

Math 6070-1: Spring 2013  
Solutions for problem set 3

1. Let  $X_1, X_2, \dots$  be an i.i.d. sample from a density function  $f$ . We assume that  $f$  is differentiable in an open neighborhood  $V$  of a fixed point  $x$ , and  $B := \max_{z \in V} |f'(z)| < \infty$ .

(a) Prove that for all  $\lambda > 0$ ,  $m \geq 1$ , and all  $x \in \mathbf{R}$ ,

$$\mathbb{P} \left\{ \min_{1 \leq j \leq m} |X_j - x| \geq \lambda \right\} = \left[ 1 - \int_{x-\lambda}^{x+\lambda} f(z) \, dz \right]^m.$$

**Solution:** Let  $\mathcal{M}_j$  denote the event that  $|X_j - x| \geq \lambda$ . Then,

$$\begin{aligned} \mathbb{P} \left\{ \min_{1 \leq j \leq m} |X_j - x| \geq \lambda \right\} &= \mathbb{P}(\mathcal{M}_1 \cap \dots \cap \mathcal{M}_m) \\ &= \mathbb{P}(\mathcal{M}_1) \times \dots \times \mathbb{P}(\mathcal{M}_m) \\ &= [\mathbb{P}(\mathcal{M}_1)]^m, \end{aligned}$$

since  $X_1, \dots, X_m$  are i.i.d. The claim follows, since

$$\mathbb{P}(\mathcal{M}_1) = 1 - \mathbb{P}\{|X_1 - x| < \lambda\} = 1 - \int_{x-\lambda}^{x+\lambda} f(z) \, dz.$$

(b) Prove that for all  $\epsilon > 0$  small enough,

$$\max_{z \in [x-\epsilon, x+\epsilon]} |f(x) - f(z)| \leq 2B\epsilon.$$

Use this to estimate  $|\int_{x-\epsilon}^{x+\epsilon} f(z) \, dz - 2\epsilon f(x)|$ .

**Solution:** The first bound is from freshman calculus. Namely, we use the fundamental theorem of calculus to see that

$$f(x) - f(z) = \int_z^x f'(w) \, dw.$$

Therefore, by the triangle inequality for integrals,

$$\begin{aligned} |f(x) - f(z)| &\leq \int_z^x |f'(w)| \, dw \\ &\leq B|x - z|. \end{aligned}$$

This yields the bound

$$\max_{z \in [x-\epsilon, x+\epsilon]} |f(x) - f(z)| \leq B\epsilon,$$

which is slightly better than the one we are asked to derive.

Next we use a trick that we used in the context of density estimation [for convolutions]. Namely, we write

$$\int_{x-\epsilon}^{x+\epsilon} f(z) dz - 2\epsilon f(x) = \int_{x-\epsilon}^{x+\epsilon} [f(z) - f(x)] dz.$$

Then, apply the triangle inequality for integrals in order to obtain

$$\begin{aligned} \left| \int_{x-\epsilon}^{x+\epsilon} f(z) dz - 2\epsilon f(x) \right| &= \left| \int_{x-\epsilon}^{x+\epsilon} [f(z) - f(x)] dz \right| \\ &\leq \int_{x-\epsilon}^{x+\epsilon} |f(z) - f(x)| dz \\ &\leq 2B\epsilon^2. \end{aligned}$$

(c) Suppose that as  $m \rightarrow \infty$ ,  $\lambda_m \rightarrow \infty$  and  $\lambda_m^2/m \rightarrow 0$ . Then, prove that

$$\lim_{m \rightarrow \infty} \frac{-1}{2\lambda_m} \ln \mathbb{P} \left\{ \min_{1 \leq j \leq m} |X_j - x| \geq \frac{\lambda_m}{m} \right\} = f(x).$$

**Solution:** Throughout, let us write

$$\delta_m := \int_{x-(\lambda_m/m)}^{x+(\lambda_m/m)} f(z) dz.$$

We know that  $\lim_{m \rightarrow \infty} \delta_m = 0$ , thanks to freshman calculus. In fact,

$$\delta_m = \frac{2\lambda_m}{m} f(x) \pm \frac{2B\lambda_m^2}{m^2},$$

thanks to part (b).

Part (a) shows that

$$\frac{-1}{2\lambda_m} \ln \mathbb{P} \left\{ \min_{1 \leq j \leq m} |X_j - x| \geq \frac{\lambda_m}{m} \right\} = \frac{-m}{2\lambda_m} \ln(1 - \delta_m).$$

According to Taylor's expansion, from freshman calculus,

$$\begin{aligned} \ln(1 - \delta_m) &= -\delta_m + \frac{\delta_m^2}{2} - \dots \\ &= -\frac{2\lambda_m}{m} f(x) \mp \frac{2B\lambda_m^2}{m^2} + \frac{1}{2} \left( \frac{2\lambda_m}{m} f(x) \pm \frac{2B\lambda_m^2}{m^2} \right)^2 \\ &= -\frac{2\lambda_m}{m} f(x) \mp \frac{(2B + [2f(x)]^2)\lambda_m^2}{m^2} + \dots \end{aligned}$$

The point is that the error of approximation is bounded above by some finite constant times  $\lambda_m^2/m^2$ . In particular,

$$\lim_{m \rightarrow \infty} \frac{-m}{2\lambda_m} \ln(1 - \delta_m) = f(x),$$

which is another way to rewrite the statement.

(d) *Devise an estimator of  $f(x)$  based on the previous steps.*

**Solution:** Suppose  $X_1, \dots, X_n$  is i.i.d. data from pdf  $f$ , and suppose that we can choose and fix an integer  $1 \ll m \ll n$  such that  $n$  divides  $m$ . Block your data into  $(n/m) - 1$  sub-blocks of length  $m$ ; that is,

$$\underbrace{X_1, \dots, X_m}_{\text{first block}}, \underbrace{X_{m+1}, \dots, X_{2m}}_{\text{second block}}, \dots, \underbrace{X_{n-m+1}, \dots, X_n}_{\text{last block}}$$

These blocks are i.i.d. Moreover, we know from part (c) that

$$P \left\{ \text{Every term in the } j\text{th block is at least } \frac{\lambda_m}{m} \text{ away from } x \right\} \approx e^{-\lambda_m f(x)},$$

provided only that  $m$  is large. By the law of large numbers, since  $n \gg m$ , and since there are  $(n/m) - 1 \approx (n/m)$  blocks,

$$\frac{m}{n} \sum_{j=1}^{(n/m)-1} I \left\{ \text{Every term in the } j\text{th block is at least } \frac{\lambda_m}{m} \text{ away from } x \right\} \approx e^{-\lambda_m f(x)}.$$

Therefore, a reasonable estimator is

$$\hat{f}(x) := \frac{-1}{\lambda_m} \ln \left( \frac{m}{n} \sum_{j=1}^{(n/m)-1} I \left\{ \text{Every term in the } j\text{th block is at least } \frac{\lambda_m}{m} \text{ away from } x \right\} \right).$$

If you cannot find such an  $(n, m)$  pair exactly, then divide the data into  $m$  blocks that are equally-sized, inasmuch as possible. Any such division works just as well as any other, and works as in the preceding.