

# Continuous Probability Distributions

## 1) Normal (or Gaussian) D<sup>n</sup>.

$$X \sim N(\mu, \sigma^2) = \frac{(x-\mu)^2}{2\sigma^2}$$

$$f_x(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} ; \sigma > 0$$

$$E(X) = \mu$$

$$\text{Var}(X) = \sigma^2$$

## 2) Uniform D<sup>n</sup>

$$X \sim \text{uniform}(a, b)$$

$$f_x(x) = \frac{1}{b-a} ; a \leq x \leq b$$

$$E(X) = \frac{a+b}{2} ; \text{Var}(X) = \frac{(b-a)^2}{12}$$

3) Pareto D" ( $\mathcal{D}''$ : of wealth in a society - fitting the trend that a large portion of wealth is held by a small fraction of the population).

$$X \sim \text{Pareto}(\alpha, \beta)$$

↑  
scale      ↗ shape

$$f_X(x) = \frac{\beta \alpha^\beta}{x^{\beta+1}} ; x > \alpha$$

$\alpha, \beta > 0$

$$E(X) = \frac{\beta \alpha}{\beta - 1} ; \beta > 1 \quad (\infty \text{ otherwise})$$

$$\text{Var}(X) = \frac{\beta \alpha^2}{(\beta - 1)^2 (\beta - 2)} ; \beta > 2$$

4) Exponential D". (Waiting times)

$X \sim \exp(\theta) ; \theta \sim \frac{1}{\lambda}$ ; where  $\lambda$  is rate (think Poisson D").

$$f_X(x) = \frac{1}{\theta} e^{-\frac{x}{\theta}} ; x \geq 0 ; \theta > 0 ; E(X) = \theta, \text{Var}(X) = \theta^2$$

5) Gamma D<sup>n</sup> (Waiting time D<sup>n</sup>)

\* generalization of exponential &  $\chi^2$  D<sup>n</sup>

$$X \sim \Gamma(\alpha, \beta)$$

↑      ↑  
Shape    scale  
 $x^{\alpha-1} e^{-x/\beta}$

$$f_x(x) = \frac{1}{\Gamma(\alpha)\beta^\alpha}$$

$$E(X) = \alpha\beta$$

$$\text{Var}(X) = \alpha\beta^2$$

6) Other examples of continuous D<sup>n</sup> :-  $X_1, t_2, F_{x_1, x_2}$

these are sampling D<sup>n</sup>.

Joint D<sup>n</sup> and Marginal D<sup>n</sup> : (for multivariate D<sup>n</sup>) .

$$P(X=x, Y=y) = P(Y=y | X=x) \underbrace{P(X=x)}_{\text{Marginal D}^n} \longleftrightarrow \text{this is analogous to cond'l prob.}$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Where  $\iint_{y|x} P(x=x, Y=y) = 1$  (Continuous case) → will discuss this below

or  $\sum_i \sum_j P(X=x_i, Y=y_i) = 1$  (Discrete case).

Continuous case :

$$f_{XY}(x, y) = f_{Y|X} \underbrace{(y|x) f_X(x)}_{\text{Marginal pdf}} = f_{X|Y}(x|y) \underbrace{f_Y(y)}_{\text{Marginal pdf}} \quad \text{where} \quad \iint_{x,y} f_{XY}(x, y) = 1$$

So far we have discussed examples of simple RVs.

Now! Some examples of D<sup>n</sup> of composite RVs.

Q) Let X & Y be independent geom. ( $p$ )

(a) Find the D<sup>n</sup> min (X, Y)

(b)  $P(Y \geq X)$

(c) Find D<sup>n</sup> of X + Y

(d)  $P(Y=y | X+Y=z)$  for  $z \geq 2$ ;  $y=1, \dots, z-1$

Soln:- (b)  $P(Y \geq X) = \sum_{x=1}^{\infty} P(X=x, Y \geq x) \xrightarrow{\text{think}} P(X=1, Y \geq X) + P(X=2, Y \geq X) + \dots P(X=\infty, Y \geq X)$

$$P(X=x, Y=y) = P(X=x) \cdot P(Y=y)$$

when X & Y  
are indepd.

$$= \sum_{x=1}^{\infty} P(X=x, Y \geq x)$$

$$\underset{x \geq 1}{\text{indep}} \sum_{x=1}^{\infty} P(X=x) P(Y \geq x) = \sum_{x=1}^{\infty} p(1-p)^{x-1} (1-p)^{x-1}$$
$$= p \sum_{x=1}^{\infty} (1-p)^{2(x-1)} = \frac{p}{2p(1-p)}$$