AI1103-Assignment 3

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Download all python codes from

https://github.com/DineshAvulaMohanaDurga/ AI1103/blob/main/assignment_3/codes/ ai1103_assignment3.py

and latex codes from

https://github.com/DineshAvulaMohanaDurga/ AI1103/blob/main/assignment_3/main.tex

1 Question

(GATE - 2021 (ME, set - 1) problem - 5)

Consider a binomial random variable X. If $X_1, X_2, ..., X_n$ are independent and identically distributed samples from the distribution of **X** with sum $Y = \sum_{i=1}^{n} X_i$, then the distribution of **Y** as $n \to \infty$ can be approximated as

- 1) Exponential
- 2) Bernoulli
- 3) Binomial
- 4) Normal

2 Answer

Given a binomial random variable X

$$\Rightarrow X \sim B(r, p)$$
 (2.0.1)

also given that $X_1, X_2, ..., X_n$ are independent and identically distributed samples

$$\Rightarrow X_1 = X_2 = \dots = X_n = X \sim B(r, p)$$
 (2.0.2)

also given that

$$Y = \sum_{i=1}^{n} X_i \tag{2.0.3}$$

We know that the characteristic equation of binomial trials with n elements is

$$\phi_X(t) = (1 - p + pe^{it})^n$$
 (2.0.4)

consider two sets of Bernoulli trials containing $r_1\&r_2$ elements respectively where both trials have

the same probability 'p'($X \sim B(r_1, p), Y \sim (r_2, p)$). Now considering both as a whole set

$$B(r_{1}, p) + B(r_{2}, p) = \Phi_{X+Y}(t)$$

$$= \Phi_{X}(t) \times \Phi_{Y}(t)$$

$$= (1 - p + pe^{it})^{r_{1}}$$

$$\times (1 - p + pe^{it})^{r_{2}}$$
(2.0.6)

$$= (1 - p + pe^{it})^{r_1 + r_2}$$
 (2.0.6)
= $B(r_1 + r_2, p)$

$$\therefore B(r_1, p) + B(r_2, p) = B(r_1 + r_2, p)$$
 (2.0.8)

using this recursively we get

$$Y = B(rn, p) \tag{2.0.9}$$

⇒ using standard formulae

mean of Y
$$\mu_Y = nrp$$

and variance $\sigma_Y^2 = nrp(1-p)$ (2.0.10)

By central limit theorem(CLT)

$$Z_{n} = \sqrt{n} \left(\frac{\frac{Y}{n} - \mu_{Y}}{\sigma_{Y}} \right)$$

$$= \frac{Y - n\mu_{Y}}{\sqrt{n}\sigma_{Y}}$$
(2.0.11)

 $\lim_{n\to\infty}Z_n\sim N(0,1)$

Which is a normal distribution
∴ the correct answer is option D

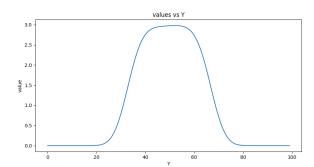


Fig. 4: distribution of Y