

0.1 Another cute lemma?

Lemma 1

Suppose $I : \mathcal{G} \mapsto [0, \infty)$ is a pseudonorm. So $I(g) + I(f) \geq I(g + f)$ and $I(cg) = cI(g) \forall c > 0$. Then

$$\left| \frac{\partial}{\partial t} I(g + th) \right| \leq I(h)$$

Proof:

By the triangle inequality for I , we have

$$|I(g + th) - I(g)| \leq I(th)$$

then diving by t and taking the limit, we have

$$\frac{\partial}{\partial t} I(g + th) = \lim_{t \rightarrow 0} \frac{|I(g + th) - I(g)|}{t} \leq \lim_{t \rightarrow 0} \frac{I(th)}{t} = I(h)$$

Lemma 2

Suppose the conditions in lemma 1, as well as

$$I^v(g) \leq M \|g\|_n^2 + M_0$$

For any $\lambda < \tilde{\lambda}$,

$$\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}\|_n^2 = O_p(1) v \tilde{\lambda}^2 I^v(\hat{g}_{\lambda})$$

Proof:

Since \hat{g}_{λ} is the minimizer of the penalized criterion, then for h s.t. $\hat{g}_{\lambda} + th \in \mathcal{G} \forall t > 0$

$$\frac{\partial}{\partial t} (\|y - (\hat{g}_{\lambda} + th)\|_n^2 + \lambda^2 I^v(\hat{g}_{\lambda} + th)) \big|_{t=0} = 0$$

But we also know that

$$\frac{\partial}{\partial t} (\|y - (\hat{g}_{\lambda} + th)\|_n^2 + \lambda^2 I(\hat{g}_{\lambda} + th)) \big|_{t=0} = -2(y - \hat{g}_{\lambda}, h)_T + \lambda^2 v I^{v-1}(\hat{g}_{\lambda}) \frac{\partial}{\partial t} I(\hat{g}_{\lambda} + th)$$

Then for $\lambda < \tilde{\lambda}$, we have

$$-2(y - \hat{g}_{\lambda}, h) + \lambda^2 v I^{v-1}(\hat{g}_{\lambda}) \frac{\partial}{\partial t} I(\hat{g}_{\lambda} + th) = -2(y - \hat{g}_{\tilde{\lambda}}, h)_T + \tilde{\lambda}^2 v I^{v-1}(\hat{g}_{\tilde{\lambda}}) \frac{\partial}{\partial t} I(\hat{g}_{\tilde{\lambda}} + th)$$

Rearranging, we get

$$\begin{aligned} 0 &= 2(\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}, h)_T + v \left(\tilde{\lambda}^2 I^{v-1}(\hat{g}_{\tilde{\lambda}}) \frac{\partial}{\partial t} I(\hat{g}_{\tilde{\lambda}} + th) - \lambda^2 I^{v-1}(\hat{g}_{\lambda}) \frac{\partial}{\partial t} I(\hat{g}_{\lambda} + th) \right) \\ &\leq 2(\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}, h)_T + v I^{v-1}(\hat{g}_{\lambda}) I(h) \tilde{\lambda}^2 \end{aligned}$$

where the first inequality follows from Lemma 1. Setting $h = \hat{g}_{\lambda} - \hat{g}_{\tilde{\lambda}}$, we get

$$\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}\|_n^2 = v \tilde{\lambda}^2 I^{v-1}(\hat{g}_{\lambda}) I(\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda})$$

Since

$$I(\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}) \leq I(\hat{g}_{\tilde{\lambda}}) + I(\hat{g}_{\lambda})$$

then

$$\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}\|_n^2 = O_p(1) v \tilde{\lambda}^2 I^v(\hat{g}_{\lambda})$$

(Note that if we assume $I^v(\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}) \leq O_p(1) \|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}\|_n^2 + C$, then $\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\lambda}\|_n^{2-2/v} = O_p(1) v I^{v-1}(\hat{g}_{\lambda}) \tilde{\lambda}^2 + C$.)

Lemma 3

Suppose

$$H(\delta, \{g : I(g) \leq 1\}, \|\cdot\|_n) \leq A\delta^{-\alpha}$$

and

$$\frac{\|g\|_n}{I(g)} \leq K < \infty$$

Suppose for some $\hat{\lambda} \leq \tilde{\lambda}$, $\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}}\|_n$ is bounded (doesn't grow with n).
Suppose

$$I^v(\hat{g}_{\hat{\lambda}}) \leq M\|\hat{g}_{\hat{\lambda}}\|_n^2 + M_0$$

(Note: this is probably not necessary, but I'm sleepy)

Then

$$H\left(\delta, \left\{\hat{g}_{\lambda} : \lambda \geq \hat{\lambda}\right\}, \|\cdot\|_n\right) \leq A\delta^{-\alpha}$$

Proof:

Let $I(\hat{g}_{\hat{\lambda}}) = R$. We'll suppose $R \geq 1$. Otherwise, we'll be done.

By the assumptions, R is bounded since

$$I^v(\hat{g}_{\hat{\lambda}}) \leq M\|\hat{g}_{\hat{\lambda}}\|_n^2 + M_0 \leq M(\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}}\|_n + \|\hat{g}_{\tilde{\lambda}}\|_n)^2 + M_0$$

Note that $\{\hat{g}_{\lambda} : \lambda \geq \hat{\lambda}\} \subseteq \{g : I(g) \leq R\} = \{Rg : I(g) \leq 1\}$. Note that if h is the closest function in the δ -cover for $\{g : I(g) \leq 1\}$, then

$$\|g - h\|_n \leq \delta \implies \|Rg - \delta \lfloor \frac{R}{\delta} \rfloor h\|_n \leq \|Rg - Rh\|_n + \delta \|h\|_n \leq \delta(1 + R + K)$$

Then for some constant \tilde{A} dependent on A, R, K ,

$$H\left(\delta, \left\{\hat{g}_{\lambda} : \lambda \geq \hat{\lambda}\right\}, \|\cdot\|_n\right) \leq H(\delta, \{Rg : I(g) \leq 1\}, \|\cdot\|_n) \leq \tilde{A}\delta^{-\alpha}$$

Lemma 4

Suppose

$$H(\delta, \{\hat{g}_{\lambda} : \lambda \in \Lambda\}, \|\cdot\|_n) \leq A\delta^{-\alpha}$$

Suppose $\hat{\lambda}$ is the CV-fitted lambda and $\tilde{\lambda}$ is the oracle lambda given in Vandegeer.

Suppose

$$\|g^* - \hat{g}_{\tilde{\lambda}}\|_V = O_p(1) \|g^* - \hat{g}_{\tilde{\lambda}}\|_T$$

Then

$$\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}}\|_V = O_p(n^{-1/(2+\alpha)})$$

Proof:

The basic inequality gives us

$$\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}}\|_V^2 \leq 2|(\epsilon, \hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}})_V| + 2|(g^* - \hat{g}_{\tilde{\lambda}}, \hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}})_V|$$

If $|(\epsilon, \hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}})_T|$ is the bigger term, then

$$\|\hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}}\|_V^2 \leq O_P(1) |(\epsilon, \hat{g}_{\tilde{\lambda}} - \hat{g}_{\hat{\lambda}})_V|$$

Use the same arguments as Thrm 9.1 to show that $\|\hat{g}_{\bar{\lambda}} - \hat{g}_{\hat{\lambda}}\|_V = O_p(n^{-1/2})$.
 If $|(g^* - \hat{g}_{\bar{\lambda}}, \hat{g}_{\bar{\lambda}} - \hat{g}_{\hat{\lambda}})_T|$ is the bigger term, then by Cauchy Schwarz

$$\begin{aligned} \|\hat{g}_{\bar{\lambda}} - \hat{g}_{\hat{\lambda}}\|_V^2 &\leq O_p(1) |(g^* - \hat{g}_{\bar{\lambda}}, \hat{g}_{\bar{\lambda}} - \hat{g}_{\hat{\lambda}})_V| \\ &\leq O_p(1) \|g^* - \hat{g}_{\bar{\lambda}}\|_V \|\hat{g}_{\bar{\lambda}} - \hat{g}_{\hat{\lambda}}\|_V \end{aligned}$$

Hence

$$\|\hat{g}_{\bar{\lambda}} - \hat{g}_{\hat{\lambda}}\|_V \leq O_p(n^{-1/(2+\alpha)})$$

Need to show:

If we show that $I(\hat{g}_{\hat{\lambda}}) = O_p(n^{\frac{2}{(2+\alpha)v}}) \leq O_p(n^{1/\alpha})$, then by Lemma 2, 3, 4, we have that $\|\hat{g}_{\bar{\lambda}} - \hat{g}_{\hat{\lambda}}\|_V \leq O_p(n^{-1/(2+\alpha)})$.

It might be easier to show that for many penalties, $I(\hat{g}_{\lambda=0}) = O_p(n^{\frac{2}{(2+\alpha)v}})???$

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Lemma 3:

Suppose the function class \mathcal{F} is bounded s.t. $\sup_{f \in \mathcal{F}} \|f\|_n \leq R < \infty$. Let

$$\tilde{\mathcal{F}} = \{\gamma f : f \in \mathcal{F}, \gamma \in (0, 1]\}$$

$$H\left(\delta(1 + R + \delta), \tilde{\mathcal{F}}, \|\cdot\|_n\right) \leq \log(1 + \lfloor \frac{1}{\delta} \rfloor) + H(\delta, \mathcal{F}, \|\cdot\|_n)$$

Proof: Let $\{h_i\}_{i=1}^N$ be the δ -cover for \mathcal{F} . Consider any $f \in \mathcal{F}$ and let $h_{(f)}$ be the closest function in δ -cover for \mathcal{F} . Choose $j \in \mathbb{Z}^+$ such that $|\gamma - \delta j| < \delta$.

$$\begin{aligned} \|\gamma f - \delta j h_{(f)}\|_n &\leq \|\gamma f - \gamma h_{(f)}\|_n + \|\gamma h_{(f)} - \delta j h_{(f)}\|_n \\ &\leq \gamma \|f - h_{(f)}\|_n + |\gamma - \delta j| \|h_{(f)}\|_n \\ &\leq \gamma \delta + \delta (\|f - h_{(f)}\|_n + \|f\|_n) \\ &\leq \gamma \delta + \delta (\delta + R) \\ &\leq \delta (1 + R + \delta) \end{aligned}$$

Hence we have found that the following $N(1 + \lfloor \frac{1}{\delta} \rfloor)$ functions form a $\delta(1 + R + \delta)$ -cover for $\tilde{\mathcal{F}}$:

$$\{h_i\}_{i=1}^N \cup \left\{ j \delta h_i : j \in 1 : \lfloor \frac{1}{\delta} \rfloor, i \in 1 : N \right\}$$

Lemma 4:

Define function classes $\{\mathcal{F}_j\}_{j=1}^J$ and

$$\tilde{\mathcal{F}} = \left\{ \sum_{j=1}^J f_j : f_j \in \mathcal{F}_j \right\}$$

Then

$$H\left(J\delta, \tilde{\mathcal{F}}, \|\cdot\|_n\right) \leq \sum_{j=1}^J H(\delta, \mathcal{F}_j, \|\cdot\|_n)$$

Proof: For every $j = 1 : J$, consider any $f_j \in \mathcal{F}_j$ and let $h_{(j)}$ be the closest function in the δ -cover for \mathcal{F}_j .

$$\left\| \sum_{j=1}^J f_j - \sum_{j=1}^J h_{(j)} \right\| \leq \sum_{j=1}^J \|f_j - h_{(j)}\| \leq J\delta$$

Hence $\exp\left(\sum_{j=1}^J H(\delta, \mathcal{F}_j, \|\cdot\|_n)\right)$ functions form a $J\delta$ -cover for $\tilde{\mathcal{F}}$.

Lemma 5:

Suppose for all $j = 1, \dots, J$, there is some $\alpha_j > 0$ and $A_j > 0$ s.t. the following entropy bound holds for all $\delta > 0$

$$H\left(\delta, \left\{ \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\|_T\right) \leq A\delta^{-\alpha_j}$$

Then for sufficiently small $\delta > 0$, we have

$$H\left(\delta, \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sup_{j \in 1:J} (I(g_j) + I(g_j^*))} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\|_T\right) \leq 2JA \left(\frac{\delta}{2J(1+R)} \right)^{-\alpha_{max}}$$

where $\alpha_{max} = \max_{j \in 1:J} \alpha_j$.

Proof: By Lemma 3,

$$H\left(\delta(1+R+\delta), \left\{ \gamma \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0, \gamma \in (0, 1] \right\}, \|\cdot\|_T\right) \leq \log(1 + \lfloor \frac{1}{\delta} \rfloor) + A\delta^{-\alpha_j}$$

Note that

$$\frac{\sum_{j=1}^J g_j - g_j^*}{\sup_{j \in 1:J} (I(g_j) + I(g_j^*))} = \sum_{j=1}^J \left(\frac{I(g_j) + I(g_j^*)}{\sup_{\ell \in 1:J} (I(g_\ell) + I(g_\ell^*))} \right) \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)}$$

By Lemma 4,

$$H\left(J\delta(1+R+\delta), \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sup_{j \in 1:J} (I(g_j) + I(g_j^*))} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\|_T\right) \leq J \log(1 + \lfloor \frac{1}{\delta} \rfloor) + JA\delta^{-\alpha_j}$$

Hence for sufficiently small δ ,

$$H\left(J\delta(1+R+\delta), \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sup_{j \in 1:J} (I(g_j) + I(g_j^*))} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\|_T\right) \leq 2JA\delta^{-\alpha_{max}}$$

Rearranging, we get

$$\begin{aligned} H\left(\delta, \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sup_{j \in 1:J} (I(g_j) + I(g_j^*))} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\|_T\right) &\leq 2AJ \left(\sqrt{\left(\frac{1+R}{2}\right)^2 + \frac{\delta}{J}} - \frac{1+R}{2} \right)^{-\alpha_{max}} \\ &\leq 2AJ \left(\frac{\delta}{2J(1+R)} \right)^{-\alpha_{max}} \end{aligned}$$

(Used the fact that for $b > 0$ small enough, $\sqrt{a^2 + b} - a \geq \sqrt{(a + \frac{b}{4a})^2} - a = \frac{b}{4a}$)

Lemma 5b:

Suppose for all $j = 1, \dots, J$, there is some $\alpha_j > 0$ and $A_j > 0$ s.t. the following entropy bound holds for all $\delta > 0$

$$H \left(\delta, \left\{ \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\| \right) \leq A\delta^{-\alpha_j}$$

Then for sufficiently small $\delta > 0$, we have

$$H \left(\delta, \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sum_{j=1}^J I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\| \right) \leq 2JA(STUFF)^{-\alpha_{max}}$$

where $\alpha_{max} = \max_{j \in 1:J} \alpha_j$.

Proof: Note that

$$\frac{\sum_{j=1}^J g_j - g_j^*}{\sum_{j=1}^J I(g_j) + I(g_j^*)} = \sum_{j=1}^J \left(\frac{I(g_j) + I(g_j^*)}{\sum_{\ell=1}^J I(g_\ell) + I(g_\ell^*)} \right) \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)}$$

So we can express

$$\left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sum_{j=1}^J I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\} \subseteq \left\{ \sum_{j=1}^J \gamma_j \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0, \sum_{j=1}^J \gamma_j = 1 \right\}$$

Let \mathcal{H}_j be the set of functions that form a δ -cover for $\left\{ \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}$. Consider the set of functions

$$\left\{ \sum_{j=1}^J \delta k_j h_j : h_j \in \mathcal{H}_j, 1 - \frac{1}{\delta} \leq \delta \sum_{j=1}^J k_j \leq 1, k_j \in 1 : \lfloor \frac{1}{\delta} \rfloor \right\}$$

Let $|\delta k_j - \gamma_j| < \delta/2$. Then

$$\begin{aligned} \left\| \sum_{j=1}^J \gamma_j \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} - \sum_{j=1}^J \delta k_j h_j \right\| &\leq \sum_{j=1}^J \left\| \gamma_j \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} - \delta k_j h_j \right\| \\ &\leq \sum_{j=1}^J \left\| \gamma_j \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} - \gamma_j h_j \right\| + |\delta k_j - \gamma_j| \|h_j\| \\ &\leq \sum_{j=1}^J \left(\gamma_j \delta + \frac{\delta}{2} \left(\left\| \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} - h_j \right\| + \left\| \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} \right\| \right) \right) \\ &\leq \delta(1 + JR + J\delta) \end{aligned}$$

Hence these $(\Pi_{j=1}^J N_j) \left(\lfloor \frac{1}{\delta} \rfloor + J - 1 \right)$ functions form a $\delta(1 + JR + J\delta)$ cover. Hence the entropy is

$$H \left(\delta(1 + JR + J\delta), \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sum_{j=1}^J I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\| \right) \leq (J-1) \log(1 + J + \lfloor \frac{1}{\delta} \rfloor) + A \sum_{j=1}^J \delta^{-\alpha_j}$$

Note:

$$\binom{\lfloor \frac{1}{\delta} \rfloor + J - 1}{J - 1} \leq \left(\lfloor \frac{1}{\delta} \rfloor + J - 1 \right)^{J-1}$$

Hence for sufficiently small δ ,

$$H \left(\delta(1 + JR + J\delta), \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sum_{j=1}^J I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\| \right) \leq 2JA\delta^{-\alpha_{max}}$$

Rearranging, we get

$$\begin{aligned} H \left(\delta, \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sum_{j=1}^J I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\| \right) &\leq 2AJ \left(\frac{-JR + 1 + \sqrt{(JR + 1)^2 + 4\delta J}}{2J} \right)^{-\alpha_{max}} \\ &\leq 2AJ \left(\frac{\sqrt{2\delta J^{3/2}}}{1 + JR} \right)^{-\alpha_{max}} \end{aligned}$$

(Used the fact that for $b > 0$ small enough, $\sqrt{a^2 + b} - a \geq \sqrt{(a + \frac{b}{4a})^2} - a = \frac{b}{4a}$)

Lemma 6:

Suppose ϵ_i are sub-gaussian errors and for the function class \mathcal{F} , we have that for some $0 < \alpha < 2$, $A' > 0$, and $J > 0$

$$H(\delta, \mathcal{F}, \|\cdot\|_T) \leq A' J^\tau \delta^{-\alpha} \quad \forall \delta > 0$$

Then for $T = 2C_1 C A'^{1/2} J^{\tau/2} 2^{1-\alpha/2}$

$$Pr \left(\sup_{f \in \mathcal{F}} \frac{\left| \frac{1}{\sqrt{n}} \sum_{i=1}^n \epsilon_i f(z_i) \right|}{\|f\|_n^{1-\alpha/2}} \geq T \right) \leq c \exp(-T^2/c^2)$$

Proof: Follow proof for Lemma 8.4 in Vandegeer, but with $A = A' J^{-\alpha}$. Note that we then have $A_0 = A'^{1/2} J^{\tau/2}$. We then get

$$Pr \left(\sup_{f \in \mathcal{F}} \frac{\left| \frac{1}{\sqrt{n}} \sum_{i=1}^n \epsilon_i f(z_i) \right|}{\|f\|_n^{1-\alpha/2}} \geq 2C_1 C A'^{1/2} J^{\tau/2} 2^{1-\alpha/2} \right) \leq c \exp(-T^2/c^2)$$

Note that we can write via shorthand that

$$\sup_{f \in \mathcal{F}} \frac{\left| \frac{1}{n} \sum_{i=1}^n \epsilon_i f(z_i) \right|}{\|f\|_n^{1-\alpha/2}} = O_p(J^{\tau/2} n^{-1/2})$$

Result 3: Single λ , Multiple Penalties, Optimal $\tilde{\lambda}_T$ over X_T

Consider function classes \mathcal{G}_j that are cones. Also, suppose we have an additive model:

$$y = \sum_{j=1}^J g_j^* + \epsilon$$

where $g_j^* \in \mathcal{G}_j$.

We fit the model by least squares with separate penalties for each function g_j :

$$\{\hat{g}_j\}_{j=1}^J = \arg \min_{g_j \in \mathcal{G}_j} \|y - \sum_{j=1}^J g_j\|_T^2 + \lambda^2 \sum_{j=1}^J I_j^{v_j}(g_j)$$

Suppose $v_j \geq 1$ for all j . (This requirement on v_j stricter than Vandegeer Thrm 10.2)
 Suppose for all j , there is some $0 < \alpha < 2$ s.t. for all $\delta > 0$,

$$H(\delta, \{g_j \in \mathcal{G}_j : I(g_j) \leq 1\}, \|\cdot\|_T) \leq A\delta^{-\alpha}$$

and that for all j

$$\sup_{g_j \in \mathcal{G}_j} \frac{\|g_j\|_T}{I(g_j)} \leq R < \infty$$

If we choose λ s.t.

$$\tilde{\lambda}_T^{-1} = O_p\left(n^{1/(2+\alpha)}\right) \left(J + \sum_{j=1}^J I_j^{v_j}(g_j^*)\right)^{(2-\alpha)/2(2+\alpha)}$$

then

$$\left\| \sum_{j=1}^J g_j - g_j^* \right\|_T = O_p\left(\tilde{\lambda}_T\right) J \left(\sum_{j=1}^J I_j^{v_j}(g_j^*)\right)^{1/2}$$

and

$$\sum_{j=1}^J I_j(\hat{g}_j) \leq O_p(J) \left(J + \sum_{j=1}^J I_j^{v_j}(g_j^*)\right)$$

Proof:

The basic inequality gives us:

$$\left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T^2 + \lambda^2 \sum_{j=1}^J I_j^{v_j}(\hat{g}_j) \leq 2 \left| \left(\epsilon_T, \sum_{j=1}^J \hat{g}_j - g_j^* \right) \right| + \lambda^2 \sum_{j=1}^J I_j^{v_j}(g_j^*)$$

Case 1: $\left| \left(\epsilon_T, \sum_{j=1}^J \hat{g}_j - g_j^* \right) \right| \leq \lambda^2 \sum_{j=1}^J I_j^{v_j}(g_j^*)$

$$\left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T \leq O_p(\lambda) \left(\sum_{j=1}^J I_j^{v_j}(g_j^*)\right)^{1/2}$$

Case 2: $\left| \left(\epsilon_T, \sum_{j=1}^J \hat{g}_j - g_j^* \right) \right| \geq \lambda^2 \sum_{j=1}^J I_j^{v_j}(g_j^*)$

By Lemma 3,

$$H\left(\delta, \left\{ \frac{\sum_{j=1}^J g_j - g_j^*}{\sum_{j=1}^J I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\|_T\right) \leq \tilde{A} J^{1-\alpha} \delta^{-\alpha}$$

Hence by (10.6) in Vandegeer,

$$\sup_{g_j \in \mathcal{G}_j} \frac{\left| \left(\epsilon_T, \sum_{j=1}^J g_j - g_j^* \right) \right|}{\left\| \sum_{j=1}^J g_j - g_j^* \right\|^{1-\alpha/2} \left(\sum_{j=1}^J I(g_j) + I(g_j^*) \right)^{\alpha/2}} = O_p\left(n^{-1/2}\right) J^{1-\alpha}$$

and the basic inequality becomes

$$\left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T^2 + \lambda^2 \sum_{j=1}^J I_j^{v_j}(\hat{g}_j) \leq O_p\left(n^{-1/2}\right) J^{1-\alpha} \left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T^{1-\alpha/2} \left(\sum_{j=1}^J I(\hat{g}_j) + I(g_j^*)\right)^{\alpha/2}$$

Case 2a: Suppose $\sum_{j=1}^J I(\hat{g}_j) \leq \sum I(g_j^*)$.
Then

$$\left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T \leq O_p \left(n^{-1/(2+\alpha)} \right) J^{\frac{2(1-\alpha)}{\alpha+2}} \left(\sum_{j=1}^J I(g_j^*) \right)^{\alpha/(2+\alpha)}$$

Case 2b: Suppose $\sum_{j=1}^J I(\hat{g}_j) \geq \sum I(g_j^*)$.

First note that by assuming $v_j \geq 1$, we must have $I_j(\hat{g}_j) \leq I_j^{v_j}(\hat{g}_j) + 1$. So

$$\begin{aligned} \sum_{j=1}^J I_j(\hat{g}_j) &\leq J + \sum_{j=1}^J I_j^{v_j}(\hat{g}_j) \\ &\leq J + O_p \left(n^{-1/2} \right) J^{1-\alpha} \lambda^{-2} \left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T^{1-\alpha/2} \left(\sum_{j=1}^J I(\hat{g}_j) \right)^{\alpha/2} \end{aligned}$$

Case 2ba: If the second term on the RHS in the inequality above is bigger, then

$$\sum_{j=1}^J I_j(\hat{g}_j) \leq O_p \left(n^{-1/(2-\alpha)} \right) J^{2(1-\alpha)/(2-\alpha)} \lambda^{-4/(2-\alpha)} \left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T$$

which implies

$$\left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T \leq O_p \left(n^{-1/(2-\alpha)} \right) J^{\alpha(1-\alpha)/(2-\alpha)} \lambda^{-2\alpha/(2-\alpha)}$$

and

$$\sum_{j=1}^J I_j(\hat{g}_j) \leq J^{(\alpha+2)(1-\alpha)/(2-\alpha)} \left(J + \sum_{j=1}^J I_j^{v_j}(g_j^*) \right)$$

Case 2bb: If the first term on the RHS in the inequality above is bigger, then

$$\sum_{j=1}^J I_j(\hat{g}_j) \leq J \implies \left\| \sum_{j=1}^J \hat{g}_j - g_j^* \right\|_T \leq O_p \left(n^{-1/(2+\alpha)} \right) J^{(2-\alpha)/(2+\alpha)}$$

Result 5: Multiple λ , Multiple Penalties, Optimal λ on X_T

Consider an additive model:

$$y = \sum_{j=1}^J g_j^* + \epsilon$$

We fit the model by least squares with separate penalties and separate λ for each function g_j :

$$\{\hat{g}_j\}_{j=1}^J = \arg \min_{g_j \in \mathcal{G}_j} \|y - \sum_{j=1}^J g_j\|_T^2 + \frac{1}{J} \sum_{j=1}^J \lambda_j^2 I_j^{v_j}(g_j)$$

Suppose $v_j > \frac{2\alpha_j}{2+\alpha_j}$ for all j .

Suppose for all j , there is some $0 < \alpha_j < 2$ s.t. for all $\delta > 0$,

$$H \left(\delta, \left\{ \frac{g_j - g_j^*}{I(g_j) + I(g_j^*)} : g_j \in \mathcal{G}_j, I(g_j) + I(g_j^*) > 0 \right\}, \|\cdot\|_T \right) \leq \frac{A}{J} \delta^{-\alpha_j}$$

and for all j ,

$$\sup_{g_j \in \mathcal{G}_j} \frac{\|g_j - g_j^*\|_T}{I(g_j) + I(g_j^*)} \leq R < \infty$$

If we choose λ s.t.

$$\tilde{\lambda}_j^{-1} = ???$$

then

$$\left\| \sum_{j=1}^J g_j - g_j^* \right\|_T = ???$$

and

$$\sum_{j=1}^J I^{v_j}(\hat{g}_{\lambda,j}) = ???$$

Example 1: Sobelov norm (NOT DONE)

Consider the functions

$$\mathcal{G} = \left\{ g : [0, 1] \mapsto \mathbb{R} : \int_0^1 g^{(m)}(z)^2 dz < \infty \right\}$$

Suppose x_i are all unique. Then the Sobelov norm for the class $\{\hat{g}_\lambda \in \mathcal{G} : \lambda \in \Lambda\}$ is bounded above by its $L_2(P_n)$ norm.

$$I^2(\hat{g}_\lambda) = \int_0^1 \left(\hat{g}_\lambda^{(m)}(z) \right)^2 dz \leq 2\|\hat{g}_\lambda\|_n^2 + 4I^2(\tilde{g}) + 4\|y\|_n^2 \quad \forall \lambda \in \Lambda$$

PROBLEM: as defined, it is possible that $I^2(\tilde{g})$ grows with n , which is not okay!

Proof:

Let \tilde{g} satisfy $\tilde{g}(x_i) = y_i$ and have the smallest value for $\int_0^1 \left(\tilde{g}^{(m)}(z) \right)^2 dz$. This function \tilde{g} should always exist.

Case 1: $\lambda \leq 1/2$

By definition of \hat{g}_λ

$$\|y - \hat{g}_\lambda\|_n^2 + \lambda^2 I^2(\hat{g}_\lambda) \leq \|y - (\tilde{g} - \lambda \hat{g}_\lambda)\|_n^2 + \lambda^2 I^2(\tilde{g} - \lambda \hat{g}_\lambda)$$

Note that

$$\begin{aligned} I^2(\tilde{g} - \lambda \hat{g}_\lambda) &= \int_0^1 \left(\tilde{g}^{(m)} - \lambda \hat{g}_\lambda^{(m)} \right)^2 dz \\ &= 2 \int_0^1 \max \left(\left| \tilde{g}^{(m)} \right|^2, \left| \lambda \hat{g}_\lambda^{(m)} \right|^2 \right) dz \\ &= 2 \left(\int_0^1 \left| \tilde{g}^{(m)} \right|^2 dz + \int_0^1 \left| \lambda \hat{g}_\lambda^{(m)} \right|^2 dz \right) \end{aligned}$$

Hence

$$\lambda^2 I^2(\hat{g}_\lambda) \leq \lambda^2 \|\hat{g}_\lambda\|_n^2 + 2\lambda^2 I^2(\tilde{g}) + 2\lambda^4 I^2(\hat{g}_\lambda)$$

The following ineq follows, where the RHS is maximized when $\lambda = 1/2$

$$I^2(\hat{g}_\lambda) \leq \frac{\lambda^2}{\lambda^2 - 2\lambda^4} (\|\hat{g}_\lambda\|_n^2 + 2I^2(\tilde{g})) \leq 2\|\hat{g}_\lambda\|_n^2 + 4I^2(\tilde{g})$$

Case 2: $\lambda > 1/2$

By definition of \hat{g}_λ

$$\|y - \hat{g}_\lambda\|_n^2 + \lambda^2 I^2(\hat{g}_\lambda) \leq \|y\|_n^2$$

The RHS is maximized when $\lambda = 1/2$, so

$$I^2(\hat{g}_\lambda) \leq 4\|y\|_n^2$$

Hence we have an upper bound for the Sobelov norm

$$I^2(\hat{g}_\lambda) \leq 2\|\hat{g}_\lambda\|_n^2 + 4I^2(\tilde{g}) + 4\|y\|_n^2$$

Appendix

A cute lemma I found but never used: Supposing that $I^v(\hat{g}_\lambda)$ is continuous in λ , then given training data T ,

$$\frac{\partial}{\partial \lambda} L_T(\hat{g}_\lambda, \lambda) = 2\lambda I^v(\hat{g}_\lambda)$$

Also, L_T is convex in λ .

Proof:

By definition,

$$L_T(\hat{g}_\lambda, \lambda) = \|y - \hat{g}_\lambda\|_T^2 + \lambda^2 I^v(\hat{g}_\lambda) \leq \|y - \hat{g}_{\lambda'}\|_T^2 + \lambda^2 I^v(\hat{g}_{\lambda'}) = L_T(\hat{g}_{\lambda'}, \lambda)$$

Then we can provide upper and lower bounds for $L_T(\hat{g}_{\lambda_2}, \lambda_2) - L_T(\hat{g}_{\lambda_1}, \lambda_1)$:

$$\begin{aligned} L_T(\hat{g}_{\lambda_2}, \lambda_2) - L_T(\hat{g}_{\lambda_1}, \lambda_1) &\leq L_T(\hat{g}_{\lambda_1}, \lambda_2) - L_T(\hat{g}_{\lambda_1}, \lambda_1) \\ &= \|y - \hat{g}_{\lambda_1}\|_T^2 + \lambda_2^2 I^v(\hat{g}_{\lambda_1}) - \|y - \hat{g}_{\lambda_1}\|_T^2 - \lambda_1^2 I^v(\hat{g}_{\lambda_1}) \\ &= (\lambda_2^2 - \lambda_1^2) I^v(\hat{g}_{\lambda_1}) \end{aligned}$$

$$\begin{aligned} L_T(\hat{g}_{\lambda_2}, \lambda_2) - L_T(\hat{g}_{\lambda_1}, \lambda_1) &\geq L_T(\hat{g}_{\lambda_2}, \lambda_2) - L_T(\hat{g}_{\lambda_2}, \lambda_1) \\ &= \|y - \hat{g}_{\lambda_2}\|_T^2 + \lambda_2^2 I^v(\hat{g}_{\lambda_2}) - \|y - \hat{g}_{\lambda_2}\|_T^2 - \lambda_1^2 I^v(\hat{g}_{\lambda_2}) \\ &= (\lambda_2^2 - \lambda_1^2) I^v(\hat{g}_{\lambda_2}) \end{aligned}$$

So suppose WLOG $\lambda_2 > \lambda_1$:

$$(\lambda_2 + \lambda_1) I^v(\hat{g}_{\lambda_2}) \leq \frac{L_T(\hat{g}_{\lambda_2}, \lambda_2) - L_T(\hat{g}_{\lambda_1}, \lambda_1)}{\lambda_2 - \lambda_1} \leq (\lambda_2 + \lambda_1) I^v(\hat{g}_{\lambda_1})$$

So as $\lambda_1 \rightarrow \lambda_2 = \lambda$, we have by the sandwich theorem,

$$\frac{\partial}{\partial \lambda} L_T(\hat{g}_\lambda, \lambda) = 2\lambda I^v(\hat{g}_\lambda)$$

Furthermore, given training data T

$$\frac{\partial}{\partial \lambda} L_T(\hat{g}_\lambda, \lambda) = \frac{\partial}{\partial \lambda} \|y - \hat{g}_\lambda\|_T^2 + 2\lambda I^v(\hat{g}_\lambda) + \lambda^2 \frac{\partial}{\partial \lambda} I^v(\hat{g}_\lambda)$$

then, combining this with the lemma, we have that

$$\frac{\partial}{\partial \lambda} \|y - \hat{g}_\lambda\|_T^2 = -\lambda^2 \frac{\partial}{\partial \lambda} I^v(\hat{g}_\lambda)$$

Finally, to see that L_T is convex in λ , note that

$$\frac{\partial^2}{\partial \lambda^2} L_T(\hat{g}_\lambda, \lambda) = 2I^v(\hat{g}_\lambda) + 2\lambda v I^{v-1}(\hat{g}_\lambda) \frac{\partial}{\partial \lambda} I(\hat{g}_\lambda) > 0$$

since $\frac{\partial}{\partial \lambda} I(\hat{g}_\lambda) > 0$.