

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 : $\{0, 1, 2, \dots, 600\}$

$\Downarrow$

X is a discrete r.v.

Repeated indep experiment $\Rightarrow$ binomial dist

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

(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 thm (LLN, CLT) to show why the prob in (b) can be approximated by value $\frac{1}{2}$.

We shall use CLT.

SLLN: If $X_1, \dots, X_n, \dots$ i.i.d. r.v., $E|X_i| < \infty$,
$$\frac{X_1 + \dots + X_n}{n} \xrightarrow{\text{a.s.}} EX_1 \quad (n \rightarrow \infty)$$

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

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

$$\begin{cases} EX = 600 \times \frac{1}{6} = 100 \\ \text{Var}(X) = 600 \times \frac{1}{6} \times \frac{5}{6} = \frac{500}{6} \end{cases}$$

Apply CLT:

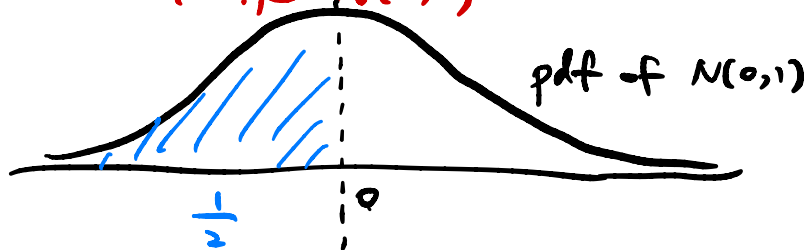
$$\frac{X - EX}{\sqrt{\text{Var}(X)}} \overset{\text{approx}}{\sim} N(0, 1)$$

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

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

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

CLT: approx $N(0,1)$



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

want to estimate $IE h(X) \Rightarrow$ Monte Carlo
 $\parallel$
 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, $\hat{f} \xrightarrow{\text{a.s.}} IE h(X) = f \quad (n \rightarrow \infty)$

so Monte Carlo converges to the true f .

Special Cases: when I want to estimate

$$IP(A) = IE I_A$$

Monte Carlo (100 simulations),

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

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

but now want to estimate $\underbrace{IP(X < 90)}_{\substack{= \\ IE I_{\{X < 90\}}}} = p$
so consider

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

Applying transformation

②: Average

estimator is $\underbrace{\frac{Z_1 + Z_2 + \dots + Z_{100}}{100}}_{\text{i.i.d. sum, has CLT}} = \hat{p} \Rightarrow \text{blue line in notebook}$

variance of $\hat{p}$:

$$\frac{\hat{p} - IE \hat{p}}{\sqrt{\text{Var } \hat{p}}} \overset{\text{approx}}{\sim} N(0, 1)$$

$$\begin{aligned} IE \hat{p} &= IE Z_1 = IE I_{\{Y_1 < 90\}} \\ &= IP(Y_1 < 90) = p \end{aligned}$$

$\hat{p}$ is unbiased

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

appears in CI.

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

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

(General Monte Carlo)

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

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

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

Build 95% CI for p:

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

find a, b such that

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

$\Downarrow$

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

$\Downarrow$

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

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

