Homework 3 10/36-702 Due Friday March 31 at 3:00 pm

1 Reproducing kernel Hilbert spaces

(a) Let \mathcal{H} be a RKHS generated by a kernel K. Consider training data $(X_1, Y_1), \dots, (X_n, Y_n)$ where $Y_i \in \{-1, +1\}$. Let $\lambda \ge 0$ be a fixed real number, and let \hat{f} minimize $\sum_{i=1}^n L(f(X_i), Y_i) + \lambda ||f||_K^2$. Show that \hat{f} has the form

$$\widehat{f}(x) = \sum_{i=1}^{n} \widehat{\alpha}_i K(X_i, X)$$

where $\hat{\alpha} = (\hat{\alpha}_1, \dots, \hat{\alpha}_n)$ minimizes

 $Q(\mathbb{K}\alpha) + \lambda \, \alpha^T \mathbb{K}\alpha$

for some function *Q* where \mathbb{K} is an $n \times n$ matrix with $\mathbb{K}(j,k) = K(X_j, X_k)$.

(b) Let \mathscr{F} be the set of all functions $f : [0,1] \to \mathbb{R}$ such that f(x) = ax for some real number *a*. Show that this is a RKHS with kernel K(x, y) = xy.

2 Basic inequalities

- (a) Show that there exists a random variable $X \ge 0$ such that Markov's inequality is an equality.
- (b) Show that there exists a random variable Y such that Chebyshev's inequality is an equality.

3 Chernoff's method

Recall Chernoff's method where we use the fact that $P(X > \delta) \le \inf_{t>0} e^{-t\delta} \mathbb{E}[e^{tX}]$. Let $X \ge 0$. Suppose that the moment generating function for X exists. Let $\delta > 0$. Show that

$$\inf_{k=0,1,2,\dots} \frac{\mathbb{E}[X^k]}{\delta^k} \le \inf_{t>0} e^{-t\delta} \mathbb{E}[e^{tX}].$$

(Hence, Chernoff's method does not necessarily give the tightest possible bounds.)

4 Sub-Gaussian variables

A random variable is sub-Gaussian if there exists a > 0 such that for all $t \in \mathbb{R}$

$$\mathbb{E}[e^{t(X-\mu)}] \le e^{\frac{a^2t^2}{2}}.$$

- (a) Let X be Rademacher. Thus P(X = 1) = P(X = -1) = 1/2. Show that X is sub-Gaussian with a = 1.
- (b) Show that every bounded random variable is sub-Gaussian.
- (c) Show that, if *X* is sub-Gaussian then

$$P(|X - \mu| \ge t) \le 2e^{-\frac{t^2}{2a^2}}$$

(d) Let *X* be sub-Gaussian with mean 0. Show that there exists b > 0 and c > 0 such that

$$\mathbb{P}(|X| \ge t) \le c \mathbb{P}(|Z| \ge t)$$

for all $t \ge 0$, where $Z \sim N(0, b^2)$.

You may use the following fact without proof: Let $Z \sim N(0, 1)$ and z > 0. Let $\phi(z)$ be the standard Normal density. Then $P(Z \ge z) \ge \phi(z)(1/z - 1/z^3)$.

Hint: Consider the ratio $\mathbb{P}(X \ge t)/\mathbb{P}(Z \ge t)$. Now consider two cases: (i) $0 \le t \le 2\sigma$ and (ii) $t > 2\sigma$.

5 Inequality on a convex function

Let $\phi : \mathbb{R} \to \mathbb{R}$ be a convex, non-decreasing function. Let \mathscr{F} be a class of functions. Let $\mathscr{G} = \{f - \mathbb{E}[f(X)] : f \in \mathscr{F}\}$. Let

$$\xi_n = \sup_{f \in \mathcal{G}} \frac{1}{n} \sum_{i=1}^n \sigma_i f(X_i)$$

where $\sigma_1, \ldots, \sigma_n$ are Rademacher. Show that

$$\mathbb{E}\left[\phi\left(\sup_{f\in\mathscr{F}}|P_n(f)-P(f)|\right)\right] \geq \mathbb{E}\left[\phi\left(\frac{1}{2}\xi_n\right)\right].$$

6 Rademacher complexity

Let \mathscr{F} denote all indicator functions of finite subsets of [0,1]. Let P be the uniform distribution on [0,1]. Show that $\operatorname{Rad}_n(\mathscr{F}) \ge 1/2$, where

$$\operatorname{Rad}_n(\mathscr{F}) = \mathbb{E}\left(\sup_{f\in\mathscr{F}}\frac{1}{n}\sum_{i=1}^n\sigma_if(Z_i)\right).$$

7 Minimax rate

Let $X_1, \ldots, X_n \sim \text{Unif}(0, \theta)$ where $0 < \theta < M$ for some constant M. Let

$$R_n = \inf_{\widehat{\theta}} \sup_{\theta \in (0,M)} \mathbb{E}_{\theta} |\widehat{\theta} - \theta|$$

be the minimax risk for estimating θ . Show that $R_n \approx 1/n$. (Recall that $a_n \approx b_n$ means that both a_n/b_n and b_n/a_n are bounded for all large n.)

8 Hellinger distance

The Hellinger distances between distributions P and Q is

$$H(P,Q) = \sqrt{\int (\sqrt{p} - \sqrt{q})^2}$$

where P has density p and Q has density q.

(a) Let TV denote the total variation distance. Show that

$$TV(P,Q) \leq H(P,Q)\sqrt{1-\frac{H^2(P,Q)}{4}}.$$

You may use the following fact without proof: $TV(P,Q) = ||P - Q||_1/2$.

(b) Let X_1, \ldots, X_n be iid. Let $p^n \equiv p(x_1, \ldots, x_n) = \prod_i p(x_i)$ and $q^n = q(x_1, \ldots, x_n) = \prod_i q(x_i)$ be joint density functions. Let $p_i = p(x_i)$ and $q_i = q(x_i)$. Show that

$$H(p^{n},q^{n}) = \sqrt{2} \sqrt{1 - \prod_{i=1}^{n} (1 - (1/2)H^{2}(p_{i},q_{i}))}$$