

CSE 107 1-29-26

11

Variance of Binomial(n, p)

If X is Binomial(n, p), then

$$X = X_1 + X_2 + \dots + X_n$$

where $X_i = \text{Bernoulli}(p)$, are independent. Thus

$$\begin{aligned}\text{Var}(X) &= \text{Var}(X_1 + \dots + X_n) \\ &= \text{Var}(X_1) + \dots + \text{Var}(X_n)\end{aligned}$$

$$= \underbrace{p(1-p) + p(1-p) + \dots + p(1-p)}_n$$

$$= np(1-p)$$



Ex. Suppose we have a weighted coin with $P(\text{head}) = p$, but we don't know p .

we compute sample mean

$$\bar{X}_n = \frac{X_1 + X_2 + \dots + X_n}{n}$$

for each $n \geq 1$. (also relative frequency)

note

$$E[X_i] = p$$

so

$$E[S_n] = E\left[\frac{1}{n}X_1 + \frac{1}{n}X_2 + \dots + \frac{1}{n}X_n\right]$$

$$= \sum_{i=1}^n \frac{1}{n} E[X_i] = \sum_{i=1}^n \frac{1}{n} \cdot p$$

$$= \frac{1}{n} \sum_{i=1}^n p = \frac{1}{\cancel{n}} \cdot \cancel{n} p = p$$

so S_n is an estimate of $E[S_n]$

$= p$, how good is the estimate?

note

$$\text{Var}(\bar{X}_n) = \text{Var}\left(\sum_{i=1}^n \frac{1}{n} X_i\right)$$

$$= \sum_{i=1}^n \text{Var}\left(\frac{1}{n} X_i\right)$$

$$= \sum_{i=1}^n \frac{1}{n^2} \text{Var}(X_i)$$

$$= \frac{1}{n^2} \sum_{i=1}^n p(1-p) = \frac{1}{n^2} \cdot n p(1-p)$$

$$= \frac{p(1-p)}{n} \longrightarrow 0$$

Exercise

If $\text{Var}(Y) = 0$, then $Y = \text{const.}$

note

even if X_i are not Bernoulli,

same calculation shows

$$\text{Var}(\Sigma_n) = \frac{\text{Var}(X_i)}{n}$$

$\rightarrow 0$ as $n \rightarrow \infty$.