

$$\xi \in \{-1; 1\}, \eta \in \{-1; 0; 1\} \Rightarrow \xi + \eta \in \{-2; -1; 0; 1; 2\}, \xi\eta \in \{-1; 0; 1\}$$

$$\xi + \eta \sim \begin{pmatrix} -2 & -1 & 0 & 1 & 2 \\ \frac{1}{8} & \frac{1}{12} & \frac{7}{24} + \frac{1}{3} & \frac{1}{6} & 0 \end{pmatrix}, \quad \xi\eta \sim \begin{pmatrix} -1 & 0 & 1 \\ \frac{7}{24} + \frac{1}{3} & \frac{1}{12} + \frac{1}{6} & \frac{1}{8} + 0 \end{pmatrix}$$

$$\begin{array}{c|ccccc} \xi\eta & -2 & -1 & 0 & 1 & 2 \\ \hline -1 & 0 & 0 & \frac{1}{3} + \frac{7}{24} & 0 & 0 \\ 0 & 0 & \frac{1}{12} & 0 & \frac{1}{6} & 0 \\ 1 & \frac{1}{8} & 0 & 0 & 0 & 0 \end{array}$$

where each cell is  $\sum_{k,l} P(\xi=k, \eta=l)$   
and  $kl, k+l$  define the cell

The joint distribution of  $\xi\eta, \xi+\eta$  is the distribution of  $\begin{bmatrix} \xi\eta \\ \xi+\eta \end{bmatrix}$ .

N11. Let  $p=0,8$  be the probability of the marksman hitting the target.  
Let  $\delta$  be the amount of shots until the marksman hits and misses at least once.  
Let  $\varphi$  be the indicator random variable that shows if the first shot hit the target.  
Let  $H \sim \text{Geo}(p)$ ,  $M \sim \text{Geo}(1-p)$  be the amount of misses and hits before a hit and a miss respectively ( $\text{Geo}(p)$  is a geometrical distribution)

$$\begin{aligned} E(\delta|\varphi=1) &= 1 + EM \\ E(\delta|\varphi=0) &= 1 + EH \end{aligned} \Rightarrow E(\delta|\varphi) \sim \begin{pmatrix} 1+EM & 1+EH \\ p & 1-p \end{pmatrix}$$

$$EM = \frac{1}{1-p} = 5$$

$$EH = \frac{1}{p} = \frac{5}{4}$$

$$E\delta = E(E(\delta|\varphi)) = p(1+5) + (1-p)(1+\frac{5}{4}) = 5,25$$

$$\text{N12. } \xi \sim \begin{pmatrix} -1 & 0 & 1 \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{pmatrix}, \quad \xi = 1 - \xi^{1000}, \quad \eta = 1 - \xi^{1001}$$

$$\xi \sim \begin{pmatrix} 0 & 1 \\ \frac{1}{3} + \frac{1}{3} & \frac{1}{3} \end{pmatrix}, \quad \eta \sim \begin{pmatrix} 0 & 1 & 2 \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{pmatrix}$$

$\xi \backslash \eta$	0	1	2
0	$\frac{1}{3}$	0	$\frac{1}{3}$
1	0	$\frac{1}{3}$	0

$$P(\xi=0, \eta=0) = \frac{1}{3} \text{ (from the table, } \xi=1)$$

$$P(\xi=0) \cdot P(\eta=0) = \frac{2}{3} \cdot \frac{1}{3} = \frac{2}{9} \neq \frac{1}{3} \text{ (from the marginal distributions)}$$

Therefore,  $\xi$  and  $\eta$  are not independent.

$$\xi\eta \sim \begin{pmatrix} 0 & 1 & 2 \\ \frac{2}{3}(\frac{1}{3} + \frac{1}{3} + \frac{1}{3}) + \frac{1}{3} \cdot \frac{1}{3} & \frac{1}{3} \cdot \frac{1}{3} & \frac{1}{3} \cdot \frac{1}{3} \end{pmatrix} = \begin{pmatrix} 0 & 1 & 2 \\ \frac{7}{9} & \frac{1}{9} & \frac{1}{9} \end{pmatrix}$$

$$\text{Cov}(\xi, \eta) = E(\xi\eta) - E\xi \cdot E\eta = 1 \cdot \frac{1}{9} + 2 \cdot \frac{1}{9} - (\frac{1}{3} \cdot 1)(\frac{1}{3} \cdot 1 + \frac{1}{3} \cdot 2) = 0$$

$$\text{Var } \xi = E\xi^2 - (E\xi)^2 = \frac{2}{9}, \quad \text{Var } \eta = E\eta^2 - (E\eta)^2 = \frac{2}{3}, \quad \rho_{\xi, \eta} = \frac{\text{Cov}(\xi, \eta)}{\sqrt{\text{Var } \xi \cdot \text{Var } \eta}} = 0$$

N13. Let  $\xi \sim \text{Geo}(p)$ ,  $\eta \sim \text{Geo}(p)$ , where  $\text{Geo}(p)$  is a geometric distribution.

$$P(\xi=k | \xi+\eta=n) = \frac{P(\xi=k, \xi+\eta=n)}{P(\xi+\eta=n)} = \frac{P(\xi=k, \eta=n-k)}{P(\xi+\eta=n)}, \text{ since } AB = A(AB) \text{ for events } A, B$$

$$P(\xi=k, \eta=n-k) \stackrel{\text{independent}}{=} P(\xi=k) \cdot P(\eta=n-k) = p \cdot (1-p)^{k-1} \cdot p \cdot (1-p)^{n-k-1} = p^2 (1-p)^{n-2} \text{ for } k \in [1; n-1], \text{ since } \xi, \eta > 0$$

$$P(\xi+\eta=n) = \sum_{i=1}^{n-1} P(\xi=i, \eta=n-i) = \sum_{i=1}^{n-1} p^2 (1-p)^{n-2} = (n-1)p^2 (1-p)^{n-2}$$

$$\text{Thus, } P(\xi=k | \xi+\eta=n) = \frac{p^2 (1-p)^{n-2}}{(n-1)p^2 (1-p)^{n-2}} = \frac{1}{n-1} \quad \square$$



N14. Let  $\xi_1 \sim \text{Geo}(p_1)$ ,  $\xi_2 \sim \text{Geo}(p_2)$ , where  $\text{Geo}(p)$  is a geometric distribution.

$$\text{Let } \xi = \min\{\xi_1, \xi_2\}$$

$$F_\xi(x) = P(\xi < x) = 1 - P(\xi \geq x) = 1 - P(\xi_1 \geq x, \xi_2 \geq x) \stackrel{\text{indep.}}{=} 1 - P(\xi_1 \geq x) \cdot P(\xi_2 \geq x) = 1 - ((1-p_1)(1-p_2))^{x-1} \\ = 1 - (1 - (p_1 + p_2 - p_1 p_2))^{x-1} \Rightarrow \xi \sim \text{Geo}(p_1 + p_2 - p_1 p_2)$$

N15.  $\mu = \begin{bmatrix} 0 \\ -3 \\ 1 \end{bmatrix}$ ,  $K = \begin{bmatrix} 5 & -2 & -1 \\ -2 & 1 & 3 \\ -1 & 3 & 35 \end{bmatrix}$ ,  $\xi = \begin{bmatrix} \xi_1 \\ \xi_2 \\ \xi_3 \end{bmatrix}$  |  $\mu = \begin{bmatrix} E\xi_1 \\ E\xi_2 \\ E\xi_3 \end{bmatrix}$ ,  $K = \begin{bmatrix} \text{Cov}(\xi_1, \xi_1) & \text{Cov}(\xi_1, \xi_2) & \text{Cov}(\xi_1, \xi_3) \\ \text{Cov}(\xi_2, \xi_1) & \text{Cov}(\xi_2, \xi_2) & \text{Cov}(\xi_2, \xi_3) \\ \text{Cov}(\xi_3, \xi_1) & \text{Cov}(\xi_3, \xi_2) & \text{Cov}(\xi_3, \xi_3) \end{bmatrix}$

a)  $\eta_a = \xi_1 - \xi_3$

$$E\eta_a = E(\xi_1 - \xi_3) = E\xi_1 - E\xi_3 = 0 - 1 = -1$$

$$\text{Var } \eta_a = \text{Var}(\xi_1 - \xi_3) = \text{Var } \xi_1 - 2\text{Cov}(\xi_1, \xi_3) + \text{Var } \xi_3 = 5 + 2 + 35 = 42$$

b)  $\eta_b = 2\xi_1 - \xi_2 + 3\xi_3$

$$E\eta_b = E(2\xi_1 - \xi_2 + 3\xi_3) = 2E\xi_1 - E\xi_2 + 3E\xi_3 = 0 + 3 + 3 = 6$$

$$\text{Var } \eta_b = \text{Var}(2\xi_1 - \xi_2 + 3\xi_3)$$

$$\text{Cov}((A \pm B), C) = E((A \pm B)C) - E(A \pm B) \cdot EC = E(AC) \pm E(BC) - EA \cdot EC \mp EB \cdot EC = \\ = \text{Cov}(A, C) \pm \text{Cov}(B, C)$$

$$\text{Var}(A^{(1)} \pm B^{(2)} \pm C) = \text{Var}(A \pm B) \pm 2\text{Cov}((A \pm B), C) + \text{Var } C = \\ = \text{Var } A + \text{Var } B + \text{Var } C \pm 2\text{Cov}(A, B) \pm 2\text{Cov}(A, C) \pm (\pm 1) \cdot 2\text{Cov}(B, C)$$

$$\text{Var } \eta_b = \text{Var}(2\xi_1) + \text{Var } \xi_2 + \text{Var}(3\xi_3) - 2\text{Cov}(2\xi_1, \xi_2) + 2\text{Cov}(2\xi_1, 3\xi_3) - 2\text{Cov}(\xi_2, 3\xi_3) = \\ = 2^2 \cdot 5 + 1 + 3^2 \cdot 35 - 2 \cdot 2 \cdot (-2) + 2 \cdot 2 \cdot 3 \cdot (-1) - 2 \cdot 3 \cdot 3 = 314$$

c)  $\eta_c = -2\xi_1 + 3\xi_2 - \xi_3$

$$E\eta_c = E(-2\xi_1 + 3\xi_2 - \xi_3) = -2E\xi_1 + 3E\xi_2 - E\xi_3 = -2 \cdot 0 - 3 \cdot 3 - 1 = -10$$

$$\text{Var } \eta_c = \text{Var}(3\xi_2 - 2\xi_1 - \xi_3) = \text{Var}(3\xi_2) + \text{Var}(2\xi_1) + \text{Var}(\xi_3) - 2\text{Cov}(3\xi_2, 2\xi_1) - 2\text{Cov}(3\xi_2, \xi_3) + 2\text{Cov}(2\xi_1, \xi_3) \\ = 3^2 \cdot 1 + 2^2 \cdot 5 + 35 - 3 \cdot 2 \cdot 2 \cdot (-2) - 3 \cdot 2 \cdot 3 + 2 \cdot 2 \cdot (-1) = 66$$

N16.  $K = \begin{bmatrix} 1 & -1 & 1 \\ -1 & 3 & \lambda \\ 1 & \lambda & 2 \end{bmatrix}$ ,  $\xi = \begin{bmatrix} \xi_1 \\ \xi_2 \\ \xi_3 \end{bmatrix}$ ,  $\xi = \lambda\xi_1 + 2\xi_2 - \xi_3$  arg min  $\text{Var } \xi$  - ?

$$\text{Var } \xi = \text{Var}(\lambda\xi_1) + \text{Var}(2\xi_2) + \text{Var}(\xi_3) + 2\text{Cov}(\lambda\xi_1, 2\xi_2) - 2\text{Cov}(\lambda\xi_1, \xi_3) - 2\text{Cov}(2\xi_2, \xi_3) = \\ = \lambda^2 \cdot 1 + 2^2 \cdot 3 + 2 + 2 \cdot 2 \cdot \lambda \cdot (-1) - 2 \cdot \lambda \cdot 1 - 2 \cdot 2 \cdot \lambda = \lambda^2 - 10\lambda + 14$$

Since  $\text{Var } \xi$  must be non-negative, we only consider  $\lambda: \lambda^2 - 10\lambda + 14 \geq 0 \Rightarrow \arg \min \text{Var } \xi = 5 \pm \sqrt{11}$

N17.  $\xi_1, \xi_2, \dots, \xi_{100}$  are independent and identically distributed,  $\xi_i \sim \text{Exp}(\lambda)$

Let  $\eta_1 = \sum_{k=1}^{100} \xi_k$ ,  $\eta_2 = \sum_{k=1}^{30} \xi_k$

$$E\eta_1 = E\left(\sum_{k=1}^{100} \xi_k\right) = 100 E\xi_1 = \frac{100}{\lambda}, \quad E\eta_2 = \frac{30}{\lambda} \Rightarrow \mu = \begin{bmatrix} \frac{100}{\lambda} \\ \frac{30}{\lambda} \end{bmatrix}$$



$$\text{Var } \eta_1 = \text{Var} \left( \sum_{k=1}^{100} \xi_k \right) = \text{Var } \xi_1 \cdot \underbrace{E(100)}_{100} + E(\xi_1)^2 \cdot \underbrace{\text{Var}(100)}_0 = 100 \text{Var } \xi_1 = \frac{100}{\lambda^2}$$

$$\text{Var } \eta_2 = \frac{30}{\lambda^2}$$

$$\begin{aligned} \text{Cov}(\eta_1, \eta_2) &= E(\eta_1 \eta_2) - E\eta_1 \cdot E\eta_2 = E \left( \sum_{k=1}^{100} \sum_{L=1}^{30} \xi_k \xi_L \right) - \frac{3000}{\lambda^2} = \sum_{k=1}^{100} \sum_{L=1}^{30} E(\xi_k \xi_L) - \frac{3000}{\lambda^2} = \\ &= \sum_{k=1}^{100} \sum_{L=1}^{30} E\xi_k E\xi_L - \frac{3000}{\lambda^2} = \frac{3000}{\lambda^2} - \frac{3000}{\lambda^2} = 0 = \text{Cov}(\eta_2, \eta_1) \end{aligned}$$

$$K = \begin{bmatrix} \frac{100}{\lambda^2} & 0 \\ 0 & \frac{30}{\lambda^2} \end{bmatrix}$$

N18. Let  $\xi$  be the amount of letters (out of  $N$ ) that have reached their destination.

Let  $\xi_i$  be the indicator that the  $i$ -th letter has reached its destination.

$$\xi = \sum_{i=1}^N \xi_i \Rightarrow \text{Var } \xi = \text{Var} \left( \sum_{i=1}^N \xi_i \right) = \sum_{i=1}^N \sum_{j=1}^N \text{Cov}(\xi_i, \xi_j)$$

$$\text{Cov}(\xi_i, \xi_j) = E(\xi_i \xi_j) - E\xi_i E\xi_j = P(\xi_i=1, \xi_j=1) - P(\xi_i=1)P(\xi_j=1)$$

$$\left. \begin{aligned} P(\xi_i=1) &= \frac{(N-1)!}{N!} = \frac{1}{N} \quad \forall i \in [1; N] \\ P(\xi_i=1, \xi_j=1) &= \frac{(N-2)!}{N!} = \frac{1}{N(N-1)} \quad \forall i, j \in [1; N], i \neq j \end{aligned} \right\} \Rightarrow \text{Cov}(\xi_i, \xi_j) = \begin{cases} \frac{N-1}{N^2}, & i=j \\ \frac{1}{N^2(N-1)}, & i \neq j \end{cases}$$

$$\text{Var } \xi = \sum_{i=1}^N \left( \frac{N-1}{N^2} + \sum_{\substack{j=1 \\ j \neq i}}^N \frac{1}{N^2(N-1)} \right) = \sum_{i=1}^N \left( \frac{N-1}{N^2} + \frac{1}{N^2} \right) = \frac{N}{N} = 1$$

N19. Let  $\xi$  be the sum of the ten random numbers from the set  $\{1, 2, \dots, 100\}$ .

Let  $\xi_i$  be the indicator that the number  $i$  was taken.

$$E\xi_i = P(\xi_i=1) = \frac{10 \cdot A_{99}^9}{A_{100}^{10}} = \frac{10 \cdot 99! \cdot 90!}{100! \cdot 90!} = \frac{1}{10}, \text{ where } A_n^k \text{ is an arrangement of } k \text{ from } n.$$

$$E(\xi_i \xi_j) = P(\xi_i=1, \xi_j=1) = \frac{A_{10}^2 \cdot A_{98}^8}{A_{100}^{10}} = \frac{10 \cdot 9 \cdot 98! \cdot 90!}{100! \cdot 90!} = \frac{90}{9900} = \frac{1}{110}, \quad i \neq j$$

$$\xi = \sum_{i=1}^{100} i \cdot \xi_i \Rightarrow E\xi = E \left( \sum_{i=1}^{100} i \cdot \xi_i \right) = \sum_{i=1}^{100} i \cdot E\xi_i = \frac{1}{10} \sum_{i=1}^{100} i = \frac{100 \cdot 101}{2 \cdot 10} = 505$$

$$\Rightarrow \text{Var } \xi = \text{Var} \left( \sum_{i=1}^{100} i \cdot \xi_i \right) = \sum_{i=1}^{100} \sum_{j=1}^{100} \text{Cov}(i \cdot \xi_i, j \cdot \xi_j) = \sum_{i=1}^{100} \sum_{j=1}^{100} i \cdot j \cdot \text{Cov}(\xi_i, \xi_j) =$$

$$= \sum_{i=1}^{100} i \cdot \sum_{j=1}^{100} j \cdot (E(\xi_i \xi_j) - E\xi_i E\xi_j) = \sum_{i=1}^{100} i \cdot \left( i \cdot (E\xi_i - E\xi_i^2) + \sum_{\substack{j=1 \\ j \neq i}}^{100} j \cdot (E(\xi_i \xi_j) - E\xi_i E\xi_j) \right) =$$

$$= \sum_{i=1}^{100} i \cdot \left( i \cdot \left( \frac{1}{10} - \frac{1}{100} \right) + \sum_{\substack{j=1 \\ j \neq i}}^{100} j \cdot \left( \frac{1}{110} - \frac{1}{100} \right) \right) = \sum_{i=1}^{100} \frac{9}{100} i^2 + \sum_{i=1}^{100} i \cdot \left( -\frac{1}{1100} \right) \cdot \left( \sum_{j=1}^{100} j \right) - i =$$

$$= \frac{9}{100} \cdot \sum_{i=1}^{100} i^2 + \left( -\frac{1}{1100} \right) \sum_{i=1}^{100} i \cdot (50 \cdot 101 - i) = \frac{100}{1100} \sum_{i=1}^{100} i^2 - \frac{5050}{1100} \cdot \sum_{i=1}^{100} i =$$

$$= \frac{100 \cdot 101 \cdot 201}{6 \cdot 11} - \frac{50^2 \cdot 101^2}{1100} = \frac{101}{11} (3350 - 2525) = 7575$$