- Let X_i be the result of the *i*-th roll.
 - All X_i are mutually independent, as the result of the dice roll do not affect one another.

•
$$\mathbf{E}[X_i] = \sum_{k=1}^{6} k \times \frac{1}{6} = \frac{7}{2}$$

$$\label{eq:Var} {\bf \hat{V}ar}[X_i] = \sum_{k=1}^6 \left(k - \mathbf{E}[X_i]\right)^2 \times \frac{1}{6} = \frac{35}{12}$$

•
$$X = \sum_{i=1}^{100} X_i$$

•
$$\mathbf{E}[X] = \sum_{i=1}^{100} \mathbf{E}[X_i] = 350$$

•
$$Var[X] = \sum_{i=1}^{100} Var[X_i] = \frac{875}{3}$$

•
$$\Pr(|X - 350| \ge 50) \le \frac{\mathbf{Var}[X]}{50^2} = \frac{7}{60}$$
 (Chevyshev's inequality)

3.6

- Let *X* be the number of flips until the k-th head appears
- Let X_i be the number of flips after the (i-1)-th head until the i-th head appears (Excluding the (i-1)-th flips and including the i-th flip)
 - All X_i are mutually independent, as the number of flips until one head does not affect the number of flip until another head.

•
$$\mathbf{Var}[X_i] = \frac{1-p}{p^2}$$

$$X = \sum_{i=1}^{k} X_i$$

$$\ \ \mathbf{Var}[X] = \sum_{i=1}^k \mathbf{Var}[X_i] = \frac{k(1-p)}{p^2}$$

•
$$\mathbf{Var}[X] = \frac{k(1-p)}{p^2}$$

- Let X_d be the price after d days

The expected value of X_d

$$^{\bullet} \mathbf{E}[X_d] = p \times r \mathbf{E}[X_{d-1}] + (1-p) \times \frac{1}{r} \mathbf{E}[X_{d-1}] = \frac{pr^2 + 1 - p}{r} \mathbf{E}[X_{d-1}]$$

•
$$\mathbf{E}[X_0] = 1$$

•
$$\mathbf{E}[X_d] = \left(\frac{pr^2 + 1 - p}{r}\right)^d$$

The variance of X_d

$$^{\bullet} \mathbf{E}[{X_d}^2] = p \times r^2 \mathbf{E}[{X_{d-1}}^2] + (1-p) \times \left(\frac{1}{r}\right)^2 \mathbf{E}[{X_{d-1}}^2] = \frac{pr^4 + 1 - p}{r^2} \mathbf{E}[{X_{d-1}}^2]$$

•
$$\mathbf{E}[X_0^2] = 1$$

$$^{\bullet} \ \mathbf{E}[{X_d}^2] = \left(\frac{pr^4 + 1 - p}{r^2}\right)^d$$

$$\bullet \ \mathbf{Var}[X_d] = \mathbf{E}\big[{X_d}^2\big] - \big(\mathbf{E}[X_d]\big)^2 = \tfrac{(pr^4 + 1 - p)^d - (pr^2 + 1 - p)^{2d}}{r^{2d}}$$

3.15

• Proposition: for two random variables X and Y, if $\mathbf{E}[XY] = \mathbf{E}[X]\mathbf{E}[Y]$, then $\mathbf{Var}[X+Y] = \mathbf{Var}[X] + \mathbf{Var}[Y]$

•
$$\mathbf{Var}[X+Y] = \mathbf{E}[(X+Y)^2] - (\mathbf{E}[X+Y])^2$$

 $= \mathbf{E}[X^2 + 2XY + Y^2] - (\mathbf{E}[X] + \mathbf{E}[Y])^2$
 $= \mathbf{E}[X^2] + 2\mathbf{E}[XY] + \mathbf{E}[Y^2] - ((\mathbf{E}[X])^2 + 2\mathbf{E}[X]\mathbf{E}[Y] + (\mathbf{E}[Y])^2)$
 $= \mathbf{E}[X^2] - (\mathbf{E}[X])^2 + \mathbf{E}[Y^2] - (\mathbf{E}[Y])^2 + 2\mathbf{E}[XY] - 2\mathbf{E}[X]\mathbf{E}[Y]$
 $= \mathbf{E}[X^2] - (\mathbf{E}[X])^2 + \mathbf{E}[Y^2] - (\mathbf{E}[Y])^2$
 $= \mathbf{Var}[X] + \mathbf{Var}[Y]$

- For n random variables X_i that satisfies $\mathbf{E}\big[X_iX_j\big] = \mathbf{E}[X_i]\mathbf{E}\big[X_j\big]$ with $1 \leq i < j \leq n$
 - For all k such that $1 \le k \le n$,

$$\begin{split} \mathbf{E}\left[\left(\sum_{i=1}^{k}X_{i}\right)X_{k+1}\right] &= \mathbf{E}\left[\sum_{i=1}^{k}X_{i}X_{k}\right] \\ &= \sum_{i=1}^{k}\mathbf{E}[X_{i}]\mathbf{E}[X_{k}] \\ &= \left(\sum_{i=1}^{k}\mathbf{E}[X_{i}]\right)\mathbf{E}[X_{k}] \\ &= \mathbf{E}\left[\left(\sum_{i=1}^{k}X_{i}\right)\right]\mathbf{E}[X_{k}] \\ &= \mathbf{Var}\left[\left(\sum_{i=1}^{n-1}X_{i}\right) + X_{n}\right] = \mathbf{Var}\left[\left(\sum_{i=1}^{n-1}X_{i}\right)\right] + \mathbf{Var}[X_{n}] \\ &= \mathbf{Var}\left[\left(\sum_{i=1}^{n-2}X_{i}\right) + X_{n-1}\right] + \mathbf{Var}[X_{n}] = \mathbf{Var}\left[\left(\sum_{i=1}^{n-2}X_{i}\right)\right] + \mathbf{Var}[X_{n-1}] + \mathbf{Var}[X_{n}] \\ &\dots \\ &= \mathbf{Var}[X_{1} + X_{2}] + \sum_{i=3}^{n}\mathbf{Var}[X_{i}] = \mathbf{Var}[X_{1}] + \mathbf{Var}[X_{2}] + \sum_{i=3}^{n}\mathbf{Var}[X_{i}] \\ &= \sum_{i=1}^{n}\mathbf{Var}[X_{i}] \end{split}$$

Upper bound: $Pr[Y \neq 0] \leq E[Y]$

•
$$\Pr[Y \neq 0] = \Pr[Y \geq 1]$$
 (Y has nonnegative integer-value)
 $\leq \frac{E[Y]}{1} = E[Y]$ (Markov's Inequality)

Lower bound: $\Pr[Y \neq 0] \ge \frac{(\mathbf{E}[Y])^2}{\mathbf{E}[Y^2]}$

- Let Z be a positive-valued integer random variable
 - That satisfies $\Pr(Z=k) = \frac{\Pr(Y=k)}{\Pr(Y\neq 0)}$ for all positive integer k

$$\mathbf{E}[Z] = \sum_{k=1}^{n} k \cdot \frac{\Pr(Y = k)}{\Pr(Y \neq 0)}$$

$$= \frac{1}{\Pr(Y \neq 0)} \sum_{k=1}^{n} k \cdot \Pr(Y = k)$$

$$= \frac{1}{\Pr(Y \neq 0)} \sum_{k=0}^{n} k \cdot \Pr(Y = k)$$

$$= \frac{1}{\Pr(Y \neq 0)} \times \mathbf{E}[Y]$$

$$\begin{split} \bullet & \mathbf{E}[Z^2] = \sum_{k=1}^n k^2 \cdot \frac{\Pr(Y=k)}{\Pr(Y\neq 0)} \\ &= \frac{1}{\Pr(Y\neq 0)} \sum_{k=1}^n k^2 \cdot \Pr(Y=k) \\ &= \frac{1}{\Pr(Y\neq 0)} \sum_{k=0}^n k^2 \cdot \Pr(Y=k) \\ &= \frac{1}{\Pr(Y\neq 0)} \times \mathbf{E}[Y^2] \end{split}$$

•
$$\mathbf{Var}[Z] \ge 0 \Leftrightarrow \mathbf{E}[Z^2] \ge (\mathbf{E}[Z])^2$$

$$\Leftrightarrow \frac{1}{\Pr(Y \ne 0)} \times \mathbf{E}[Y^2] \ge \left(\frac{1}{\Pr(Y \ne 0)} \times \mathbf{E}[Y]\right)^2$$

$$\Leftrightarrow \Pr(Y \ne 0) \ge \frac{(\mathbf{E}[Y])^2}{\mathbf{E}[Y^2]}$$

• Let
$$X = \frac{1}{n} \sum_{i=1}^{n} X_i$$

•
$$\mathbf{E}[X] = \frac{1}{n} \sum_{i=1}^{n} \mathbf{E}[X_i] = \frac{n\mu}{n} = \mu$$

$${\bf ^{\bullet}} \ {\bf Var}[X] = \frac{1}{n^2} \sum_{i=1} {\bf E}[X_i] = \frac{n\sigma^2}{n^2} = \frac{\sigma^2}{n}$$

•
$$\Pr(|X - \mu| > \varepsilon) \leq \frac{\mathbf{Var}(X)}{\varepsilon^2} = \frac{\sigma^2}{n\varepsilon^2} \text{ (Chebyshev's inequality)}$$

$$\lim_{n \to \infty} \Pr(|X - \mu| > \varepsilon) \le \lim_{n \to \infty} \frac{\sigma^2}{n\varepsilon^2} = 0$$