

## Discrete Mathematics Homework #5

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\* Let  $P(E)$  : probability of event  $E$  and let  $E(X)$  : expectation of a random variable  $X$ .

**7.1.38** Let H, T denote heads and tails, respectively. Also note that for the sample space  $S$ ,  $|S| = 8$ , and each element of the sample space have equally likely outcomes.

- (a)  $E_1 = \{THH, THT, TTH, TTT\}$ , and  $E_2 = \{HHH, HHT, THH, THT\}$ . Thus  $P(E_1) = P(E_2) = 4/8 = 1/2$ . And since  $E_1 \cap E_2 = \{THH, THT\}$ ,  $P(E_1 \cap E_2) = 2/8 = 1/4$ , which is equal to  $P(E_1)P(E_2) = 1/4$ .  $E_1, E_2$  are independent.
- (b)  $E_1 = \{THH, THT, TTH, TTT\}$ , and  $E_2 = \{HHT, THH\}$ . Thus  $P(E_1) = 1/2$ ,  $P(E_2) = 1/4$ . Since  $E_1 \cap E_2 = \{THH\}$ ,  $P(E_1 \cap E_2) = 1/8$ , which is equal to  $P(E_1)P(E_2) = 1/8$ .  $E_1, E_2$  are independent.
- (c)  $E_1 = \{HTH, HTT, TTH, TTT\}$ , and  $E_2 = \{HHT, THH\}$ . Thus  $P(E_1) = 1/2$ ,  $P(E_2) = 1/4$ . Since  $E_1 \cap E_2 = \emptyset$ ,  $P(E_1 \cap E_2) \neq P(E_1)P(E_2)$ . Thus  $E_1, E_2$  are dependent.

□

**7.2.24** Let

$A$  : First flip comes up tails when a fair coin is flipped 5 times.

$B$  : Exactly 4 heads appear when a fair coin is flipped 5 times.

We want to calculate  $P(B | A)$ .

$P(A) = 1/2$  since we only care about the outputs of the first flip, and  $P(A \cap B) = 1/32$  since THHHH is the only possible case. From the definition,  $P(B | A) = P(A \cap B)/P(A) = 1/16$ . □

**7.2.34** The probability of success is  $p$ , and failure is  $1 - p$ .

- (a) Fails for  $n$  times. Thus  $(1 - p)^n$ .
- (b) Since  $1 = P(\text{No successes}) + P(\text{At least 1 success})$ , the wanted probability is  $1 - (1 - p)^n$ .
- (c) For the case  $n = 0$ , it is trivial that the answer is 1. For  $n > 0$ , the only possible cases are: No successes, and only 1 success. The probability of 1 success is  $np(1 - p)^{n-1}$  since we fail for  $n - 1$  times and succeed once, and we can succeed on any  $n$  trials, out of  $n$  trials. Thus the wanted probability is:  $(1 - p)^n + np(1 - p)^{n-1}$ .
- (d) For the case  $n = 0$ , it is trivial that the answer is 0. For  $n > 0$ , since  $1 = P(\text{At most 1 success}) + P(\text{At least 2 successes})$ , the wanted probability is equal to  $1 - (\text{answer of (c)})$ . Thus the wanted probability is  $1 - [(1 - p)^n + np(1 - p)^{n-1}]$ .

□

**7.3.14** From Bayes' theorem,

$$\begin{aligned} P(F_2|E) &= \frac{P(E \cap F_2)}{P(E)} = \frac{P(E|F_2)P(F_2)}{P(E \cap F_1) + P(E \cap F_2) + P(E \cap F_3)} \\ &= \frac{P(E|F_2)P(F_2)}{P(E|F_1)P(F_1) + P(E|F_2)P(F_2) + P(E|F_3)P(F_3)} \\ &= \frac{3/8 \cdot 1/2}{2/7 \cdot 1/6 + 3/8 \cdot 1/2 + 1/2 \cdot 1/3} = \frac{7}{15} \end{aligned}$$

□

**7.3.22 (a)** We can estimate that  $P(S) = s/(s+h)$ , and  $P(\bar{S}) = h/(s+h)$ .

**(b)** Let  $W$  be the event that an incoming message contains the word  $w$ . We want to calculate  $P(S|W)$ . By Bayes' theorem,

$$\begin{aligned} P(S|W) &= \frac{P(S \cap W)}{P(W)} = \frac{P(W|S)P(S)}{P(W \cap S) + P(W \cap \bar{S})} = \frac{P(W|S)P(S)}{P(W|S)P(S) + P(W|\bar{S})P(\bar{S})} \\ &= \frac{p(w) \cdot \frac{s}{s+h}}{p(w) \cdot \frac{s}{s+h} + q(w) \cdot \frac{h}{s+h}} = \frac{sp(w)}{sp(w) + hq(w)} \end{aligned}$$

□

**7.4.18** Using that  $X, Y$  are non-negative and by the definition of  $Z$ ,  $Z(s) \leq X(s) + Y(s)$ . Thus,

$$\mathbf{E}(Z) = \sum_{s \in S} Z(s)P(s) \leq \sum_{s \in S} (X(s) + Y(s))P(s) = \sum_{s \in S} X(s)P(s) + \sum_{s \in S} Y(s)P(s) = \mathbf{E}(X) + \mathbf{E}(Y)$$

□

**7.4.38 (a)** Let  $X$  be the number of cans filled on a day. By Markov's inequality,

$$P(X > 11000) < \frac{\mathbf{E}(X)}{11000} = \frac{10000}{11000} = \frac{10}{11}$$

**(b)** By Chebyshev's inequality,

$$\begin{aligned} P(9000 < X < 11000) &= P(|X - \mathbf{E}(X)| < 1000) = 1 - P(|X - \mathbf{E}(X)| \geq 1000) \\ &\geq 1 - \frac{\mathbf{V}(X)}{1000^2} = \frac{999}{1000} \end{aligned}$$

□

**7.4.48** Define a random variable  $X_i$  as follows.

$$X_i = \begin{cases} 1 & \text{if } i\text{-th ball is in the first bin} \\ 0 & \text{otherwise} \end{cases}$$

And define  $X = \sum_{i=1}^m X_i$ . Then the expected number of balls that fall into the first bin is  $\mathbf{E}(X)$ .

By the linearity of expectation,  $\mathbf{E}(X) = \sum_{i=1}^m \mathbf{E}(X_i)$ . Now we calculate  $\mathbf{E}(X_i)$ .

By definition,  $\mathbf{E}(X_i) = 1 \cdot P(X_i = 1) + 0 \cdot P(X_i = 0) = 1/n$ . (Clearly  $P(X_i = 1) = 1/n$ , and this situation is symmetric for all  $i$ ). Thus the expected number of balls in the first bin is  $m/n$ . □