10/36-702 Homework 2 Due: Friday 3/1/2013

<u>Instructions</u>: Hand in your homework to Michelle Martin (GHC 8001) before 3:00 pm on Friday 3/1/2013.

- 1. Let $X_1, \ldots, X_n \sim P$ where P has density p and $0 \leq X_i \leq 1$. Find the asymptotic bias of $\hat{p}_h(0)$ where \hat{p}_h is the kernel density estimator. Assume that $p \in \Sigma(2, L)$ and assume a Gaussian kernel. Note: because 0 is a boundary point, the bias is different than the bias we computed in class.
- 2. Let $X_1, \ldots, X_n \sim P$ where P has density p and $X_i \in \mathbb{R}$. Let

$$\widehat{p}_h(x) = \frac{1}{n} \sum_{i=1}^n \frac{1}{h} K\left(\frac{x - X_i}{h}\right)$$

be the usual kernel density estimator. Assume that $h = h_n$ is such that $h_n \to 0$ and $nh_n \to \infty$ as $n \to \infty$.

(a) Assuming only that p(x) is continuous in a neighborhood of x, show that

$$\widehat{p}_h(x) \xrightarrow{P} p(x).$$

(b) Let $p_h(x) = \mathbb{E}(\widehat{p}_h(x))$. It can be shown that

$$\frac{\widehat{p}_h(x) - p_h(x)}{\operatorname{se}_n(x)} \rightsquigarrow N(0, 1)$$

where $\operatorname{se}_n(x) = \sqrt{\operatorname{Var}(\widehat{p}_h(x))}$. Note that $\operatorname{se}_n(x)$ is the standard error of the mean as $\widehat{p}_h(x) = \frac{1}{n} \sum_{i=1}^n Z_i$ where

$$Z_i = \frac{1}{h} K\left(\frac{x - X_i}{h}\right).$$

Let s^2 be the sample variance of Z_1, \ldots, Z_n :

$$s^{2} = \frac{1}{n} \sum_{i=1}^{n} (Z_{i} - \overline{Z}_{n})^{2}.$$

Let $\widehat{\operatorname{se}}_n(x) = s/\sqrt{n}$. Show that

$$\frac{\widehat{\operatorname{se}}_n(x)}{\operatorname{se}_n(x)} \xrightarrow{P} 1$$

and that

$$\frac{\widehat{p}_h(x) - p_h(x)}{\widehat{\operatorname{se}}_n(x)} \rightsquigarrow N(0, 1).$$

Note: keep in mind that $h = h_n$ is changing with n. You may make extra assumptions if needed.

(c) Suppose that h_n takes the form of the optimal bandwidth, e.g., $h_n = C n^{-1/5}$ for $p \in \Sigma(2, L)$. Show that

$$\frac{\widehat{p}_h(x) - p(x)}{\operatorname{se}_n(x)} \rightsquigarrow N(b(x), 1)$$

for some function b(x).

(d) Continuing from part (c), suppose we use the confidence interval

$$C_n = [\widehat{p}_h(x) - z_{\alpha/2} \,\widehat{se}_n(x), \ \widehat{p}_h(x) + z_{\alpha/2} \,\widehat{se}_n(x)].$$

Explain why $P(p(x) \in C_n)$ does not converge to $1 - \alpha$ as $n \to \infty$. (In fact, no one really knows how to compute a confidence interval for a density estimator.)

3. Let \widehat{m} be a linear smoother so that $\widehat{m}(x) = \sum_{i} \ell_i(x) Y_i$ and $\widehat{Y} = LY$. Prove the leave-one-out identity:

$$\frac{1}{n}\sum_{i=1}^{n}(Y_{i}-\widehat{m}_{(-i)}(X_{i}))^{2}=\frac{1}{n}\sum_{i=1}^{n}\left(\frac{Y_{i}-\widehat{m}(X_{i})}{1-L_{ii}}\right)^{2}.$$

4. Let \mathcal{H} be a Hilbert space of functions. Suppose that the evaluation functionals $\delta_x f = f(x)$ are continuous. Show that \mathcal{H} is a reproducing kernel Hilbert space and find the kernel.