} - g^*\|_V + \delta$$

Proof for Theorem 1

The basic inequality gives us

$$\left\|\hat{g}_{\hat{\lambda}} - g^*\right\|_V^2 \le \left\|\hat{g}_{\tilde{\lambda}}(\cdot|T) - g^*\right\|_V^2 + 2\langle \epsilon, \hat{g}_{\hat{\lambda}} - \hat{g}_{\tilde{\lambda}}\rangle_V$$

Note that if $\|\hat{g}_{\tilde{\lambda}} - g^*\|_{V} \leq \|\hat{g}_{\hat{\lambda}} - g^*\|_{V}$, then

$$\left(\left\| \hat{g}_{\hat{\lambda}} - g^* \right\|_{V} - \left\| \hat{g}_{\tilde{\lambda}} - g^* \right\|_{V} \right)^2 \leq \left\| \hat{g}_{\hat{\lambda}} - g^* \right\|_{V}^2 - \left\| \hat{g}_{\tilde{\lambda}} - g^* \right\|_{V}^2$$

By a peeling argument, we have

$$Pr\left(\left\|\hat{g}_{\hat{\lambda}} - g^{*}\right\|_{V} - \left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V} \ge \delta\right)$$

$$= \sum_{s=0}^{\infty} Pr\left(2^{s}\delta \le \left\|\hat{g}_{\hat{\lambda}} - g^{*}\right\|_{V} - \left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V} \le 2^{s+1}\delta\right)$$

$$\le \sum_{s=0}^{\infty} Pr\left(\left\|\hat{g}_{\hat{\lambda}} - g^{*}\right\|_{V} - \left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V} \ge 2^{s}\delta \wedge \left\|\hat{g}_{\hat{\lambda}} - \hat{g}_{\tilde{\lambda}}\right\|_{V} \le 2\left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V} + 2^{s+1}\delta\right)$$

$$= \sum_{s=0}^{\infty} Pr\left(\left(\left\|\hat{g}_{\hat{\lambda}} - g^{*}\right\|_{V} - \left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V}\right)^{2} \ge 2^{2s}\delta^{2} \wedge \left\|\hat{g}_{\hat{\lambda}} - \hat{g}_{\tilde{\lambda}}\right\|_{V} \le 2\left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V} + 2^{s+1}\delta\right)$$

$$\le \sum_{s=0}^{\infty} Pr\left(\left\|\hat{g}_{\hat{\lambda}} - g^{*}\right\|_{V}^{2} - \left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V}^{2} \ge 2^{2s}\delta^{2} \wedge \left\|\hat{g}_{\hat{\lambda}} - \hat{g}_{\tilde{\lambda}}\right\|_{V} \le 2\left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V} + 2^{s+1}\delta\right)$$

$$\le \sum_{s=0}^{\infty} Pr\left(\sup_{\left\|\hat{g}_{\lambda} - \hat{g}_{\tilde{\lambda}}\right\|_{V} \le 2\left\|\hat{g}_{\tilde{\lambda}} - g^{*}\right\|_{V} + 2^{s+1}\delta\right)$$

To apply the lemma based on vandegeer corollary 8.3 (see below), we must check that δ satisfies all the conditions. By our choice of δ and the fact that $\psi(a+u)/u$ is nonincreasing, we have that for for all s

$$\sqrt{n_V} 2^{2s-1} \delta^2 \ge c \left[\psi_T \left(2 \| \hat{g}_{\tilde{\lambda}} - g^* \|_V + 2^{s+1} \delta \right) \vee \left(2 \| \hat{g}_{\tilde{\lambda}} - g^* \|_V + 2^{s+1} \delta \right) \right]$$

So by the lemma below, we have

$$Pr\left(\left\|\hat{g}_{\hat{\lambda}} - g^*\right\|_{V} - \|\hat{g}_{\tilde{\lambda}} - g^*\|_{V} \ge \delta \wedge \|\epsilon\|_{V} \le 2\sigma \wedge \|\epsilon\|_{T} \le 2\sigma\right)$$

$$\le C \sum_{s=0}^{\infty} \exp\left(-n_{V} \frac{2^{4s-2}\delta^{4}}{4C^{2} \left(2\left\|\hat{g}_{\tilde{\lambda}} - g^*\right\|_{V} + 2^{s+1}\delta\right)^{2}}\right)$$

$$\le C \sum_{s=0}^{\infty} \exp\left(-n_{V} \frac{2^{4s-2}\delta^{4}}{64C^{2} \left\|\hat{g}_{\tilde{\lambda}} - g^*\right\|_{V}^{2}}\right) \vee \exp\left(-n_{V} \frac{2^{2s}\delta^{2}}{196C^{2}}\right)$$

$$\le c \exp\left(-\frac{n_{V}\delta^{4}}{c^{2} \left\|\hat{g}_{\tilde{\lambda}} - g^*\right\|_{V}^{2}}\right) + c \exp\left(-\frac{n_{V}\delta^{2}}{c^{2}}\right)$$

for some constant c.

By Bernstein's inequality, we also know

$$Pr(\|\epsilon\|_T \ge 2\sigma) \le \exp\left(-n_T \frac{\sigma^2}{K}\right)$$

and similarly for $Pr(\|\epsilon\|_V \geq 2\sigma)$.

Hence we have found for the given δ choice, we have

$$Pr\left(\left\|\hat{g}_{\hat{\lambda}} - g^*\right\|_{V} - \left\|\hat{g}_{\hat{\lambda}} - g^*\right\|_{V} \ge \delta\right)$$

$$\leq c \exp\left(-\frac{n_{V}\delta^{4}}{c^{2}\left\|\hat{g}_{\hat{\lambda}} - g^*\right\|_{V}^{2}}\right) + c \exp\left(-\frac{n_{V}\delta^{2}}{c^{2}}\right) + \exp\left(-n_{T}\frac{\sigma^{2}}{K}\right) + \exp\left(-n_{V}\frac{\sigma^{2}}{K}\right)$$

Corollary to Theorem 1

Let $\Lambda = [n^{-t_{min}}, n^{t_{max}}]^J$.

Suppose that if $\|\epsilon\|_T \leq 2\sigma$, there are constants C, κ such that for any u > 0, we have for all $\lambda \in \Lambda$

$$\|\boldsymbol{\lambda}_1 - \boldsymbol{\lambda}_2\| < Cn^{\kappa}u \implies \|\hat{q}_{\boldsymbol{\lambda}_1} - \hat{q}_{\boldsymbol{\lambda}_2}\|_V < u$$

Then there are constants c, c_1, c_2 s.t. with high probability,

$$\|\hat{g}_{\hat{\lambda}} - g^*\|_{V} \le \|\hat{g}_{\tilde{\lambda}} - g^*\|_{V} + \frac{c_1 \left(J(\log n_V + c_2)\right)^{1/2}}{\sqrt{n_V}} + \sqrt{c \left(J(\log n_V + c_2)\right)^{1/2} \|\hat{g}_{\tilde{\lambda}} - g^*\|_{V} n_V^{-1/2}}$$

Proof

By Lemma param covering cube, we have

$$H(u, \mathcal{G}(T), \|\cdot\|_V) \le \log \frac{1}{C_J} + J \log \left(\frac{2n^{t_{max} - \kappa} + 2Cu}{Cu}\right)$$

Let
$$R_1 = R \wedge \sqrt{n^{t_{max} - \kappa}/C}$$
.

Then

$$\begin{split} \int_{0}^{R} H^{1/2}(u,\mathcal{G}(T),\|\cdot\|_{V}) du & \leq \int_{0}^{R} \left[\log \frac{1}{C_{J}} + J \log \left(\frac{2n^{t_{max}-\kappa} + 2Cu}{Cu} \right) \right]^{1/2} du \\ & < \int_{0}^{R} \left[\log \frac{1}{C_{J}} + J \log 4 + J \log \left(\frac{4n^{t_{max}-\kappa}}{Cu} \right) \right]^{1/2} du \\ & = R \int_{0}^{1} \left[\log \frac{1}{C_{J}} + J \log 4 + J \log \left(\frac{4n^{t_{max}-\kappa}}{CRv} \right) \right]^{1/2} dv \\ & \leq R \left[\int_{0}^{1} \log \frac{1}{C_{J}} + J \log 4 + J \log \left(\frac{4n^{t_{max}-\kappa}}{CRv} \right) dv \right]^{1/2} \\ & = R \left[\log \frac{1}{C_{J}} + J(1 + \log 4) + J \log \left(\frac{4n^{t_{max}-\kappa}}{C} \right) + J \log \frac{1}{R} \right]^{1/2} \\ & \leq R \left(\left[\log \frac{1}{C_{J}} + J(1 + \log 4) + J \log \left(\frac{4n^{t_{max}-\kappa}}{C} \right) \right]^{1/2} + \sqrt{J \log \frac{1}{R}} \vee 0 \right) \end{split}$$

The second bound is crazy loose (but is okay I think). The third inequality follows from concavity of the square root.

The term $\log \frac{1}{R}$ is nasty. When choosing δ , we will replace it with $\log n_V$ since for all $R \geq \frac{1}{n_V}$, we have

$$\log \frac{1}{R} \vee 0 \le \log n_V$$

Hence it suffices to choose δ s.t.

$$\sqrt{n_V}\delta^2 \ge c \left(\left\| \hat{g}_{\tilde{\lambda}} - g^* \right\|_V + \delta \right) \left(\left[\log \frac{1}{C_J} + J(1 + \log 4) + J \log \left(\frac{4n^{t_{max} - \kappa}}{C} \right) \right]^{1/2} + \sqrt{J \log n_V} \right)$$

Let

$$K = c \left(\left[\log \frac{1}{C_J} + J(1 + \log 4) + J \log \left(\frac{4n^{t_{max} - \kappa}}{C} \right) \right]^{1/2} + \sqrt{J \log n_V} \right)$$

and

$$\omega = \left\| \hat{g}_{\tilde{\lambda}} - g^* \right\|_V$$

So we are interested in δ s.t.

$$\sqrt{n_V}\delta^2 - K\delta - K\omega \ge 0$$

The quadratic formula tells us that this is satisfied when

$$\delta \geq \frac{K + \sqrt{K^2 + 4K\omega\sqrt{n_V}}}{2\sqrt{n_V}}$$

A more simple lower bound for δ is

$$\delta \ge \frac{K}{\sqrt{n_V}} + \sqrt{K \left\| \hat{g}_{\tilde{\lambda}} - g^* \right\|_V n_V^{-1/2}}$$

Lemma (Based on Vandegeer Corollary 8.3)

Let Q_m be the empirical distribution of m observations at covariates x_i .

Suppose ϵ are m independent sub-gaussian errors. Suppose the model class $\mathcal{F}(T)$ has elements $\sup_{f \in \mathcal{F}_n(T)} \|f\|_{Q_m} \leq R$ and satisfies

$$\psi_T(R) \ge \int_0^R H^{1/2}(u, \mathcal{F}(T), \|\cdot\|_{Q_m}) du$$

There is C dependent only on the sub-gaussian constants such that for all $\delta > 0$ such that

$$\sqrt{m}\delta \ge C(\psi_T(R) \lor R)$$

we have

$$Pr\left(\sup_{f\in\mathcal{F}_n(T)}\left|\frac{1}{m}\sum_{i=1}^m\epsilon_if(x_i)\right|\geq\delta\wedge\|\epsilon\|_{Q_m}\leq2\sigma\right)\leq C\exp\left(-\frac{m\delta^2}{4C^2R^2}\right)$$

Lemma param covering cube

Suppose we have $\Lambda = [\lambda_{min}, \lambda_{max}]^J$, we have

$$N(\delta, \Lambda, \|\cdot\|_2) \le \frac{1}{C_J} \left(\frac{2(\lambda_{max} - \lambda_{min}) + 2\delta}{\delta} \right)^J$$

Proof

(Based on Lemma 2.5 in vandegeer)

Let $C = \{c_j\}_{j=1}^N \subset \Lambda$ be the largest set s.t. two distinct points c_{j_1}, c_{j_2} are at least δ apart. Then balls with radius δ centered at C cover Λ .

If the balls have radius $\delta/4$ and are centered at C, then the balls must be disjoint and are completely contained in the box $[\lambda_{min} - \delta/4, \lambda_{max} + \delta/4]^J$.

The volume of a ball with radius ρ is $C_J \rho^J$. Therefore

$$NC_J(\delta/4)^J \le (\lambda_{max} - \lambda_{min} + \delta/2)^J$$