

(a) **Solution:**

$$\begin{aligned}
 E[X] &= E[X_1 + \dots + X_n] = \sum_{i=1}^n E[X + i] && \text{linearity of expectation} \\
 &= \sum_{i=1}^n \sum_x x p (1-p)^{i-1} && \text{defn. of E} \\
 &= \sum_i i = 1^n \frac{1-p}{p} && \text{expectation for geometric} \\
 &= \frac{n(1-p)}{p}
 \end{aligned}$$

$$\begin{aligned}
 E[X^2] &= E[(X_1 + \dots + X_n)^2] && X = \sum_i X_i \\
 &= E[X_1^2 + X_1 X_1 + \dots + X_n^2] \\
 &= \sum_{i=1}^n E[X_i^2] + \sum_{i,j \in [n], i \neq j} E[X_i X_j] && \text{linearity of expectation} \\
 E[X]^2 &= (E[X_1] + \dots + E[X_n])^2 && \text{linearity of expectation} \\
 &= \sum_i i = 1^n E[X + i]^2 + \sum_{i,j \in [n], i \neq j} E[X_i X_j] && \text{indep. of } X_i X_j \text{ } i \neq j \\
 \text{Var}(X) &= E[X^2] - E[X]^2 && \text{defn. of var in terms of E} \\
 &= \sum_i i = 1^n E[X_i^2] - \sum_{i=1}^n E[X_i]^2 && E[X^2] \text{ and } E[X]^2 \text{ from above} \\
 &= \sum_{i=1}^n (E[X_i^2] - E[X_i]^2) \\
 &= \sum_{i=1}^n \text{Var}(X_i)
 \end{aligned}$$

■

(b) **Solution:** When applying Chebyshev's inequality  $Pr[|(X - E[X])| \geq \epsilon] \leq \frac{\text{Var}(X)}{\epsilon^2}$  to  $Pr[X \geq c E[X]]$ , because  $X$  is always positive (as is  $E[X]$  naturally), there are two cases for the absolute value.  $X - E[X]$  where  $X \geq E[X] \geq 0$  is a positive or zero value, and thus would never deviate from  $|X - E[X]|$ .

$X - E[X]$  where  $0 \leq X < E[X]$  is a negative value  $v$  in the range  $E[X] - v < 0$ . Thus, in order for Chebyshev's to be applied to  $(X - E[X])$ ,  $\epsilon$  must satisfy  $\epsilon \geq -E[X]$ , such that there is no difference of evaluation of  $|X - E[X]| \geq \epsilon$  and  $X - E[X] \geq \epsilon$ . Because  $\epsilon = c E[X]$  satisfies this requirement for  $c \geq 2$ , we get  $Pr[X \geq E[X]] \leq \frac{\text{Var}(X)}{E[X]^2}$  applying Chebyshev's, and using values from the previous part. ■

(c) **Solution:** The event  $X > t$  describes the occurrence in which  $X > t$  flips are required to yield  $n$  flips of heads. As a result, by definition this means that the first  $t$  flips yielded fewer than  $n$  heads, e.g.  $Y < n$ . As a result,  $Pr[X > t] = Pr[Y < n]$ . Considering the case where  $Pr[X = t]$ , there are  $\binom{t-1}{n-1}$  ways to partition the  $t$  flips into  $n$  groups (delineated by a successful flip of heads, using stars and bars). Each of these ways to partition have probability  $p^n (1-p)^{t-n}$  (based on number of successes and failures), therefore  $Pr[X = t] = \binom{t-1}{n-1} p^n (1-p)^{t-n}$ .  $Pr[Y = n]$  is modeled by a binomial distribution,

thus  $\Pr[Y = n] = \binom{t}{n} p^n (1-p)^{t-n}$ . This leads to  $\Pr[X = t] \leq \Pr[Y = n]$ , and thus  $\Pr[X \geq t] \leq \Pr[Y \leq n]$ . ■

- (d) **Solution:** In this application,  $\epsilon = c - 1$ , e.g.  $\epsilon$  is over the bound  $\epsilon \geq 1$ . Applying part 1c i for the first applicable portion of this bound, we can say that  $\Pr[X \geq c \mathbb{E}[X]] \leq e^{-(c-1)^2 \mathbb{E}[X]/4}$  for  $2 \leq c \leq 2 + \ln 4$ . Similarly for second portion of the bound of  $c$ , we can apply part 1c ii and say that  $\Pr[X \geq c \mathbb{E}[X]] \leq 2^{-(c-1) \mathbb{E}[X]/2}$  for  $c \geq \ln 4$ , and thus we have bound  $\Pr[X \geq c \mathbb{E}[X]]$  for  $c \geq 2$ . ■