

The another random variables is

Continuous random variables.

The special prob. dist. for Continuous Random variables are

1- The uniform density:

Def:- a random variable X has a uniform density and it is referred to as a continuous uniform random variable iff its prob. density is given by

$$f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta - \alpha} & \text{for } \alpha \leq x \leq \beta \\ 0 & \text{o.w.} \end{cases}$$

Theorem:- The mean and variance of this dist. is given by $\mu = \frac{\alpha + \beta}{2}$ and variance $(\sigma_x^2) = \frac{(\beta - \alpha)^2}{12}$

i.e. $X \sim U(\alpha, \beta)$ it is parameters or constant.

For Ex: $X \sim U(2, 4)$ Find the uniform dist.

② The mean of this dist.

③ The variance

solⁿ

$$f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta - \alpha} & \alpha \leq x \leq \beta \\ 0 & \text{o.w.} \end{cases}$$

$$\text{① } f(x; 2, 4) = \begin{cases} \frac{1}{2} & 2 \leq x \leq 4 \\ 0 & \text{o.w.} \end{cases}$$

$$\text{② } \mu = \frac{\alpha + \beta}{2} = \frac{2 + 4}{2} = \frac{6}{2} = 3$$

$$\text{③ } \text{variance}(x) = \text{var}(x) = \frac{(\beta - \alpha)^2}{12} = \frac{(4 - 2)^2}{12} = \frac{2^2}{12} = \frac{4}{12} = \frac{1}{3}$$

2- The Gamma distribution

Def:- A random variable X has a gamma dist. and it is referred to as a gamma random variable iff its prob. dist. is given by

$$f(x) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} \cdot x^{\alpha-1} e^{-\frac{x}{\beta}} & \text{for } x > 0 \\ 0 & \text{o.w.} \end{cases} \quad \alpha, \beta > 0$$

$\Gamma(\alpha)$ is equal to $(\alpha-1)!$ for ex. $\Gamma(3) = (3-1)! = 2! = 2$

The mean of this dist. is $\mu = E(X) = \alpha\beta$
 $\text{Var}(X) = \alpha\beta^2$

Then $X \sim G(\alpha, \beta) \Rightarrow \alpha, \beta$ are parameters.

Ex:- $X \sim G(3, 5)$, Find the dist.

\hookrightarrow mean
 \hookrightarrow var(X)

Solⁿ

$$f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} \cdot x^{\alpha-1} e^{-\frac{x}{\beta}} & x > 0 \\ 0 & \text{o.w.} \end{cases} \quad \alpha, \beta > 0$$

$$f(x; 3, 5) = \begin{cases} \frac{1}{5^3 \Gamma(3)} \cdot x^{3-1} e^{-\frac{x}{5}} & x > 0 \\ 0 & \text{o.w.} \end{cases} = \begin{cases} \frac{1}{5^3 (2)} x^2 e^{-\frac{x}{5}} & x > 0 \\ 0 & \text{o.w.} \end{cases}$$

$$E(X) = \alpha\beta = 3(5) = 15$$

$$\text{Var}(X) = \alpha\beta^2 = 3(5^2) = 3(25) = 75$$

3- Exponential distribution :-

Def: A random variable X has an exponential dist. iff prob. is given by $P(x; \theta) = \begin{cases} \frac{1}{\theta} e^{-\frac{x}{\theta}} & \text{for } x \geq 0 \\ 0 & \text{o.w.} \end{cases}$

or we can write another formula

$$f(x; \beta) = \begin{cases} \frac{1}{\beta} e^{-\frac{x}{\beta}} & \text{for } x \geq 0 \\ 0 & \text{o.w.} \end{cases}$$

we can derive this dist. from the gamma dist.

where $\alpha = 1$

$$f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-\frac{x}{\beta}} & \\ 0 & \text{o.w.} \end{cases}$$

← Gamma dist.

$$f(x; 1, \beta) = \begin{cases} \frac{1}{\beta^1 \Gamma(1)} x^{1-1} e^{-\frac{x}{\beta}} & \\ 0 & \text{o.w.} \end{cases} = \begin{cases} \frac{1}{\beta} e^{-\frac{x}{\beta}} & x \geq 0 \\ 0 & \text{o.w.} \end{cases}$$

Because $\Gamma(1) = (1-1)! = 0! = 1$, $x^0 = 1$

$$\text{The Mean} = \mu = E(x) = \beta$$

$$\text{var}(x) = \beta^2 = \sigma^2$$

Ex: $X \sim \text{Exp}(5)$, Find the dist.

$$1- f(x; \theta) = \begin{cases} \frac{1}{\theta} e^{-\frac{x}{\theta}} & x \geq 0 \\ 0 & \text{o.w.} \end{cases}$$

ε mean
ε variance

$$= \begin{cases} \frac{1}{5} e^{-\frac{x}{5}} & x \geq 0 \\ 0 & \text{o.w.} \end{cases}$$

$$\text{Mean} = E(X) = \theta = 5$$

$$\text{var}(X) = \sigma^2 = \theta^2 = (5)^2 = 25$$

4- Chi-square distribution χ^2

Def: A random variable X has a chi-square dist iff its prob. dist. is given by:

$$f(x) = \begin{cases} \frac{1}{2^{\frac{V}{2}} \Gamma(\frac{V}{2})} x^{\frac{V}{2}-1} e^{-\frac{x}{2}} & \text{for } x > 0 \\ 0 & \text{o.w.} \end{cases}$$

The mean $= E(X) = \mu = V$ where V means the degree of freedom
 $\text{var}(X) = \sigma_x^2 = 2V$

$$X \sim \chi^2(V)$$

Ex: $X \sim \chi^2(4)$, find the dist.

$$f(x) = \begin{cases} \frac{1}{2^{\frac{4}{2}} \Gamma(\frac{4}{2})} x^{\frac{4}{2}-1} e^{-\frac{x}{2}} & \begin{matrix} \text{mean}(X) \\ \text{varian}(X) \end{matrix} \\ 0 & \text{o.w.} \end{cases} \quad x > 0$$

$$= \begin{cases} \frac{1}{4} x e^{-\frac{x}{2}} & x > 0 \\ 0 & \text{o.w.} \end{cases}$$

The mean $= \mu = V = 4$
 $\text{var}(X) = \sigma_x^2 = 2(V) = 8$

5- Beta distribution

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Def: A random variable  $X$  has a beta distribution iff its prob. distr is given by

$$f(x) = \begin{cases} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1} & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\mu = E(X) = \text{Mean} = \frac{\alpha}{\alpha+\beta}$$

$$\text{Var}(X) = \sigma_x^2 = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$$

Ex:  $X \sim \text{Beta}(\alpha, \beta) \Rightarrow X \sim \text{Beta}(3, 5)$

$$f(x) = \begin{cases} \frac{\Gamma(8)}{\Gamma(3)\Gamma(5)} x^2 (1-x)^4 & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\mu = E(X) = \frac{\alpha}{\alpha+\beta} = \frac{3}{8}$$

$$\text{Var}(X) = \sigma_x^2 = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)} = \frac{15}{8^2(9)576} = \frac{15}{576} = 0.026$$

## 6- Normal distribution

Def: A random variable  $X$  has a normal distribution iff its prob. dist. is given by

$$f(x; \mu, \sigma) = \begin{cases} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} & -\infty \leq x \leq \infty \\ 0 & \text{o.w.} \end{cases} \quad \sigma > 0$$

The Mean =  $E(X) = \mu = \underline{\underline{M}}$   
 $\text{var}(X) = \sigma_x^2 = \underline{\underline{\sigma^2}}$

Ex<sup>2</sup>:  $X \sim N(3, 3)$ , find the dist.

$$f(x, 3, 3) = \begin{cases} \frac{1}{3\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-3}{3}\right)^2} & \begin{matrix} \text{Mean}(x) \\ \text{variance}(x) \end{matrix} \\ 0 & \text{o.w.} \end{cases} \quad -\infty \leq x \leq \infty$$

Mean =  $\mu = \underline{\underline{3}}$   
 $\text{var}(X) = \sigma_x^2 = \underline{\underline{3^2 = 9}}$

Note: if the random dist.  $\mu=0$  and  $\sigma=1$  is referred to as the standard normal dist.

Ex:  $X \sim N(\mu, \sigma^2) \Rightarrow X \sim N(0, 1)$  then

$$f(x, 0, 1) = \begin{cases} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} & -\infty \leq x \leq \infty \\ 0 & \text{o.w.} \end{cases}$$