

Math 152 - Spring 2019
 HW02 - Solution.

1) We flip a fair coin ten times.

a) Define random variables.

with prob. $\frac{1}{2}$

$$X_i = \begin{cases} \text{Head} & \text{if we get head in the } i\text{th flip,} \\ \text{Tail} & \text{otherwise} \end{cases}$$

a) There are 10 flips of which we choose 5 heads, and there are total of 2^{10} ways to flip the coin. Thus, the probability is

$$\frac{\binom{10}{5}}{2^{10}} = \frac{63}{256}$$

b) Let $S = X_1 + \dots + X_{10}$ be the number of heads.

$P(\text{more heads than tails})$

$$= P(S=6) + P(S=7) + P(S=8) + P(S=9) + P(S=10)$$

$$= \sum_{k=6}^{10} P(S=k)$$

$$= \sum_{k=6}^{10} \frac{\binom{10}{k}}{2^{10}}$$

$$= \frac{386}{1024}$$

$$= \frac{47}{128}$$

c) Let $P_4(n)$ be the probability that we have at least 4 consecutive heads after n flips. Then

$$P_4(0) = P_4(1) = P_4(2) = P_4(3) = 0$$

$$\text{and } P_4(4) = \frac{1}{2^4}.$$

For more flips, either we have achieved 4 consecutive heads already or we have a string without 4 consecutive heads followed by THHTHHTH. Hence for $n > 4$,

$$P_4(n) = P_4(n-1) + \frac{1}{2^5} (1 - P_4(n-5)).$$

$$\therefore P_4(10) = P_4(9) + \frac{1}{2^5} (1 - P_4(5)).$$

$$= P_4(8) + \frac{1}{2^5} (1 - P_4(4)) + \frac{1}{2^5} (1 - P_4(4) - \frac{1}{2^5}).$$

$$= P_4(7) + \frac{1}{2^5} (1 - P_4(3)) + \frac{1}{2^5} (1 - P_4(4)) + \frac{1}{2^5} (1 - P_4(4)) - \frac{1}{2^{10}}.$$

$$= \dots = P_4(4) + \frac{1}{2^5} \sum_{k=0}^4 (1 - P_4(k)) + \frac{1}{2^5} (1 - P_4(4)) - \frac{1}{2^{10}}.$$

$$= 2^4 + \frac{1}{2^5} \cdot 3(1-0) + \frac{2}{2^5} (1 - 2^4) - \frac{1}{2^{10}}.$$

2) a) $E[\max(X_1, X_2)]$.

$$\max(X_1, X_2) = \begin{cases} X_1 & \text{if } X_1 \geq X_2 \\ X_2 & \text{if } X_2 < X_1 \end{cases}$$

$$E[\max(X_1, X_2)] = \sum_{j=1}^k j \cdot P(\max(X_1, X_2) = j)$$

~~$\sum_{j=1}^k j$~~ We need to find $P(\max(X_1, X_2) = j)$.

$$\mathbb{P}(\max(X_1, X_2) = j) = \mathbb{P}(X_1 = j, X_2 \leq j) + \mathbb{P}(X_1 < j, X_2 = j)$$

since X_1 and X_2 are independent

$$= \mathbb{P}(X_1 = j) \mathbb{P}(X_2 \leq j) + \mathbb{P}(X_1 < j) \mathbb{P}(X_2 = j)$$

$$= \frac{1}{k} \frac{j}{k} + \frac{j-1}{k} \frac{1}{k}$$

$$= \frac{2j-1}{k^2}$$

$$\therefore \mathbb{E}[\max(X_1, X_2)] = \sum_{j=1}^k j \frac{2j-1}{k^2} = \frac{1}{k^2} \sum_{j=1}^k 2j^2 - \frac{1}{k^2} \sum_{j=1}^k 1$$

$$= \frac{2}{k^2} \sum_{j=1}^k j^2 - \frac{1}{k} = \cancel{\frac{2}{k^2} \frac{k(k+1)}{2}} - \cancel{\frac{k}{k^2}}$$

$$= \frac{2}{k^2} \frac{k(k+1)(2k+1)}{6} - \frac{1}{k} = \cancel{\frac{k+1}{k}}$$

$$= \frac{(k+1)(2k+1)}{3k} - \frac{1}{k}$$

$$\text{b) } \mathbb{E}[\min(X_1, X_2)] = \sum_{j=1}^k j \mathbb{P}(\min(X_1, X_2) = j)$$

$$\Rightarrow \mathbb{P}(\min(X_1, X_2) = j) = \mathbb{P}(X_1 = j, X_2 \geq j) + \mathbb{P}(X_1 > j, X_2 = j)$$

$$= \mathbb{P}(X_1 = j) \mathbb{P}(X_2 \geq j) + \mathbb{P}(X_1 > j) \mathbb{P}(X_2 = j)$$

$$= \frac{1}{k} \frac{k-j+1}{k} + \frac{k-j}{k} \frac{1}{k}$$

$$= \frac{k-j+1+k-j}{k^2}$$

$$= \frac{2(k-j)+1}{k^2}$$

$$\begin{aligned}\therefore \mathbb{E}[\min(X_1, X_2)] &= \sum_{j=1}^k j \cdot \frac{2(k-j)+1}{k^2} \\ &= \dots \quad (\text{simplify it!}).\end{aligned}$$

Math 452 - HW02.

3) Let $X_i = \begin{cases} 1 & \text{if Alice wins, with probability 0.7} \\ 0 & \text{if Alice loses, with probability 0.3} \end{cases}$

(X_i is the random variable indicating whether Alice wins or not in the i th game).

Let $X = X_1 + X_2 + \dots + X_n$. Thus, X is the random variable indicating the number of games Alice wins in the tournament.

Then, $\mathbb{E}[X_i] = 0.7$ and $\mu = \mathbb{E}[X] = n(0.7)$.
 APP Recall that Alice loses the tournament if she wins less than half of the games, i.e., $X \leq \frac{n-1}{2}$.

By Chernoff's bound for $\{0,1\}$ -valued r.v.'s. (See other useful forms in the lecture note):

$$P(X \leq (1-\epsilon)\mu) \leq e^{-\mu\epsilon^2/2}$$

set $(1-\epsilon)\mu = \frac{n-1}{2}$, and find ϵ ,

$$0.7n(1-\epsilon) = 0.5n - 0.5$$

$$0.7n - \epsilon 0.7n = 0.5n - 0.5$$

$$0.2n + 0.5 = \epsilon 0.7n.$$

$$\frac{2}{7} + \frac{5}{7n} = \epsilon$$

$$\Rightarrow \epsilon = \frac{2}{7} + \frac{5}{7n} > \frac{2}{7} \Rightarrow \epsilon^2 > \frac{4}{49}.$$

$$\therefore P(X \leq \frac{n-1}{2}) = P(X \leq (1-\epsilon)\mu) \leq e^{-\mu\epsilon^2/2} < e^{-0.7n(\frac{4}{49})/2}$$

$$\therefore P(X \leq \frac{n-1}{2}) \leq e^{-n/5}$$

2) Let X be the number of times that a 6 occurs over n throws of the die.

$$\text{Let } p = P(X \geq \frac{n}{4}).$$

a) Markov's inequality:

$$p = P(X \geq \frac{n}{4}) \leq \frac{\mathbb{E}[X]}{\frac{n}{4}} = \frac{4\mathbb{E}[X]}{n}.$$

We need to evaluate $\mathbb{E}[X]$:

$$\mathbb{E}[X] = \mathbb{E}\left[\sum_{i=1}^n X_i\right] = n \sum_{i=1}^n \mathbb{E}[X_i] = \frac{n}{6}.$$

where X_i is the result we get at the i th throw, i.e. $X_i = \begin{cases} 1 & \text{w/ prob } 1/6 \\ 0 & \text{otherwise} \end{cases}$

$$\therefore \boxed{P \leq \frac{4(\frac{n}{6})}{n} = \frac{4}{6} = \frac{2}{3}}.$$

b) Chebyshov's inequality:

$$P(|X - \mathbb{E}[X]| > t) \leq \frac{\text{Var}[X]}{t^2}$$

$$\text{Var}[X] = \text{Var}\left[\sum_i X_i\right] = \sum_{i=1}^n \text{Var}[X_i] = n \cdot \frac{5}{36} = \frac{5n}{36}.$$

$$\begin{aligned} P\left(X \geq \frac{n}{4}\right) &= P\left(X - \frac{n}{6} \geq \frac{n}{4} - \frac{n}{6}\right) \\ &= P\left(X - \mathbb{E}[X] \geq \frac{n}{12}\right) \\ &\leq P\left(|X - \mathbb{E}[X]| \geq \frac{n}{12}\right) \end{aligned}$$

$$\text{Chebyshov's ineq.} \leq \frac{\text{Var}[X]}{(n/12)^2}$$

$$\begin{aligned} \text{with } t = \frac{n}{12} \\ &= \frac{12^2}{n^2} \cdot \frac{5n}{36} \\ &= \frac{20}{n} \end{aligned}$$

$$\therefore \boxed{P \leq \frac{20}{n}}$$

c) Chernoff's inequality:

$$P = P(X \geq \frac{n}{4})$$

$$P(X \geq (1+\varepsilon)\mu) \leq e^{-\mu\varepsilon^2/2}.$$

$$\text{We want } (1+\varepsilon)\mu = \frac{n}{4} \Rightarrow (1+\varepsilon)\frac{n}{6} = \frac{n}{4}$$

$$1+\varepsilon = \frac{3}{2} \Rightarrow \varepsilon = \frac{1}{2}.$$

$$\therefore P = P(X \geq \frac{n}{4}) = P(X \geq (1+\varepsilon)\mu) \leq e^{-\frac{n}{6} \cdot (\frac{1}{2})^2/2} = e^{-n/96}$$

$$\therefore \boxed{P \leq e^{-n/96}}.$$

5) $X_i = \begin{cases} 1 & \text{w/p } 1/2 \\ -1 & \text{w/p } 1/2 \end{cases}$

$$\text{Let } S = \sum_{i=1}^n X_i.$$

a) Since $Y = |S|$, Y is a non-negative random variable, Markov's inequality holds for Y . That is,

for $t > 0$,

$$P(Y \geq t) \leq \frac{\mathbb{E}[Y]}{t}.$$

b) Chebyshov's inequality: for S

$$P(|S - \mathbb{E}[S]| \geq t) \leq \frac{\text{Var}[S]}{t^2}.$$

$$\mathbb{E}[S] = \mathbb{E}\left[\sum_i X_i\right] = \sum_i \mathbb{E}[X_i] = 0.$$

$$\text{Var}[S] = \text{Var}\left[\sum_i X_i\right] = \sum_i \text{Var}[X_i] = n$$

$$\therefore P(|S| \geq t) \leq \frac{n}{t^2}.$$

c) (This is similar to the way we prove Chernoff's bound).

$$\begin{aligned} P(S \geq a) &= P(X_1 + \dots + X_n \geq a) \\ &= P(\lambda \sum_i X_i \geq \lambda a) \quad \text{for } \lambda \geq 0 \\ &= P(e^{\lambda \sum_i X_i} \geq e^{\lambda a}) \quad \text{since } e^x \text{ is monotone.} \end{aligned}$$

$$\text{Markov's} \leftarrow e^{-\lambda a} E[e^{\lambda \sum_i X_i}]$$

Now, let's consider $E[e^{\lambda \sum_i X_i}]$.

$$E[e^{\lambda \sum_i X_i}] = E[e^{\lambda X_1}, e^{\lambda X_2}, \dots, e^{\lambda X_n}]$$

$$= \prod_i E[e^{\lambda X_i}] \quad \text{since } X_i \text{'s are independent.}$$

$$= \prod_i \left(e^\lambda \cdot P(X_i = 1) + \bar{e}^\lambda \cdot P(X_i = -1) \right)$$

$$= \prod_i \frac{e^\lambda + \bar{e}^\lambda}{2}$$

$$= \prod_i \cosh(\lambda)$$

$$= (\cosh(\lambda))^n$$

consider $\lambda \in [0, 1]$, then $\cosh(\lambda) \leq e^{\lambda^2/2}$.

$$(\cosh(\lambda))^n \leq e^{\lambda^2 n / 2}$$

$$\therefore P(S \geq a) \leq \bar{e}^{\lambda a} E[e^{\lambda \sum_i X_i}] = e^{\frac{\lambda^2 n}{2} - \lambda a}$$

Recall that $\min_{\lambda \in [0, 1]} \frac{\lambda^2 n}{2} - \lambda a$ occurs at $\lambda = \frac{a}{n}$. \Leftarrow take this one

$$\therefore P(S \geq a) \leq e^{\frac{a^2 \cdot n}{2} - \frac{a^2}{n}} = e^{-a^2/2n}$$