

Math for CS I/Discrete Mathematics  
Assignment 6 Solutions

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## I Problem 1 Solution

### 1.1 Part (a)

In this part of the question we have 2 players  $A$  and  $B$  and they take turns rolling a die, they each need different value to win.

$A$  wins with probability  $\alpha$  and  $B$  wins with probability  $\beta$  in a certain rolling attempt.

Let  $A$  be the event that  $A$  wins and  $B$  be the event that  $B$  wins. I have to find the probability of  $A$  being the winner for 2 cases:

**A rolls first:**

*Solution.* The series of events possible in this case are:

$$A, \overline{A}B, \overline{A}\overline{B}A, \overline{A}\overline{B}\overline{A}B, \dots$$

The probability of  $A$  winning i.e.  $P(A)$  is:

$$P(A) = \alpha + \alpha(1-\alpha)(1-\beta) + \alpha(1-\alpha)^2(1-\beta)^2 + \dots$$

$$P(A) = \frac{\alpha}{1 - (1-\alpha)(1-\beta)}$$

$$P(A) = \frac{\alpha}{\alpha + \beta - \alpha\beta}$$

□

**A rolls second:**

*Solution.* The series of events possible in this case are:

$$\overline{A}B, \overline{A}\overline{B}A, \overline{A}\overline{B}\overline{A}B, \dots$$

The probability of  $A$  winning i.e.  $P(A)$  is:

$$P(A) = \alpha(1-\beta) + \alpha(1-\beta)^2(1-\alpha) + \alpha(1-\beta)^3(1-\alpha)^2 + \dots$$

$$P(A) = \frac{\alpha(1-\beta)}{1 - (1-\alpha)(1-\beta)}$$

$$P(A) = \frac{\alpha(1-\beta)}{\alpha + \beta - \alpha\beta}$$

□

### 1.2 Part (b)

*Solution.* 2 coins  $A$  and  $B$  show heads with respective probabilities  $\alpha$  and  $\beta$ .

They are tossed alternately such that  $A$  is tossed first and I have to find the probability such that  $A$  is first to show a head.

This case is same as

$$A, \overline{A}B, \overline{A}\overline{B}A, \overline{A}\overline{B}\overline{A}B, \dots$$

here,  $\overline{A}$  denotes  $A$  not showing head and same applies to  $\overline{B}$ .

$$P(A) = \alpha + \alpha(1-\alpha)(1-\beta) + \alpha(1-\alpha)^2(1-\beta)^2 + \dots$$

$$P(A) = \frac{\alpha}{1 - (1-\alpha)(1-\beta)}$$

$$P(A) = \frac{\alpha}{\alpha + \beta - \alpha\beta}$$

□

### 1.3 Part (c)

*Solution.* Consider the 2 events as below:

$E_1$ : Out of 3 coin tosses, first 2 coins result in same outcome.

$E_2$ : Outcomes of all 3 coin tosses are alike.

In the argument of the problem, it is true that out of 3 coin toss outcomes, atleast 2 must be same and the probability that the third is a head or a tail is  $\frac{1}{2}$  which is independent of other 2.

The mistake in the argument is assuming that the 2 same outcomes are of the first 2 coins.

The probability given by the above argument is  $\frac{1}{2}$  but this probability is  $P(E_2 | E_1)$  and not  $P(E_2)$ .

$$P(E_2 | E_1) = \frac{P(E_2 \cap E_1)}{P(E_1)}$$

$$P(E_2 | E_1) = \frac{P(E_1 | E_2) \cdot P(E_2)}{P(E_1)}$$

$P(E_1) = \frac{1}{2}$ , it can be trivially calculated.

$P(E_1 | E_2) = 1$  as event  $E_1$  lies within event  $E_2$ .

$$P(E_2) = P(E_2 | E_1) \cdot P(E_1)$$

$$P(E_2) = \frac{1}{2} \cdot \frac{1}{2}$$

$$P(E_2) = \frac{1}{4}$$

The above value for the event  $E_2$  is the one which it should be i.e.  $\frac{1}{4}$ , so it is consistent with the argument I made. □

### 1.4 Part (d)

*Solution.* There are 2 players  $A$  and  $B$ ,  $A$  flips  $n + 1$  fair coins and  $B$  flips  $n$  fair coins.

I have to find the probability of  $A$  having more heads than  $B$ , consider the following events:

**E**:  $A$  gets more number of heads than  $B$ .

**F**:  $A$  gets more number of tails than  $B$ .

The events  $E$  and  $F$  are the only possible events as it is not possible for both  $A$  and  $B$  to have equal number of heads since they have unequal number of coins.

If  $A(\text{heads}) > B(\text{heads})$ , it will be event  $E$  and if  $A(\text{heads}) \leq B(\text{heads})$ , it will be event  $F$  and thus, these 2 events are exhaustive.

Also, we will have  $P(E) = P(F)$  as heads and tails can be interchanged.

$$P(E) + P(F) = 1$$

$$2 \cdot P(E) = 1$$

$$P(E) = \frac{1}{2}$$

□

## II Problem 2 Solution

### 2.1 Part (a)

*Solution.* We have a  $s$ -sided fair die, it is rolled  $r$  times, so the possible number of outcomes are  $s^r$ .

Consider the following events:

$E_1$ : Side 1 did not appear in the rolls.

$E_2$ : Side 2 did not appear in the rolls.

$\vdots$

$E_s$ : Side  $s$  did not appear in the rolls.

Probability that atleast 1 side of the side does not show in the  $r$  rolls is  $P(E_1 \cup E_2 \cup \dots \cup E_s)$ .

$$P(E_1 \cup E_2 \cup \dots \cup E_s) = \frac{\binom{s}{1}(s-1)^r - \binom{s}{2}(s-2)^r \dots (-1)^{s-2} \binom{s}{s-1}(s-s+1)^r}{s^r}$$

$$P(E_1 \cup E_2 \cup \dots \cup E_s) = \frac{\sum_{i=1}^{s-1} (-1)^{i-1} \binom{s}{i} (s-i)^r}{s^r}$$

The probability  $p$  that each side has turned up atleast once if it was rolled  $r$  times.

$$p = 1 - P(E_1 \cup E_2 \cup \dots \cup E_s)$$

$$p = \frac{s^r - \sum_{i=1}^{s-1} (-1)^{i-1} \binom{s}{i} (s-i)^r}{s^r}$$

$$p = \frac{\sum_{i=0}^s (-1)^i \binom{s}{i} (s-i)^r}{s^r}$$

□

### 2.2 Part (b)

*Solution.* Assuming there are 365 days in the year of people's birthdays for calculation, even if it is 366 the final answer will not change.

Let  $n$  be the number of students who have announced their birthdays before you, we have:

$$P(n=0) = 0$$

$$P(n=1) = \frac{365}{365} \cdot \frac{1}{365}$$

$$P(n=2) = \frac{365}{365} \cdot \frac{364}{365} \cdot \frac{2}{365}$$

$$P(n) = \frac{\binom{365}{n} n! \binom{n}{1}}{365^{n+1}}$$

I have to find out max value of  $P(n)$ , it will increase with increasing  $n$  till a specific value and then it will keep on decreasing.

$$P(n+1) < P(n)$$

$$\frac{\binom{365}{n+1} (n+1)! \binom{n+1}{1}}{365^{n+2}} < \frac{\binom{365}{n} n! \binom{n}{1}}{365^{n+1}}$$

$$(365-n)(n+1) < 365n$$

$$n^2 + n - 365 > 0$$

The above equation is valid for  $n > 18.611$ , so using the first integer value, the **maxima** will occur at  $n = 19$ .

So, the max probability of winning the prize will be when 19 students before you have announced their birthday.

So, state your birthday at  $20^{th}$  place and you will have best probability of winning. □

### III Problem 3 Solution

$X$  has a mass function  $f(x)$ , i.e.  $f(x) = P(X = x)$ .

$$\Omega = \{\omega_1, \omega_2, \dots, \omega_n\}$$

$$X = \{x_1, x_2, \dots, x_m\}$$

$$P(X = x_i) = P(\omega_j \mid \omega_j \in \Omega, X(\omega_j) = x_i)$$

Using the above definitions of the mass function and probability of  $X$ , I will find the mass function for the given sets of  $X'$  in the following parts.

#### 3.1 Part (a)

*Solution.* Consider the mass function for  $-X$  to be  $g(x)$ .

We will have  $g(x) = P(-X = x)$  or  $g(x) = P(X = -x)$ .

Thus, the function  $g(x)$  is:

$$g(x) = f(-x)$$

□

#### 3.2 Part (b)

*Solution.* Consider the mass function for  $X^+ = \max(0, X)$  to be  $g(x)$ .

$X^+$  will have only non-negative values of  $x$  and all the negative values will result in zero, so the final function will be as below.

$$g(x) = \begin{cases} f(x), & x > 0 \\ \sum_{x \leq 0} f(x), & x = 0 \\ 0, & x < 0 \end{cases}$$

□

#### 3.3 Part (c)

*Solution.* Consider the mass function for  $X^- = \max(0, -X)$  to be  $g(x)$ .

$X^-$  will have only non-negative values of  $x$  and all the positive values will result in zero as it is  $-X$ , so the final function will be as below.

$$g(x) = \begin{cases} f(-x), & x > 0 \\ \sum_{x \leq 0} f(-x), & x = 0 \\ 0, & x < 0 \end{cases}$$

□

#### 3.4 Part (d)

*Solution.* Consider the mass function for  $|X| = X^+ + X^-$  to be  $g(x)$ .

$|X|$  will have only non-negative values of  $x$  and all the values will be result of  $X$  and  $-X$ , so the final function will be as below.

$$g(x) = \begin{cases} f(-x) + f(x), & x > 0 \\ f(0), & x = 0 \\ 0, & x < 0 \end{cases}$$

□

### 3.5 Part (e)

*Solution.* Consider the mass function to be  $g(x)$  for:

$$\text{sgn}(X) = \begin{cases} \frac{X}{|X|}, & X \neq 0 \\ 0, & X = 0 \end{cases}$$

$\text{sgn}(X)$  will have only 3 values of  $x = \{-1, 0, 1\}$  i.e. all negative values will result in  $-1$  and all positive values will result in  $1$ , so the final function will be as below.

$$g(x) = \begin{cases} \sum_{x < 0} f(x), & x = -1 \\ f(0), & x = 0 \\ \sum_{x > 0} f(x), & x = 1 \\ 0, & \text{otherwise.} \end{cases}$$

□

## IV Problem 4 Solution

### 4.1 Part (a)

The probability mass function is  $f(x) = P(X = x)$  by definition and  $x$  here is the number on the card, since the box contains  $n$  cards numbered from 1 to  $n$  and one is picked at random. So the following will be the probability mass function:

$$f(x) = \begin{cases} \frac{1}{n}, & \text{if } x = \{1, 2, \dots, n\}; \\ 0, & \text{otherwise.} \end{cases}$$

*Solution.* I have to find the value of  $E[X]$  which is also called the mean expectation value.

$$\begin{aligned} E[X] &= \sum_{x=1}^n x \cdot f(x) \\ E[X] &= \frac{1}{n} \sum_{x=1}^n x \\ E[X] &= \frac{1}{n} \cdot \frac{n(n+1)}{2} \\ E[X] &= \frac{n+1}{2} \end{aligned}$$

□

*Solution.* I have to find the value of  $E[X^2]$  which is also called the second moment.

$$\begin{aligned} E[X^2] &= \sum_{x=1}^n x^2 \cdot f(x) \\ E[X^2] &= \frac{1}{n} \sum_{x=1}^n x^2 \\ E[X^2] &= \frac{1}{n} \cdot \frac{n(n+1)(2n+1)}{6} \\ E[X^2] &= \frac{(n+1)(2n+1)}{6} \end{aligned}$$

□

*Solution.* I have to find the value of second central moment which is also called variance. The second central moment is  $E[(X - E[X])^2] = E[X^2] - (E[X])^2$ .

$$\begin{aligned} E[(X - E[X])^2] &= E[X^2] - (E[X])^2 \\ E[(X - E[X])^2] &= \frac{(n+1)(2n+1)}{6} - \left(\frac{n+1}{2}\right)^2 \\ E[(X - E[X])^2] &= \frac{n^2 - 1}{12} \end{aligned}$$

□



## 4.2 Part (b)

Assuming  $E[X]$  exists, I have to prove that:

$$(E[X])^2 \leq (E[|X|])^2 \leq E[X^2]$$

*Proof.* We know that  $x \leq |x|$ .

Multiplying the above equation by a non-negative number will not change the validity of the inequality and we know that  $p(x) \geq 0$ , so multiply by it.

We get  $x \cdot p(x) \leq |x| \cdot p(x)$ , since it is valid for every value of  $x$ , we can do summation of it and still the inequality will be valid.

$$\begin{aligned} x \cdot p(x) &\leq |x| \cdot p(x) \\ \sum x \cdot p(x) &\leq \sum |x| \cdot p(x) \\ (\sum x \cdot p(x))^2 &\leq (\sum |x| \cdot p(x))^2 \\ (E[X])^2 &\leq (E[|X|])^2 \end{aligned}$$

By definition of variance for absolute  $x$  which is:

$$Var[|X|] = \sum (|x| - E[|X|])^2 \cdot p(|x|)$$

It is clear from above that  $Var[|X|] \geq 0$ .

$$\begin{aligned} E[X^2] - (E[|X|])^2 &\geq 0 \\ E[X^2] &\geq (E[|X|])^2 \end{aligned}$$

By above equations, I have proved the required result for a discrete random variable.

$$(E[X])^2 \leq (E[|X|])^2 \leq E[X^2]$$

For proving the result for a continuous random variable we can combine the above arguments with **Riemann Integral Hypothesis** i.e. breaking integral into summation of small strips.  $\square$