

**FUNCTIONS OF RANDOM VARIABLES**

- i. **METHOD OF DISTRIBUTION FUNCTIONS**
- ii. **ONE-TO-ONE TRANSFORMATIONS.**
- iii. **TWO-TO TWO TRANSFORMATIONS. (JOINT DISTRIBUTION OF FUNCTIONS OF RANDOM VARIABLES )**
- iv. **METHOD OF MOMENT-GENERATING FUNCTIONS.**

**Q1)** If  $X \sim \text{Uniform}(0,1)$ , find the pdf of  $Y = -2\ln X$ . Name the distribution and its parameter values.

**Q2)** If  $X \sim \text{Normal}(\mu, \sigma^2)$ , find the pdf of  $Y = e^X$ .

**Q3)** If  $X \sim \text{Exponential}(1)$ , find the pdf of  $Y = -\ln X$ .

**Q4)** If  $X \sim \text{Uniform}(0,1)$ , find the pdf of  $Y = \sqrt{X}$ .

**Q5)** The pdf of  $X$  is given by  $f_X(x) = \frac{1}{2}x$ ;  $0 < x < 2$ .

a. Find the pdf of  $Y = X^3$ .

b. Find  $P\left(\frac{1}{2} < X < 1\right)$  and  $P\left(\frac{1}{8} < Y < 1\right)$ . Are they the same or different? Why?

**Q6)** If  $X \sim \text{Uniform}(0,1)$  independent of  $Y \sim \text{Exponential}(1)$ , find the distribution of  $Z = X + Y$ .

**Q7)** Let  $X_1$  and  $X_2$  be independent  $\text{Exp}(\lambda)$  r.v. Find the joint density of  $Y_1 = X_1 + X_2$  and  $Y_2 = e^{X_1}$ .

**Q8)** Let  $X_1 \sim \text{Exp}(\lambda_1)$  independent of  $X_2 \sim \text{Exp}(\lambda_2)$  r.v.. Find:

a. The cumulative distribution function of  $Z = \frac{X_1}{X_2}$ .

b.  $P(X_1 < X_2)$ .

**Q9)** Let  $X$  and  $Y$  be distributed as independent  $\text{Uniform}(0,1)$  r.v. ,

a. Find the joint density function of  $Z_1 = X + Y$  and  $Z_2 = Y$ .

b. Find the marginal pdf of  $Z_1$  from the joint density.

**Q10)** Let  $X$  and  $Y$  be distributed as independent  $\text{Exp}(1)$  r.v., find:

a. The joint density function of  $Z = X + Y$  and  $U = \frac{X}{X+Y}$ .

b. Find the marginal pdf of  $U$ .

**Q11)** Let  $X$  and  $Y$  have independent  $\text{Gamma}(\alpha, \lambda)$  distributions.

a. Find the joint pdf of  $U = \frac{X}{X+Y}$  and  $V = X + Y$ .

b. Show that the marginal density of  $U$  is a Beta distribution.

- If  $X_i$  indpt.  $Exp(\lambda)$ , then the sum  $\sum_{i=1}^{i=n} X_i \sim Gamma(n, \lambda)$
- If  $X_i$  indpt.  $Gamma(\alpha_i, \beta)$ , then the sum  $\sum_{i=1}^{i=n} X_i \sim Gamma(\sum_{i=1}^{i=n} \alpha_i, \beta)$
- If  $X_i$  indpt.  $Normal(\mu_i, \sigma_i^2)$ , then the sum  $\sum_{i=1}^{i=n} X_i \sim Normal(\sum_{i=1}^{i=n} \mu_i, \sum_{i=1}^{i=n} \sigma_i^2)$
- If  $X_i$  indpt.  $Normal(\mu_0, \sigma_0^2)$ , then the sum  $\sum_{i=1}^{i=n} X_i \sim Normal(n\mu_0, n\sigma_0^2)$
- If  $Z \sim Normal(0,1)$ , then the  $Z^2 \sim \chi_1^2$
- If  $X \sim \chi_n^2$ , ind, of  $Y \sim \chi_m^2$ , then the  $X + Y \sim \chi_{n+m}^2$
- If  $Z_1 \sim Normal(0,1)$ , ind, of  $Z_2 \sim Normal(0,1)$  then the  $Z_1 + Z_2 \sim \chi_2^2$