## Summary of Continuous Distributions for a random Variable X

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- 1 Uniform  $X \sim U(a, b)$
- 1.1 Probability Density Function f(x)

$$f(x) = \begin{cases} \frac{1}{b-a} & a \le x \le b\\ 0 & \text{Otherwise} \end{cases}$$

1.2 Cumulative Distribution Function F(x)

$$F(x) = \begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & a \le x \le b \\ 1 & x > b \end{cases}$$

Proof.

$$F(X) = \int_{a}^{b} f(u)du$$

$$= \left[\frac{u}{(b-a)}\right]_{a}^{x}$$

$$= \frac{x}{b-a} - \frac{a}{b-a}$$

$$= \frac{x-a}{b-a}$$

QED

1.3 Moment Generating Function  $M(t) = E(e^{tX})$ 

$$M(t) = \frac{e^{tb} - e^{ta}}{t(b-a)}$$

Proof.

$$M(t) = E(e^{tX})$$

$$= \int_a^b e^{tx} f(x) dx$$

$$= \int_a^b \frac{e^{tx}}{(b-a)}$$

$$= \left[\frac{e^{tx}}{b-a}\right]_a^b$$

$$= \frac{e^{tb} - e^{ta}}{b-a}$$

QED

QED

1.4 E(X)

$$E(X) = \frac{a+b}{2}$$

Proof. Symmetry Argument makes this proof trivial

Proof. Integration

$$E(X) = \int_{a}^{b} x f(x) dx$$

$$= \int_{a}^{b} \frac{x}{b-a} dx$$

$$= \left[ \frac{x^{2}}{2(b-a)} \right]_{a}^{b}$$

$$= \frac{1}{2(b-a)} [b^{2} - a^{2}]$$

$$= \frac{1}{2(b-a)} [(b-a)(b+a)]$$

$$= \frac{a+b}{2}$$

QED

1.5 Var(X)

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

*Proof.* Integration

$$E(X^{2}) = \int_{a}^{b} x^{2} f(x) dx$$

$$= \int_{a}^{b} \frac{x^{2}}{b - a} dx$$

$$= \left[ \frac{x^{3}}{3(b - a)} \right]_{a}^{b}$$

$$= \frac{1}{3(b - a)} [b^{3} - a^{3}]$$

$$= \frac{1}{3(b - a)} [(b - a)(b^{2} + ab + a^{2})]$$

$$= \frac{a^{2} + ab + b^{2}}{3}$$

Then we minus the two to obtain

$$\begin{split} Var(X) &= \frac{a^2 + ab + b^2}{3} - \frac{(a+b)^2}{4} \\ &= \frac{a^2 + ab + b^2}{3} - \frac{a^2 + 2ab + b^2}{4} \\ &= \frac{4a^2 + 4ab + 4b^2}{12} - \frac{3a^2 + 6ab + 3b^2}{12} \\ &= \frac{a^2 - 2ab + b^2}{12} \\ &= \frac{(b-a)^2}{12} \end{split}$$

QED

- 2 Exponential  $X \sim Exp(\lambda)$
- **2.1** Probability Density Function f(x)

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & x > 0\\ 0 & \text{Otherwise} \end{cases}$$

**2.2** Cumulative Distribution Function F(x)

$$F(x) = 1 - e^{-\lambda x}$$
 where  $x > 0$ 

2.3 Moment Generating Function  $M(t) = E(e^{tX})$ 

$$M(t) = \frac{\lambda}{\lambda - t}$$

2.4 E(X)

$$E(X) = \frac{1}{\lambda}$$

 $2.5 \quad Var(X)$ 

$$Var(X) = \frac{1}{\lambda^2}$$

- 3 Gamma  $X \sim \Gamma(\alpha, \lambda)$
- **3.1** Probability Density Function f(x)

$$f(x) = \begin{cases} \frac{\lambda^{\alpha} x^{\alpha - 1} e^{-\lambda x}}{\Gamma(\alpha)} & x \ge 0\\ 0 & \text{Otherwise} \end{cases}$$

Where we have

$$\Gamma(\alpha) = \int_0^\infty x^{\alpha - 1} e^{(-x)} dx$$
$$= (\alpha - 1)!$$

**3.2** Moment Generating Function  $M(t) = E(e^{tX})$ 

$$M(t) = \frac{\lambda^{\alpha}}{(\lambda - t)^{\alpha}}$$

 $3.3 \quad E(X)$ 

$$E(X) = \frac{\alpha}{\lambda}$$

 $3.4 \quad Var(X)$ 

$$Var(X) = \frac{\alpha}{\lambda^2}$$

- 4 Beta  $X \sim B(\alpha, \beta)$
- 4.1 Probability Density Function f(x)

$$f(x) = \begin{cases} \frac{x^{\alpha-1}(1-x)^{\beta-1}}{B(\alpha,\beta)} & x \in (0,1) \\ 0 & \text{Otherwise} \end{cases}$$

Where we have

$$B(\alpha, \beta) = \int_0^1 x^{\alpha - 1} (1 - x)^{\beta - 1} dx$$

4.2 E(X)

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

 $4.3 \quad Var(X)$ 

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

- 5 Normal  $X \sim N(\mu, \sigma^2)$
- 5.1 Probability Density Function f(x)

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right]$$

Where we have  $x \in \mathbb{R}$ 

**5.2** Moment Generating Function M(t)

$$M(t) = exp\left[\mu t - \frac{\sigma^2 t^2}{2}\right]$$

5.3 E(X)

$$E(X) = \mu$$

5.4 Var(X)

$$Var(X) = \sigma^2$$

Also for normal distributions we can have that X can become the standard normal distribution  $Z \sim N(0,1)$  by the calculation

$$\frac{X-\mu}{\sigma^2}$$