## Group Assignments

## 12.1 Group Assignment 1

- 1. Suppose that A and B are independent events. Show that  $A^c$  and  $B^c$  are independent.
- 2. The probability that a child has brown hair is 1/4. Assume independence between children and assume there are three children.
  - (a) If it is known that at least one child has brown hair, what is the probability that at least two children have brown hair?
  - (b) If it is known that the oldest child has brown hair, what is the probability that at least two children have brown hair?
- 3. Let (X,Y) be uniformly distributed on the unit disc,  $\{(x,y) \in \mathbb{R}^2 \mid x^2 + y^2 \leq 1\}$ . Set  $R = \sqrt{X^2 + Y^2}$ . What is the CDF and PDF of R?
- 4. A fair coin is tossed until a head appears. Let X be the number of tosses required. What is the expected value of X?
- 5. Let  $X_1, \ldots, X_n$  be i.i.d. from Bernoulli(p).
  - (a) Let  $\alpha > 0$  be fixed and define

$$\epsilon_n = \sqrt{\frac{1}{2n} \log \frac{2}{\alpha}}.$$

Let  $\hat{p}_n = \frac{1}{n} \sum_{i=1}^n X_i$  and define the confidence interval  $I_n = [\hat{p}_n - \epsilon_n, \hat{p}_n + \epsilon_n]$ . Use Hoeffding's inequality to show that  $P(p \in I_n) \ge 1 - \alpha$ .

- (b) Let  $\alpha = 0.05$  and p = 0.4. Conduct a simulation study to see how often the confidence interval  $I_n$  contains p (called coverage). Do this for n = 10, 100, 1000, 10000. Plot the coverage as a function of n.
- (c) Plot the length of the confidence interval as a function of n.
- (d) Say that  $X_1, \ldots, X_n$  represents if a person has a disease or not. Let us assume that unbeknownst to us the true proportion of people with the disease has changed from p = 0.4 to p = 0.5. We use

the confidence interval to make a decision, that is, when presented with evidence (samples) we calculate  $I_n$  and our decision is that the true proportion of people with the disease is in  $I_n$ . Conduct a simulation study to answer the following question: Given that the true proportion has changed, what is the probability that our decision is correct? Again using n = 10, 100, 1000, 10000.

## 12.2 Group Assignment 2

1. Consider a supervised learning problem where we assume that  $Y \mid X$  is Poisson distributed. That is, the conditional density of  $Y \mid X$  is given by

$$f_{Y|X}(y,x) = \frac{\lambda^y e^{-\lambda}}{y!}, \quad \lambda(x) = \exp(\alpha \cdot x + \beta).$$

Here  $\alpha$  is a vector (slope) and  $\beta$  is a number (intercept). Follow the calculations from Section 4.2.1 to derive a loss that needs to be minimized with respect to  $\alpha$  and  $\beta$ . Note: do we really need the factorial term?

- 2. Let  $X_1, \ldots, X_n$  be i.i.d. from Uniform $(0, \theta)$ . Let  $\hat{\theta} = \max(X_1, \ldots, X_n)$ . First, find the distribution function of  $\hat{\theta}$ . Then compute the bias $(\hat{\theta})$ , se $(\hat{\theta})$  and  $MSE_n(\hat{\theta})$ .
- 3. Consider the continuous distribution with density

$$p(x) = \frac{1}{2}\cos(x), \quad -\frac{\pi}{2} < x < \frac{\pi}{2}.$$

- (a) Find the distribution function F.
- (b) Find the inverse distribution function  $F^{-1}$ .
- (c) To sample using an Accept-Reject sampler, Algorithm 1, we need to find a density g such that  $p(x) \leq Mg(x)$  for some M > 0. Find such a density g and find the value of M.
- 4. Let  $Y_1, Y_2, \ldots, Y_n$  be a sequence of i.i.d. discrete random variables, where  $P(Y_i = 0) = 0.1$ ,  $P(Y_i = 1) = 0.3$ ,  $P(Y_i = 2) = 0.2$ , and  $P(Y_i = 3) = 0.4$ . Let  $X_n = \max\{Y_1, \ldots, Y_n\}$ . Let  $X_0 = 0$  and verify that  $X_0, X_1, \ldots, X_n$  is a Markov chain. Find the transition matrix P.

5. Let  $X_1, \ldots, X_n$  be i.i.d. from some distribution F that is unknown. Let  $\hat{F}_n$  be the empirical distribution function, use this to find an estimate of the p quantile of F (call it q). Use Theorem 5.28 to find a confidence interval for q.

## 12.3 Group Assignment 3

1. Consider a three-state (1, 2, 3) Markov chain with transition matrix

$$P = \begin{pmatrix} 0.5 & 0.5 & 0 \\ 0.5 & 0 & 0.5 \\ 0.5 & 0 & 0.5 \end{pmatrix}.$$

- (a) Draw the transition diagram.
- (b) Find the stationary distribution  $\pi$ .
- (c) Given that the chain is in state 1 at time 1, what is the probability that the chain is in state 2 at time 4?
- (d) Given that the chain is in state 1 at time 1, what is the expected time until the chain is in state 3 for the first time?
- (e) What is the period of each state?
- 2. Assume that we are trying to classify a binary outcome Y, i.e., our data is of the form  $(X,Y) \sim F_{X,Y}$ , where  $Y \in \{0,1\}$  and  $X \in \mathbb{R}^d$ . We have used data to train a classifier g(X). We can evaluate the performance of the classifier using i.i.d. testing data,  $(X_1, Y_1), \ldots, (X_n, Y_n)$ . We are interested in estimating the following quantities:

Precision : 
$$P(Y = 1 | g(X) = 1)$$
,  
Recall :  $P(g(X) = 1 | Y = 1)$ .

- (a) Write down the empirical version of the precision and recall.
- (b) Let us now think that the variable Y denotes if a battery's health has deteriorated or not, and let X denote a bunch of constructed health indicators about the battery. If the model g(X) predicts that the battery has deteriorated, you need to run a test to confirm this. The cost of running the test is c when the battery is not deteriorated. On the other hand, if the battery is in fact deteriorated and the test is not run, the battery will die during use

- and the cost of this is d. Define a random variable representing the cost of the decision g(X) and write down the formula for the expected cost in terms of the precision and recall.
- (c) Advanced question: Can you produce a confidence interval for the expected cost? What about the precision and recall?
- 3. Let X and Y be two d-dimensional zero mean, unit variance Gaussian random vectors. Show that X and Y are nearly orthogonal by calculating their dot product. Can you, for instance, also bound the probability that the dot product is larger than  $\epsilon$ ?
- 4. Let  $u_1, \ldots, u_r$  be  $n \times 1$  unit length vectors that are linearly independent, i.e.,

$$\sum_{i=1}^{r} \alpha_i u_i = 0 \implies \alpha_i = 0 \text{ for all } i.$$

- (a) Verify that the matrix  $u_i u_i^T$  is a rank one matrix for all i. What is the null-space and range of  $u_i u_i^T$ ?
- (b) Verify that the matrix  $U = \sum_{i=1}^{r} u_i u_i^T$  is a rank r matrix.
- (c) i. If we perform SVD on U, are the vectors  $u_1, \ldots, u_r$  the same as the right singular vectors? If not, can you give an example?
  - ii. What if the vectors  $u_1, \ldots, u_r$  are all orthogonal? In this case, what are the singular values of U?
- 5. Let  $X \sim \text{Uniform}(B_1)$  and define  $Y = ||X||_2$  (the Euclidean norm).
  - (a) Find the distribution function of Y.
  - (b) What is the distribution of  $\ln(1/Y)$ ?
  - (c) Calculate  $E[\ln(1/Y)]$ , first by using the distribution function of Y and then by using the distribution function of  $\ln(1/Y)$ .