Assignment 2

15-359 Probability and Computing

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Section: B

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Problem 1: Cell Block (exercise)

By definition of expected value,

$$E\left(\frac{X}{Y} \mid X^2 + Y \le 10\right) = \frac{\left(\frac{1}{3} * \frac{1}{12} + \frac{2}{3} * \frac{2}{12} + \frac{1}{8} * \frac{3}{12}\right)}{\frac{6}{12}}$$
$$= \boxed{\frac{49}{144}}.$$

Problem 2: Friend of a friend (exercise)

By definition of conditional probability and conditional independence,

$$\begin{split} P(A|B\cap C) &= \frac{P(A\cap B\cap C)}{P(B\cap C)} \\ &= \frac{P(A\cap B|C)\cdot P(C)}{P(B\cap C)} \\ &= \frac{P(A|C)\cdot P(B|C)\cdot P(C)}{P(B\cap C)} \\ &= \frac{P(A|C)\cdot P(B\cap C)\cdot P(C)}{P(B\cap C)\cdot P(C)} \\ &= P(A|C). \quad \blacksquare \end{split}$$

Problem 3: Expecting something different? (exercise)

Let X be a non-negative, discrete, integer-valued random variable. Let $A = \{(i, j) \in \mathbb{N} \times \mathbb{N} \mid i \geq 1, j < i\} = \{(i, j) \in \mathbb{N} \times \mathbb{N} \mid j \geq 0, i > j\}$. By definition of expected value,

$$E(X) = \sum_{i=0}^{\infty} iP(X=i) = \sum_{i=1}^{\infty} iP(X=i)$$

$$= \sum_{i=1}^{\infty} \sum_{j=0}^{i-1} P(X=i) = \sum_{(i,j)\in A} P(X=i) = \sum_{j=0}^{\infty} \sum_{i=j+1}^{\infty} P(X=i)$$

$$= \sum_{j=0}^{\infty} P(X>j). \quad \blacksquare$$

Problem 4: Big data (exercise)

A. Let n be the number of files in the database, and let $\{F_i\}_{i=1}^n$ be a decreasing sequence of the file sizes in the database. Suppose, for sake of contradiction, that, for some $m \geq \frac{n}{2}$, for $i \in \{1, 2, ..., m\}$, $F_i > 12K$. Then, the average file size A of the files in the database is given by:

$$A = \frac{1}{n} \sum_{i=1}^{n} F_i \ge \frac{1}{n} \sum_{i=1}^{m} F_i > \frac{1}{n} \sum_{i=1}^{m} 12K = \frac{1}{n} m \cdot 12K \ge \frac{1}{n} \frac{n}{2} 12K = 6K,$$

contradicting the given that A = 6K.

B. Let n and $\{F_i\}_{i=1}^n$ be as in the solution to part A. Suppose, for sake of contradiction, that, for some $m \geq \frac{n}{3}$, for some $i \in \{1, 2, ..., m\}$, $F_i > 12K$. Then, the average file size A of the files in the database is given by:

$$A = \frac{1}{n} \sum_{i=1}^{n} F_i = \frac{1}{n} \left(\sum_{i=1}^{m} F_i + \sum_{i=m+1}^{n} F_i \right)$$

$$> \frac{1}{n} \left(\sum_{i=1}^{m} 12K + \sum_{i=m+1}^{n} 3K \right)$$

$$= \frac{1}{n} \left(m \cdot 12K + (n-m) \cdot 3K \right) = \frac{1}{n} \left(m \cdot (9K) + n \cdot 3K \right)$$

$$\geq \frac{1}{n} \left(\frac{n}{3} (9K) + n \cdot 3K \right) = 6K,$$

contradicting the given that A = 6K. Therefore, at fewer than a third of the files can have size > 12K.

Problem 5: Making a stack of coins out of fish

Let $\lambda = pn$, so that $p = \lambda/n$. Let X be a random variable such that $X \sim \text{Binomial}(n, p)$. Since

$$\lim_{t \to \infty} \left(\frac{n!}{n^{k+i}(n-k-i)!} \right) = 1,$$

by the Binomial Theorem,

$$\lim_{t \to \infty} \left(\frac{n!}{(n-k)!n^k} (1-p)^{n-k} \right) = \lim_{t \to \infty} \left(\frac{n!}{(n-k)!n^k} \left(1 - \frac{-\lambda}{n} \right)^{n-k} \right)$$

$$= \lim_{t \to \infty} \left(\frac{n!}{(n-k)!n^k} \sum_{i=0}^{n-k} \binom{n-k}{i} \left(\frac{-\lambda}{n} \right)^i \right)$$

$$= \lim_{t \to \infty} \left(\sum_{i=0}^{n-k} \frac{(n-k)!}{(n-k)!} \frac{n!}{n^{k+i}(n-k-i)!} \frac{(-\lambda)^i}{i!} \right)$$

$$= \sum_{i=0}^{\infty} \frac{(-\lambda)^i}{i!} = e^{-\lambda}.$$

$$\lim_{t \to \infty} (P_X(k)) = \lim_{t \to \infty} \left(\binom{n}{k} p^k (1-p)^{n-k} \right)$$

$$= \lim_{t \to \infty} \left(\frac{\lambda^k n!}{k! (n-k)! n^k} (1-p)^{n-k} \right)$$

$$= \frac{\lambda^k e^{-\lambda}}{k!}.$$

Thus, for large n, the binomial distribution Binomial(n,p) is well-approximated by the Poisson distribution Poisson(np).

Problem 6: Coffee-theorem automata

Let X_1 be a random variable denoting the time (in hours) the student takes to get home from work, and let X_2 be a random variable denoting the time (in hours) the student takes to get home from the coffee shop. Let $W = E(X_1)$, and let $C = E(X_2)$. Then, conditioning on whether the student goes home or goes to get coffee,

$$W = 1 * \frac{1}{3} + (1+C) * \frac{2}{3}.$$

Similarly, conditioning on whether the student goes back to work or stays at the coffee house,

$$C = (1+W) * \frac{1}{3} + (1+C) * \frac{2}{3}.$$

Since this gives a system of two linear equations in two variables, we can solve the system, yielding W = 9, C = 12. Thus, the expected time until the student goes home is 9 hours.

Problem 7: Expecting to be astonished

- A. Suppose that X is a constant random variable, taking only a single value x_1 with $P(X = x_1) = 1$. Then, $a(P(X = x_1)) = a(1) = 0$, so that $C(X) = P(X = x_1) \cdot a(P(X = x_1)) = 0$. Suppose, on the other hand, that X takes at least two values, including some distinct x_1 and x_2 , so that $0 < P(X = x_1), P(X = x_2) < 1$. Then, $C(X) \ge P(X = x_1) \cdot a(P(X = x_1)) + P(X = x_2) \cdot a(P(X = x_2)) > 1$, since a > 0 on (0, 1). Thus, C(X) is non-negative, and zero if and only if X is a constant random variable.
- B. Since

$$\begin{split} \sum_{j=1}^{m} \sum_{i=1}^{n} P(X = x_{i}, Y = y_{i}) a(P(Y = y_{i})) &= \sum_{j=1}^{m} P(Y = y_{i}) a(P(Y = y_{i})) = C(Y), \\ C(X, Y) &= \sum_{i=1}^{n} \sum_{j=1}^{m} P(X = x_{i}, Y = y_{i}) \cdot a(P(X = x_{i}, Y = y_{i})) \\ &= \sum_{i=1}^{n} \sum_{j=1}^{m} P(X = x_{i}, Y = y_{i}) \cdot a(\frac{P(X = x_{i}, Y = y_{i})P(Y = y_{i})}{P(Y = y_{i})}) \\ &= \sum_{i=1}^{n} \sum_{j=1}^{m} P(X = x_{i}, Y = y_{i}) \cdot \log_{2}(\frac{P(X = x_{i}, Y = y_{i})P(Y = y_{i})}{P(X = x_{i}, Y = y_{i})}) \\ &= \sum_{i=1}^{n} \sum_{j=1}^{m} P(X = x_{i}, Y = y_{i}) \cdot \left(\log_{2}(P(Y = y_{i})) + \log_{2}\left(\frac{P(X = x_{i}, Y = y_{i})}{P(Y = y_{i})}\right)\right) \\ &= \sum_{i=1}^{n} \sum_{j=1}^{m} P(X = x_{i}, Y = y_{i}) \cdot (a(P(Y = y_{i})) + a(P(X = x_{i}|Y = y_{i}))) \\ &= \sum_{j=1}^{m} \sum_{i=1}^{n} P(X = x_{i}, Y = y_{i}) \cdot a(P(X = x_{i}|Y = y_{i})) + \sum_{i=1}^{n} P(X = x_{i}, Y = y_{i}) \cdot a(P(Y = y_{i})) \\ &= C(X|Y) + C(Y). \quad \blacksquare \end{split}$$

C. 1/p(X) is a random variable because it is the outcome the experiment of applying the function $X \mapsto \log_2 1/p(X)$ to X, the outcome of an experiment. Furthermore, 1/p(X) can take at most as many values as X, and thus can take only finitely (and thus countably) many values, so that 1/p(X) is discrete. Since $X \log_2 1/p(X) = \log_2 1/p(x)$ whenever X = x, $E(\log_2 1/p(X)) = \sum_{x \in \Omega} P(\log_2 1/p(X)) = \log_2 1/p(x)$.

D. Since $y \mapsto \log_2 1/y$ is a convex, decreasing function, by Jensen's Inequality, $C(X) = E(\log_2 1/p(X)) \ge \log_2 1/E(p(X))$, so that C(X) is maximized when $X = \arg\min_{x \in \Omega} p(x)$, the result in the sample space of least probability.

Problem 8: You may call a k-clause a Klaus

Let $n < 2^{k-1}$ be the number of clauses in the given k-CNF. Suppose we randomly assign values to each of the variables in the given k-CNF. Let X be the event that some clause contains contains all only true or only false variables. For $i \in \{1, 2, ..., n\}$, let X_i be the event that the i^{th} k-clause contains only true or only false variables, so that $X = \bigcup_{i=1}^n X_i$. Then, for each $i \in \{1, 2, ..., n\}$,

$$P(X_i) = \frac{2}{2^k} = \frac{1}{2^{k-1}}.$$

Thus,

$$P(X) = P\left(\bigcup_{i=1}^{n} X_i\right) \le \sum_{i=1}^{n} P(X_i) = \sum_{i=1}^{n} \frac{1}{2^{k-1}} = n \frac{1}{2^{k-1}} < \frac{2^{k-1}}{2^{k-1}} = 1.$$

If Y is the event that every k-clause includes at least one true variable and at least one false variable. Then, $Y = X^c$, so that P(Y) = 1 - P(X) > 0. Since there is a non-zero probability that, under a random assignment of the variables in the given k-CNF, every clause includes at least one true variable and at least one false variable, there exists an assignment of the variables in the given k-CNF with the desired condition.

Problem 9: Shearing off projections (extra credit)