

Moment Generating Functions

$$\Sigma_{XY} = \begin{bmatrix} \text{Var}[X] & \text{Cov}[X, Y] \\ \text{Cov}[Y, X] & \text{Var}[Y] \end{bmatrix}$$

$$\text{Var}[W_n] = \sum_{i=1}^n \text{Var}[X_i] + 2 \sum_{i=1}^{n-1} \sum_{j=i+1}^n \text{Cov}[X_i, X_j]$$

$$\text{Cov}[X_1, X_2], \text{Cov}[X_1, X_3], \text{Cov}[X_2, X_3]$$

If $\mathbf{X} = [X_1 \ X_2 \ X_3]'$, the covariance matrix of $\mathbf{X}$ is

$$\mathbf{C}_\mathbf{X} = \begin{bmatrix} \text{Var}[X_1] & \text{Cov}[X_1, X_2] & \text{Cov}[X_1, X_3] \\ \text{Cov}[X_2, X_1] & \text{Var}[X_2] & \text{Cov}[X_2, X_3] \\ \text{Cov}[X_3, X_1] & \text{Cov}[X_3, X_2] & \text{Var}[X_3] \end{bmatrix} \quad (8.23)$$

Example 9.2

$$X_i = \begin{cases} 1 & \text{if person } i \text{ draws own hat} \\ 0 & \text{otherwise} \end{cases}$$

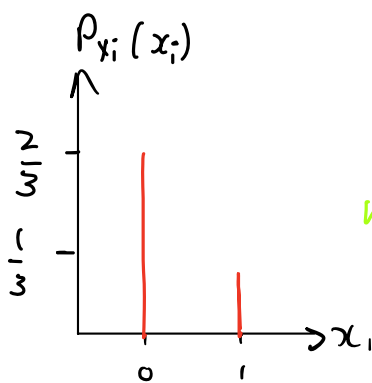
$S_{X_i} = [0, n]$

Number of matches $V_n = X_1 + X_2 + \dots + X_n$

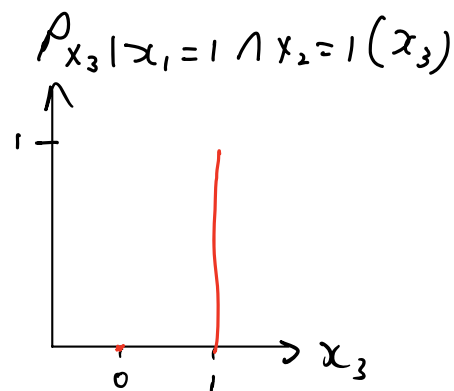
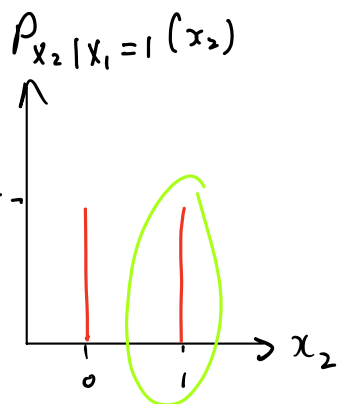
$$\begin{aligned} E[V_n] &= E[X_1 + X_2 + \dots + X_n] \\ &= E[X_1] + E[X_2] + \dots + E[X_n] = n \cdot \frac{1}{n} = 1 \end{aligned}$$

$$E[X_i] = \sum_{x \in S_{X_i}} x \cdot \underbrace{P_{X_i}(x)}_{\substack{\text{green arrow} \\ \text{from } E[X_i^2]}} = 0 \cdot \cancel{P_{X_i}(0)} + 1 \cdot P_{X_i}(1) = 1 \cdot \frac{1}{n} = \frac{1}{n} = E[X_i^2]$$

What is PMF of X_i (for $n=3$)?



$$\frac{1}{n-1} = 0.5$$



PMF of person 1 drawing own hat

person 2 drawing own hat

person 3 drawing own hat

$$P_{X_i}(1) = \frac{1}{n}$$

$$P_{X_i}(0) = 1 - \frac{1}{n} = \frac{n-1}{n}$$

$$\text{Var}[V_n] = \sum_{i=1}^n \text{Var}[X_i] + 2 \sum_{i=1}^{n-1} \sum_{j=i+1}^n \text{Cov}[X_i, X_j]$$

$$= n \text{Var}[X_i] + n(n-1) \text{Cov}[X_i, X_j] \quad \text{iid.}$$

$$= 1 - \frac{1}{n} + 1 - \frac{n-1}{n}$$

$$= 2 + \frac{-1 - n + 1}{n} = 2 - 1 = 1 \rightarrow$$

$$\begin{aligned} \text{Var}[X_i] &= E[X_i^2] - E[X_i]^2 \\ &= \frac{1}{n} - \left(\frac{1}{n}\right)^2 \end{aligned}$$

$$\text{Cov}[X_i, X_j] = r_{X_i X_j} - \mu_{X_i} \mu_{X_j}$$

$$= E[X_i X_j] - E[X_i] E[X_j]$$

$$= \frac{1}{n(n-1)} - \frac{1}{n} \cdot \frac{1}{n} = \frac{1}{n(n-1)} - \frac{1}{n^2}$$

$$E[X_i X_j] = P_{X_i X_j}(x_i, x_j) \cdot 1 \cdot 1 = P_{X_i | X_j}(1 | 1) P_{X_j}(1)$$

$$= \frac{1}{n-1} \cdot \frac{1}{n} = \frac{1}{n(n-1)}$$

$$E[X]$$

$$E[X^2]$$

$$E[X^n]$$

Moment Generating Functions

$$\phi_x(s) = E[e^{sx}]$$

$$E[X^n] = m_n = \left. \frac{d^n \phi_x(s)}{ds^n} \right|_{s=0}$$

$$\begin{aligned} f(x) &= \frac{\lambda}{x} & g(s) &= \lambda - s \\ f'(x) &= -\frac{\lambda}{x^2} & g'(s) &= -1 \end{aligned}$$

Example 9.4

$$\phi_x(s) = \frac{\lambda}{\lambda - s}$$

Chain rule: $\frac{d}{dx} f(g(x)) = f'(g(x)) \cdot g'(x)$

$$\begin{aligned} E[X] &= \left. \frac{d\phi(s)}{ds} \right|_{s=0} = \left. \frac{\lambda}{(\lambda - s)^2} \right|_{s=0} = \frac{\lambda}{\lambda^2} = \frac{1}{\lambda} \\ E[X^2] &= \left. \frac{d^2\phi(s)}{ds^2} \right|_{s=0} = \left. \frac{2\lambda}{(\lambda - s)^3} \right|_{s=0} = \frac{2\lambda}{\lambda^3} = \frac{2}{\lambda^2} \end{aligned}$$

$$E[X^n] = \left. \frac{d^n \phi(s)}{ds^n} \right|_{s=0} = \left. \frac{n! \lambda}{(\lambda - s)^{n+1}} \right|_{s=0} = \frac{n!}{\lambda^n}$$

Example 9.5

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J and K are independent random variables with probability mass functions

$$\begin{array}{c|ccc} j & 1 & 2 & 3 \\ \hline P_J(j) & 0.2 & 0.6 & 0.2 \end{array}, \quad \begin{array}{c|cc} k & -1 & 1 \\ \hline P_K(k) & 0.5 & 0.5 \end{array}. \quad (9.31)$$

Find the MGF of $M = J + K$. What are $P_M(m)$ and $E[M^3]$?

$$\begin{aligned} \phi_J(s) &= E[e^{sJ}] = \sum_{j \in S_J} e^{sj} \cdot P_J(j) \\ &= e^s \cdot 0.2 + e^{2s} \cdot 0.6 + e^{3s} \cdot 0.2 \end{aligned}$$

$$\phi_K(s) = E[e^{sK}] = e^{-s} \cdot 0.5 + e^s \cdot 0.5$$

$$\phi_M(s) = \phi_J(s)\phi_K(s) = 0.1 + 0.3e^s + 0.2e^{2s} + 0.3e^{3s} + 0.1e^{4s}$$

M	0	1	2	3	4
$P_M(m)$	0.1	0.3	0.2	0.3	0.1

$$\begin{aligned} E[M^3] &= \left. \frac{d^3 \phi_M(s)}{ds^3} \right|_{s=0} = \sum_{m \in S_M} m^3 \cdot P_M(m) \\ &= 0.3e^s + 0.2 \cdot 2^3 e^{2s} + 0.3(3^3)e^{3s} + 0.1(4^3)e^{4s} \Big|_{s=0} \\ &= 16.4 \end{aligned}$$