## Homework 8

Due November 19 at 3:00.

## Searching for solutions is forbidden.

(1) Let  $Y \in \mathbb{R}$ ,  $X \in \mathbb{R}^d$  and  $A \in \{0,1\}$ . Assuming there is no unmeasured confounding. We have that, for  $a \in \{0,1\}$ ,

$$\mathbb{E}[Y(a)] = \int \mu(x, a) p(x) dx$$

where  $\mu(x, a) = \mathbb{E}[Y|X = x, A = a] = \int yp(y|x, a)dy$ . Show that

$$\mathbb{E}[Y(a)] = \mathbb{E}\left[\frac{YI(A=a)}{\pi(A|X)}\right]$$

where  $\pi(a|x) = P(A = a|X = x)$ . Here, I(A = a) = 1 if A = a and I(A = a) = 0 otherwise.

Hint: You may take X and Y to be discrete if you wish. In that case, the integrals become sums.

Explain how would you use this formula to estimate  $\mathbb{E}[Y(a)]$ .

(2) Download the dataset SAheart.csv. The variables are

sbp systolic blood pressure tobacco cumulative tobacco (kg)

ldl low density lipoprotein cholesterol

adiposity

family history of heart disease (Present, Absent)

typea type-A behavior

obesity

alcohol current alcohol consumption

age age at onset

chd response, coronary heart disease

(Remove the first column which is just row numbers.)

- (a) Examine the data using exploratory data analysis (i.e. plots).
- (b) Do a logistic regression of chd on the other variables. Summarize the results.
- (c) Suppose we want to estimate the causal effect of Age on chd. Assume we have measured all confounding variables. Estimate the causal effect. To do this, let Y denote chd, let A denote age and let X denote the other variables. We want to estimate the function  $\psi(a) = \mathbb{E}[Y(a)]$ , where  $a \in \mathbb{R}$ . We can restrict attention to  $15 \le a \le 64$  since our data are in that range. Then recall that the plugin estimator is

$$\widehat{\psi}(a) = \frac{1}{n} \sum_{i} \widehat{\mu}(X_i, a)$$

where  $\widehat{\mu}(x,a)$  is your estimate of  $\mu(x,a) = \mathbb{E}[Y|X=x,A=a]$  which you have from the logistic regression. Plot the function  $\widehat{\psi}(a)$ .

(d) We would like to get a confidence interval for  $\psi(a)$ . To do this, you can use the bootstrap which works as follows: Draw a sample of size n from your original data. For example:

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n = length(Y)
I = sample(1:n,size=n,replace=TRUE)
XX = X[I,]
YY = Y[I]
AA = A[I]
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After you draw the bootstrap sample, estimate  $\psi(a)$  again using the bootstrap sample. Repeat this process B=1000 times to get  $\widehat{\psi}_1(a),\ldots,\widehat{\psi}_B(a)$ . For each a, let  $\ell(a)$  be the 2.5 percentile of  $\widehat{\psi}_1(a),\ldots,\widehat{\psi}_B(a)$  and let u(a) be the 97.5 percentile of  $\widehat{\psi}_1(a),\ldots,\widehat{\psi}_B(a)$ . This gives a 95 percent confidence band for  $\psi(a)$ . Plot  $\widehat{\psi}(a)$  along with  $\ell(a)$  and u(a).