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Ques ① → X, Y & Z are independent geometric R.V.

Now → taking a random instance →
suppose → $X=x, Y=y, Z=k$.

So →
$$P(X+Y \leq Z) = \sum_{x,y,k} P(x+y \leq k)$$

↪ this is behaving like compositionally joint prob. distribution of R.V. $(X+Y)$.

$$P(X+Y \leq Z) = \sum_{\forall k} P(X+Y \leq k)$$

So → we can say that →

$$P(X+Y \leq Z) = \frac{1 + P(X+Y=Z)}{2} = \frac{1 + \sum_{\forall k} P(X+Y=k)}{2}$$

Now → $P(X+Y=Z) \Rightarrow \sum_{\forall k} P(X+Y=k)$

⇒ this can be written as →

$$\Rightarrow \sum_{k=1}^{x+y} P_X(X=i) \cdot P_Y(Y=k-i) \quad ; \quad \left\{ \begin{array}{l} \text{Since } X \text{ \& } Y \\ \text{are independent} \end{array} \right.$$

$$\Rightarrow \sum_{k=1}^{x+y} \left\{ (1-p)^{i-1} \cdot p \right\} \cdot \left\{ (1-p)^{k-i-1} \cdot p \right\}$$

$$\Rightarrow \sum_{k=1}^{x+y} \left\{ p^2 \cdot (1-p)^{k-2} \right\}$$

in general ^{Q8} ⇒
$$P(X+Y=Z) = p^2 \cdot (1-p)^{x+y-2}$$

$$\text{So} \rightarrow \boxed{P(X+Y \leq Z) = \frac{1 + p^2 \cdot (1-p)^{x+y-2}}{2}}$$

(12). $X = \text{R.V. denoting no. of black balls in sample.}$

$$\text{So} \rightarrow \boxed{X = x_1 + x_2 + x_3 + \dots + x_r}$$

where x_i are R.V., such that \rightarrow

$$x_i = \begin{cases} 1 & ; \text{if } i\text{th ball taken is black.} \\ 0 & ; \text{otherwise.} \end{cases}$$

$$\text{So} \rightarrow \text{for every } x_i \rightarrow P(x_i = 1) = \left(\frac{b}{n}\right) = (P)$$

$$\text{where } \boxed{n = (b+g)}$$

$$P(x_i = 0) = \left(\frac{g}{n}\right) = (1-P)$$

$$\text{So} \rightarrow E[x_i] = 1 \cdot P + 0 \cdot (1-P) = P \Rightarrow \boxed{E[x_i] = P}$$

Now \rightarrow we know that:

$$E[X] = E\left[\sum_i x_i\right] = \sum_i E[x_i]$$

So \rightarrow (Since every x_i , distribution is identical)

$$\boxed{E[X] = r \cdot P = \left(\frac{r \cdot b}{b+g}\right)}$$

Now \rightarrow

$$\text{Var}[X] = \text{Var}\left[\sum_{i=1}^r x_i\right]$$

$$\Rightarrow \sum_{i=1}^r \text{Var}[x_i] + 2 \sum_{1 \leq i < j \leq r} \text{Cov}[x_i, x_j]$$

$$\Rightarrow \sum_{i=1}^r \left\{ E[x_i^2] - (E[x_i])^2 \right\} + 2 \sum_{1 \leq i, j \leq r} \text{Cov.}[x_i, x_j] \quad \text{--- ①}$$

(easy)

$$\text{Cov.}(x_i, x_j) = E[x_i x_j] - E[x_i] \cdot E[x_j]$$

$$\Rightarrow \sum x_i x_j \cdot P(x_i x_j) - E[x_i] \cdot E[x_j]$$

$$\Rightarrow \frac{b \cdot (b-1) / 2}{n(n-1) / 2} - \left(\frac{b}{n} \right)^2$$

$$\Rightarrow \frac{b(b-1)}{n(n-1)} - \left(\frac{b}{n} \right)^2$$

$$\boxed{\text{Cov.}(x_i, x_j) \Rightarrow \frac{b(b-n)}{n^2(n-1)}}$$

putting this in eq. ① \rightarrow

$$\Rightarrow \sum_{i=1}^r \left\{ E[x_i^2] - (E[x_i])^2 \right\} + 2 \cdot \frac{r(r-1)}{2} \cdot \frac{b(b-n)}{n^2(n-1)}$$

$$\Rightarrow \frac{br(n-b)}{n^2} + r(r-1) \cdot \frac{b(b-n)}{n^2(n-1)}$$

$$\boxed{\text{Var}[x] \Rightarrow \frac{rbg}{n^2} \cdot \left(1 - \frac{(r-1)}{(n-1)} \right)}$$

③ $\rightarrow n_R = 3$ red balls.

$n_B = 2$ black balls.

sample $(k) = 2$.

Let $R.V. \rightarrow$
 $U \rightarrow$ No. of red balls selected.
 $V \rightarrow$ " " black " " "

So \rightarrow

$$\rho(U, V) = \frac{\text{Cov}(U, V)}{\sqrt{\text{Var}(U) \cdot \text{Var}(V)}}$$

$$\text{Cov}(U, V) = E(UV) - E(U) \cdot E(V)$$

probability fn calculations \rightarrow making combinations of red & black balls calculating prob.

$$P(0, 2) \Rightarrow \frac{\binom{3}{0} \binom{2}{2}}{\binom{5}{2}} = \left(\frac{1}{10}\right)$$

$$P(1, 1) \Rightarrow \frac{\binom{3}{1} \binom{2}{1}}{\binom{5}{2}} = \left(\frac{6}{10}\right)$$

$$P(2, 0) \Rightarrow \frac{\binom{3}{2} \binom{2}{0}}{\binom{5}{2}} = \left(\frac{3}{10}\right)$$

$$E(UV) = \sum_{x,y} xy \cdot P(x,y)$$

$$\Rightarrow 0 + 1 \times 1 \times \frac{6}{10} + 0$$

$$\boxed{E(UV) = \frac{3}{5}}$$

Now, since U & V are hypergeometric distributions. So,
Directly putting expectations \rightarrow

$$E[U] = k \cdot \frac{n_R}{n_R + n_b} \Rightarrow \left(\frac{6}{5}\right)$$

$$\underline{f} \rightarrow E[V] = \frac{k \cdot n_b}{n_b + n_R} \Rightarrow \left(\frac{4}{5}\right)$$

$$\text{COV.}(X, Y) = \frac{3}{5} - \frac{24}{25} \Rightarrow \left(-\frac{9}{25}\right)$$

Now \rightarrow Since we already know that \rightarrow

$$\text{Var}(X) = k \cdot P \cdot (1-P) \cdot \left(1 - \frac{k-1}{n-1}\right)$$

if X is hypergeometric

$$\text{Var}(U) = 2 \times \frac{3}{5} \times \frac{2}{5} \times \left(1 - \frac{2-1}{5-1}\right)$$

$$\Rightarrow 2 \times \frac{3}{5} \times \frac{2}{5} \times \frac{3}{4} \Rightarrow \left(\frac{36}{25}\right)$$

$$\text{Var}(V) \Rightarrow 2 \times \frac{2}{5} \times \frac{3}{5} \times \left(1 - \frac{2-1}{5-1}\right) \Rightarrow \left(\frac{36}{25}\right)$$

Thus \rightarrow

$$\rho(U, V) = \frac{-\frac{9}{25}}{\sqrt{\frac{36}{25} \cdot \frac{36}{25}}} \Rightarrow \frac{-\frac{9}{25} \times \frac{25}{36 \times 4}}{\Rightarrow \left(-\frac{1}{4}\right)}$$

④ → R.V. → $x_1, x_2, x_3, \dots, x_n$

Since → M.G.f. →

$$M(t) = E[e^{tx}] = \sum_{x_n} e^{tx} \cdot p(x)$$

& Joint M.G.f is defined as →

$$M_{x+y}(t) = M_x(t) \cdot M_y(t)$$

Now →

$$M.G.f(x_1 + x_2 + \dots + x_n) =$$

$$M_{x_1}(t) * M_{x_2}(t) * \dots * M_{x_n}(t)$$

So →

$$M_x(t) = [M_{x_1}(t)]^n$$

But →

$$M_{x_1}(t) = \sum_x e^{tx} \cdot p(x)$$

Now →

L.H.S →

$$= E[(x_1 + \dots + x_n)^3]$$

⇒ $E[\gamma^3]$ → it is 3rd moment.