

$$1] \quad \hat{R}(h) = \frac{1}{n} \sum_{j=1}^n L(h(X_j), Y_j)$$

to show: $P(\lim_{h \rightarrow \infty} \hat{R}(h) = R(h)) = 1$

$$A_n(\varepsilon) = \{ \omega \mid | \underbrace{\hat{R}(h)}_{= \hat{R}(h)} - R(h) | > \varepsilon \}$$

$$= \frac{1}{n} \sum_{j=1}^n \underbrace{L(h(X_j), Y_j)}_{Z_j \in [0,1]}$$

$$= \hat{R}(h)$$

$$= R(h)$$

$$\Rightarrow P(| \underbrace{\bar{Z}_n}_{= \hat{R}(h)} - \underbrace{E[\bar{Z}_n]}_{= R(h)} | > \varepsilon) \leq 2e^{-2n\varepsilon^2}$$

$$\sum_{n=1}^{\infty} P(A_n(\varepsilon)) \leq 2 \sum_{n=1}^{\infty} e^{-2n\varepsilon^2} < \infty$$

Borel-Cantelli
 \Rightarrow

$$P(\lim_{n \rightarrow \infty} \hat{R}(h) = R(h)) = 1$$

$$\begin{aligned} 2] \quad R(\hat{h}) - \inf_{h \in \mathcal{H}} R(h) &\leq \sup_{h \in \mathcal{H}} \underbrace{R(\hat{h}) - R(h)}_{\hat{R}(\hat{h}) - \hat{R}(h) + R(\hat{h}) - \hat{R}(\hat{h})} \\ &\quad \underbrace{\hat{R}(\hat{h}) - \hat{R}(h)}_{\leq 0} + \hat{R}(h) - R(h) \\ &\leq 2 \sup_{h \in \mathcal{H}} |R(h) - \hat{R}(h)| \end{aligned}$$

$$\mathcal{Z} = \mathcal{X} \times \mathcal{Y}, \quad z_i = (x_i, y_i), \quad z = (x, y)$$

$$f(z_1, \dots, z_n) = \underbrace{R(h) - \frac{1}{n} \sum_{i=1}^n L(h(x_i), y_i)}_{\text{empirical risk}}$$

$$|f(z_1, \dots, z_n) - f(z_1, \dots, z'_n)| \leq \frac{1}{n} \leq \frac{2 R_n(L, \mathcal{H})}{R(h) - \hat{R}(h)}$$

$$\Rightarrow \mathbb{P}(R(\tilde{h}) - \hat{R}(\tilde{h}) > \underbrace{\sup_{h \in \mathcal{H}} R(h) - \hat{R}(h)}_{\text{empirical risk}} + t)$$

$$\leq \mathbb{P}(R(\tilde{h}) - \hat{R}(\tilde{h}) > \mathbb{E}[R(\tilde{h}) - \hat{R}(\tilde{h})] + t) \\ \leq e^{-2nt^2}$$

$$\Rightarrow \mathbb{P}(R(h) - \hat{R}(h) > 2\beta_n(L, \mathcal{H}) + t) \leq e^{-2nt^2}$$

analog: $\mathbb{P}(\hat{R}(h) - R(h) > 2\beta_n(L, \mathcal{H}) + t)$

$$\Rightarrow \mathbb{P}(\sup_{h \in \mathcal{H}} |R(h) - \hat{R}(h)| > 2\beta_n(L, \mathcal{H}) + t)$$

$$\leq \underbrace{2e^{-2nt^2}}_{=\delta} \hookrightarrow t = \sqrt{\frac{\log(2/\delta)}{2n}}$$

$$3) H \in \mathcal{H}_p : H = \underbrace{a}_{\mathbb{R}^n} + \text{span} \underbrace{\{v_1, \dots, v_p\}}_{\text{orthonormal}}$$

$$P_H : \|x - P_H(x)\|^2 = \inf_{v \in H} \|x - v\|^2$$

$$= \min_{b \in \mathbb{R}} \|x - (a + \sqrt{b})\|^2 = \underbrace{\inf_{b \in \mathbb{R}} \|x - (a + \sqrt{b})\|^2}_{= I(a, V, B)}$$

$$\Rightarrow \min_{H \in \mathcal{H}_p} \sum_{j=1}^m \|x_j - P_H(x_j)\|^2 = \min_{a, V, B} \sum_{j=1}^m \|x_j - (a + \sqrt{b_j})\|^2$$

$$\text{wlog } \sum_{j=1}^m b_j = 0$$

$$\text{with } B = (b_1, \dots, b_m) \in \mathbb{R}^{p \times m}$$

$$V = (v_1, \dots, v_p) \in \mathbb{R}^{n \times p}$$

$$\nabla_a J(a, V, B) = -2 \underbrace{\sum_{j=1}^m (x_j - a - V b_j)}_{=0} \stackrel{!}{=} 0$$

$$= -2(m\bar{x} - ma)$$

$$\Rightarrow a = \bar{x} = \frac{1}{m} \sum_{j=1}^m x_j$$

$$\nabla_{b_k} J(a, V, B) = -2 v^T (x_k - \overset{= \bar{x}}{a} - V b_k) \stackrel{!}{=} 0$$

$$= -2 (v^T (x_k - \bar{x}) - \underbrace{v^T V}_{= I_p} b_k)$$

$$b_k = V^T (x_k - \bar{x})$$

$$\leadsto \min_{V_1, \dots, V_p} \sum_{j=1}^m \|x_j - (\bar{x} + V V^T (x_j - \bar{x}))\|^2$$

$$= \min_{V_1, \dots, V_p} \sum_{j=1}^m \|y_j - V V^T y_j\|^2$$

$$\text{with } y_j = x_j - \bar{x}$$

$$\sum_{j=1}^m y_j = 0$$

Lecture (PrA)

$$\hookrightarrow v_k = u_k$$

$$\sum_{j=1}^m \|y_j - VV^T y_j\|^2 = \text{tr}((Y - VV^T Y)^T (Y - VV^T Y))$$

$$V^T V = I = \underbrace{\text{tr}(Y^T Y)} - \text{tr}(Y^T V V^T Y)$$

$$= \text{tr}(Y Y^T) = \text{tr}(C)$$

$$= \text{tr}(U \Lambda U^T) = \text{tr}(\Lambda)$$

$$= \sum_{i=1}^n \lambda_i$$

$$\text{tr}(Y^T V V^T Y) = \sum_{i=1}^p \langle v_i, C v_i \rangle \longrightarrow \max_{\substack{V^T V = I_p}} !$$

