CV Proof: Mitchell's Thrm

Consider the CV problem with

$$\hat{\lambda} = \arg\min_{\lambda} \sum_{k=1}^{K} \|y - \hat{g}_{\lambda}(\cdot|D_{-k})\|_{k}^{2}$$

We consider the behavior of

$$\hat{g}_{MCV} = \sum_{k=1}^{K} \hat{g}_{\lambda}(\cdot|D_{-k})$$

and its generalization error

$$E_D \|\hat{g}_{MCV}(\cdot|D) - g^*\|^2 = E_D \left[\int (\hat{g}_{MCV}(x|D) - g^*(x))^2 d\mu(x) \right]$$

We will assume that $\sup_{g\in\mathcal{G}}\|g\|_{\infty}\leq G,$ so Assumption A.1 and A.2 are satisfied. Suppose the errors are bounded ($\|\epsilon\|_{\infty}<\infty$). Suppose we can show

$$||g_{\lambda_1} - g_{\lambda_2}||_{\infty} \le |\lambda_1 - \lambda_2| n^{\kappa} C$$

So we get that

$$E\left[\|g^* - \frac{1}{k}\sum_{k=1}^K g_{\hat{\lambda}}(\cdot|D_{-k})\|^2\right] \le (1+a)E\left[\|g^* - \frac{1}{k}\sum_{k=1}^K g_{\hat{\lambda}}(\cdot|D_{-k})\|^2\right] + c_aJ^2 \max_{k=1:K} \frac{(\log n)^2}{n_k}$$

Proof

First note that the assumption A.2 is satisfied:

$$\int (g^*(x) - g_{\lambda}(x|D_0))^4 d\mu(x) \leq \|(g^* - g_{\lambda})^2\|_1 \|(g^* - g_{\lambda})^2\|_{\infty}$$

$$\leq 4G^2 \|g^* - g_{\lambda}\|_2^2$$

Let

$$Q(T) = \{q_{\lambda}(x, y) = (g^* - \hat{g}_{\lambda})^2 : \lambda \in \Lambda\}$$

To compute the upper bound, we need the $\|\cdot\|_{\psi_1}$ and $\|\cdot\|_{L_2}$ entropy of the function class

$$\mathcal{Q}_d^{L_2}(T) = \left\{ q \in \mathcal{Q}(T) : \|Q(Z)\|_2 \le \sqrt{d} \right\}$$

From our assumptions, we note that

$$\begin{aligned} \|q_{\lambda_1} - q_{\lambda_2}\|_{\psi_1} &= \|(g^* - g_{\lambda_1})^2 - (g^* - g_{\lambda_2})^2\|_{\psi_1} \\ &= \|(g_{\lambda_1} - g_{\lambda_2})(2g^* - g_{\lambda_1} - g_{\lambda_2})\|_{\psi_1} \\ &\leq \|g_{\lambda_1} - g_{\lambda_2}\|_{\infty} \|2g^* - g_{\lambda_1} - g_{\lambda_2}\|_{\infty} \\ &\leq |\lambda_1 - \lambda_2| \, 4Cn^{\kappa}G \end{aligned}$$

and similarly for $\|\cdot\|_{L_2}$ (we're using a loose bound for $\|\cdot\|_{L_2}$ since calculating its inverse/convex conjugate is difficult).

Hence (for a different constant κ)

$$H(u, \mathcal{Q}_d^{L_2}(T), \|\cdot\|_{\psi_1}) \le J\left(\log\frac{1}{u} + \log(4Cn^{\kappa}G)\right)$$

and

$$H(u, \mathcal{Q}_d^{L_2}(T), \|\cdot\|_{L_2}) \le J\left(\log\frac{1}{u} + \log(4Cn^{\kappa}G)\right)$$

We calculate each component of the complexity term J(d):

$$\begin{array}{lcl} \gamma_1(\mathcal{Q}_d^{L_2}(T), \|\cdot\|_{\psi_1}) & = & \int_0^G H(u, \mathcal{Q}_d^{L_2}(T), \|\cdot\|_{\psi_1}) du \\ \\ & = & JG \left(1 + \log(4Cn^{\kappa}G)\right) \end{array}$$

and

$$\begin{split} \gamma_2(\mathcal{Q}_d^{L_2}(T), \|\cdot\|_{L_2}) &= \int_0^{\sqrt{d}} \left[H(u, \mathcal{Q}_d^{L_2}(T), \|\cdot\|_{L_2}) \right]^{1/2} du \\ &= \sqrt{d} \int_0^1 \left[J\left(\log \frac{1}{u} + \log(4Cn^{\kappa}G)\right) \right]^{1/2} du \\ &\leq \sqrt{d} \left[\int_0^1 J\left(\log \frac{1}{u} + \log(4Cn^{\kappa}G)\right) du \right]^{1/2} \\ &= \sqrt{d} \left[J\left(1 + \log(4Cn^{\kappa}G)\right) \right]^{1/2} \end{split}$$

So we can define

$$J(d) \equiv J\left(1 + \log(4Cn^{\kappa}G)\right) \left[\sqrt{d} + \left(\max_{k=1:K} \frac{\log n_k}{\sqrt{n_k}}\right)\right]$$

$$\geq \sqrt{d} \left[J\left(1 + \log(4Cn^{\kappa}G)\right)\right]^{1/2} + \left(\max_{k=1:K} \frac{\log n_k}{\sqrt{n_k}}\right) JG\left(1 + \log(4Cn^{\kappa}G)\right)$$

Then $J^{-1}(b)$ is

$$J^{-1}(b) = \left(\frac{b}{J(1 + \log(4Cn^{\kappa}G))} - \left(\max_{k=1:K} \frac{\log n_k}{\sqrt{n_k}}\right)\right)^2$$

The convex conjugate of $J^{-1}(b)$ is

$$\psi(v) = \frac{1}{2} \left(vJ \left(1 + \log(4Cn^{\kappa}G) \right) \right)^2 + J \left(1 + \log(4Cn^{\kappa}G) \right) \left(\max_{k=1:K} \frac{\log n_k}{\sqrt{n_k}} \right) v$$

Therefore

$$\begin{split} \epsilon_q(1/q) & = & \psi \left(\max_{k=1:K} \frac{2q(1+a)}{a\sqrt{n_k}} \right) \\ & = & \frac{1}{2} \left(\frac{2q(1+a)}{a\sqrt{n_k}} J \left(1 + \log(4Cn^{\kappa}G) \right) \right)^2 + J \left(1 + \log(4Cn^{\kappa}G) \right) \left(\max_{k=1:K} \frac{\log n_k}{\sqrt{n_k}} \right) \frac{2q(1+a)}{a\sqrt{n_k}} \\ & \leq & \max_{k=1:K} \left(\frac{2q(1+a)}{a\sqrt{n_k}} J \left(1 + \log(4Cn^{\kappa}G) \right) \right)^2 \end{split}$$

Finally, we get

$$E\left[\|g^* - \frac{1}{k}\sum_{k=1}^K g_{\hat{\lambda}}(\cdot|D_{-k})\|^2\right] \leq (1+a)E\left[\|g^* - \frac{1}{k}\sum_{k=1}^K g_{\hat{\lambda}}(\cdot|D_{-k})\|^2\right] + \frac{c(1+a)^2}{a}\frac{J^2\left(1 + \log(4Cn^{\kappa}G)\right)^2}{\min_{k=1:K} n_k}$$