

e.g.: Roll fair dice 600 times independently,
 X be the # of sixes that appear.

(a): What type of r.v. is X ? Dist of X ?

Support of X : $\{0, 1, 2, \dots, 600\}$



X is a discrete r.v.

Repeated indep experiment \Rightarrow binomial dist

$$\underbrace{X \sim B(600, \frac{1}{6})}$$

(b): Find $IP(X \leq 100)$

$$IP(X \leq 100) = \sum_{k=0}^{100} \underbrace{IP(X=k)}_{\text{pmf of binomial}} = \sum_{k=0}^{100} \binom{600}{k} \left(\frac{1}{6}\right)^k \left(\frac{5}{6}\right)^{600-k}$$

(c): Use a limit theorem to show why the prob in (b) can be approximated by value $\frac{1}{2}$.
(LLN, CLT)

We shall use CLT.

SLLN: If X_1, \dots, X_n, \dots i.i.d. r.v., $|EX_i| < \infty$,
 $\frac{X_1 + \dots + X_n}{n} \xrightarrow{\text{a.s.}} EX_i \ (n \rightarrow \infty)$

Since $X \sim B(600, \frac{1}{6})$, let X_i be an indicator of whether the i -th roll gives six.

$X = X_1 + \dots + X_{600}$ (X is i.i.d. sum of
large enough $n=600$ r.v.)

$$\left\{ \begin{array}{l} EX = 600 \times \frac{1}{6} = 100 \\ \text{Var}(X) = 600 \times \frac{1}{6} \times \frac{5}{6} = \frac{500}{6} \end{array} \right.$$

Apply CLT:

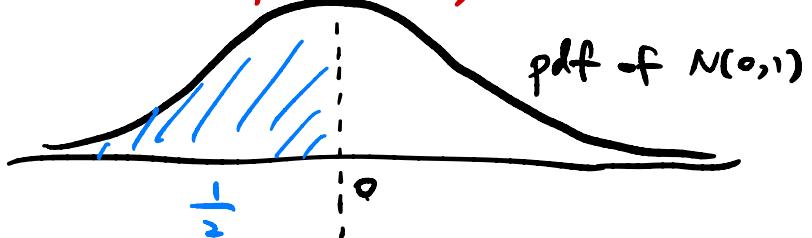
$$\frac{X - EX}{\sqrt{\text{Var}(X)}} \underset{\text{approx}}{\sim} N(0, 1)$$

$$\frac{X - 100}{\sqrt{\frac{500}{6}}} \underset{\text{approx}}{\sim} N(0, 1)$$

$$\text{IP}(X \leq 100) = \text{IP}\left(\frac{X - 100}{\sqrt{\frac{500}{6}}} \leq \frac{100 - 100}{\sqrt{\frac{500}{6}}}\right)$$

$$= \text{IP}\left(\frac{X - 100}{\sqrt{\frac{500}{6}}} \leq 0\right) \approx \left(\frac{1}{2}\right)$$

CLT: approx $N(0, 1)$



We have a dist F , we have r.v. $X \sim F$,

want to estimate $\mathbb{E} h(X) \Rightarrow$ Monte Carlo
 \hat{f}

① Sampling,

sample n i.i.d. random variable
 $X_1, \dots, X_n \sim F$.

② Average

$$\hat{f} = \frac{h(X_1) + h(X_2) + \dots + h(X_n)}{n}$$

(sample mean)

\hat{f} is an estimator for f .

By SLLN, $\hat{f} \xrightarrow{\text{a.s.}} \mathbb{E} h(X) = f \quad (n \rightarrow \infty)$

so Monte Carlo converges to the true f .

Special Cases: when I want to estimate

$$\mathbb{P}(A) = \mathbb{E} I_A$$

Monte Carlo (100 simulations),

we are sampling from $B(600, \frac{1}{6})$ for 100 time, so we have ① Sampling

$$Y_1, Y_2, \dots, Y_{100} \text{ i.i.d. } \sim B(600, \frac{1}{6})$$

but now want to estimate

$$\underbrace{\Pr(X < 90)}_{\text{II}} = p$$

so consider

$$\begin{cases} Z_1 = I_{\{Y_1 < 90\}} \\ Z_2 = I_{\{Y_2 < 90\}} \\ \vdots \\ Z_{100} = I_{\{Y_{100} < 90\}} \end{cases} \rightarrow \text{i.i.d.}$$

Applying transformation

estimator is

$$\frac{Z_1 + Z_2 + \dots + Z_{100}}{100} = \hat{P} \xrightarrow{\text{blue line in notebook}} \text{②: Average}$$

Variance of \hat{P} :

$$\frac{\hat{P} - \mathbb{E}\hat{P}}{\sqrt{\text{Var } \hat{P}}} \xrightarrow{\text{approx}} N(0, 1)$$

$$\begin{aligned} \mathbb{E} \hat{P} &= \mathbb{E} Z_1 = \mathbb{E} I_{\{Y_1 < 90\}} \\ &= \Pr(Y_1 < 90) = p \end{aligned}$$

\hat{P} is unbiased

$$\text{Var } \hat{P} = \frac{100 \cdot \text{Var } Z_1}{100^2} = \frac{\text{Var } Z_1}{100} = \boxed{\frac{P(1-P)}{100}}$$

$$Z_1 = I_{\{Y_1 < 90\}} \sim B(1, p)$$

appears in CI.

$$\underline{\text{Var } Z_1 = P(1-P)}$$

(General Monte Carlo)

e.g.: We want to figure out expectation of $B(600, \frac{1}{6})$, called f .

Sample 100 obs $Y_1, \dots, Y_{100} \sim B(600, \frac{1}{6})$

then estimator is $\frac{Y_1 + \dots + Y_{100}}{100} = \hat{f}$. i.i.d.

Build 95% CI for p:

$$\frac{\hat{P} - P}{\sqrt{\frac{P(1-P)}{100}}} \stackrel{\text{approx}}{\sim} N(0, 1) \quad (\text{result of CLT})$$

find a, b such that

$$P\left(a \leq \frac{\hat{P} - P}{\sqrt{\frac{P(1-P)}{100}}} \leq b\right) = 0.95$$



$$P\left(a \sqrt{\frac{P(1-P)}{100}} \leq \hat{P} - P \leq b \cdot \sqrt{\frac{P(1-P)}{100}}\right) = 0.95$$



$$P\left(\hat{P} - b \cdot \sqrt{\frac{P(1-P)}{100}} \leq P \leq \hat{P} - a \cdot \sqrt{\frac{P(1-P)}{100}}\right) = 0.95$$

so: CI is $\left[\hat{P} - b \cdot \sqrt{\frac{P(1-P)}{100}}, \hat{P} - a \cdot \sqrt{\frac{P(1-P)}{100}}\right]$

