

Continuous Distributions : Uniform, Exponential and Gaussian

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	Sample Space	PDF	μ	σ^2	MGF	CDF
Uniform	$x \in [a, b]$	$\frac{1}{b-a}$	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$	$\begin{cases} \frac{e^{tb} - e^{ta}}{t(b-a)} & t \neq 0 \\ 1 & t = 0 \end{cases}$	$\begin{cases} 0 & a < x \\ \frac{x-a}{b-a} & x \in [a, b] \\ 1 & x \geq b \end{cases}$
Exponential	$x \in [0, \infty)$	$\frac{e^{-\frac{x}{\theta}}}{\theta}$	θ	θ^2	$\frac{1}{1-\theta t}$	$1 - e^{-\frac{x}{\theta}}$
Gaussian	$x \in \mathbb{R}$	$\frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$	μ	σ^2	$\exp\left(\mu t + \frac{1}{2}\sigma^2 t^2\right)$	$\frac{1}{2} \left[1 + \operatorname{erf}\left(\frac{x-\mu}{\sqrt{2}\sigma}\right) \right]$

1 Uniform

1.1 PDF

$$Uni(x) = \frac{1}{b-a} = \text{const.}$$

Illustration :

- Summation of PDF = 1 : $\int_a^b \frac{1}{b-a} dx = 1$

1.2 μ_{Uni}

$$\mu_{Uni} = E(X_{Uni}) = \int_a^b x f(x) dx = \int_a^b \frac{x}{b-a} dx = \frac{a+b}{2}$$

1.3 σ_{Uni}^2

$$\begin{aligned} \sigma_{Uni}^2 &= Var(X_{Uni}) = \int_a^b x^2 f(x) dx - \mu_{Uni}^2 = \frac{1}{b-a} \int_a^b x^2 dx - \frac{a^2 + 2ab + b^2}{4} \\ &= \frac{1}{b-a} \left(\frac{b^3 - a^3}{3} \right) - \frac{a^2 + 2ab + b^2}{4} = \frac{4a^2 + 4b^2 + 4ab - 3a^2 - 6ab - 3b^2}{12} = \frac{(a-b)^2}{12} \end{aligned}$$

1.4 MFG

$$M_{Uni}(t) = E[e^{t(Uni)}] = \begin{cases} t \neq 0 : \int_a^b e^{tx} \frac{dx}{b-a} = \frac{e^{tb} - e^{ta}}{t(b-a)} \\ t = 0 : \int_a^b \frac{dx}{b-a} = 1 \end{cases}$$

1.5 CDF

$$\begin{cases} 0 & a < x \\ \frac{x-a}{b-a} & x \in [a, b) \\ 1 & x \geq b \end{cases}$$

2 Exponential

2.1 PDF

$$E(x; \theta) = \frac{1}{\theta} e^{-\frac{x}{\theta}} \quad \text{or} \quad \lambda e^{-\lambda x} \quad \frac{1}{\theta} = \lambda$$

- Summation of PDF = 1 : $\int_0^\infty \frac{1}{\theta} e^{-\frac{x}{\theta}} dx = - \int_0^\infty e^{-\frac{x}{\theta}} \left(d \frac{-x}{\theta} \right) = - \left(\exp \frac{-x}{\theta} \right)_0^\infty = 1$

2.2 μ_{Exp}

$$\begin{aligned} \mu_{Exp} = E(X_{Exp}) &= \int_0^\infty x \frac{1}{\theta} e^{-\frac{x}{\theta}} dx = - \int_0^\infty x e^{-\frac{x}{\theta}} \left(d \frac{-x}{\theta} \right) = - \int_0^\infty x (d e^{-\frac{x}{\theta}}) \\ &= - \underbrace{(x e^{-\frac{x}{\theta}})_0^\infty}_0 + \int_0^\infty e^{-\frac{x}{\theta}} dx = -\theta \underbrace{\int_0^\infty e^{-\frac{x}{\theta}} d \frac{-x}{\theta}}_{-1} = \theta \end{aligned}$$

2.3 σ_{Exp}^2

$$\sigma^2 = E[X^2] - (E[X])^2$$

Now consider the second order moment

$$\begin{aligned} E[X^2] &= \int_0^\infty x^2 \frac{1}{\theta} e^{-\frac{x}{\theta}} dx = - \int_0^\infty x^2 e^{-\frac{x}{\theta}} \left(d \frac{-x}{\theta} \right) = - \int_0^\infty x^2 d e^{-\frac{x}{\theta}} \\ &= - \underbrace{(x^2 e^{-\frac{x}{\theta}})_0^\infty}_0 + 2 \int_0^\infty x e^{-\frac{x}{\theta}} dx = 2\theta \underbrace{\int_0^\infty \frac{1}{\theta} x e^{-\frac{x}{\theta}} dx}_{E(X)} = 2\theta^2 \end{aligned}$$

Thus

$$\sigma^2 = E[X^2] - (E[X])^2 = (2\theta^2) - (\theta^2) = \theta^2$$

2.4 MGF

$$\begin{aligned} M_{Exp}(t) = E[e^{t(Exp)}] &= \int_0^\infty e^{tx} \frac{1}{\theta} e^{-\frac{x}{\theta}} dx = \frac{1}{\theta} \int_0^\infty \exp\left(-x \left(\frac{1}{\theta} - t\right)\right) dx \\ &= \frac{1}{-\left(\frac{1}{\theta} - t\right)\theta} \underbrace{\int_0^\infty d \exp\left(-x \left(\frac{1}{\theta} - t\right)\right)}_{-1} = \frac{1}{1 - \theta t} \end{aligned}$$

2.5 CDF

$$CDF = \int_0^x \frac{1}{\theta} e^{-\frac{x}{\theta}} dx = - \int_0^x e^{-\frac{x}{\theta}} d \frac{-x}{\theta} = - \left(e^{-\frac{x}{\theta}} \right)_0^x = 1 - e^{-\frac{x}{\theta}}$$

3 Gaussian

3.1 PDF

$$N(x; \mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

3.2 Integration of e^{-x^2} , Review of Double Integral and Jacobian

$$\int_{-\infty}^{\infty} e^{-kx^2} dx = \sqrt{\int_{-\infty}^{\infty} e^{-kx^2} dx \int_{-\infty}^{\infty} e^{-ky^2} dy} = \left(\int_{-\infty}^{\infty} e^{-kx^2} dx \int_{-\infty}^{\infty} e^{-ky^2} dy \right)^{\frac{1}{2}} = \left(\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-k(x^2+y^2)} dx dy \right)^{\frac{1}{2}}$$

$x = r \cos \theta$, $y = r \sin \theta$, so the partial derivatives are : $x_r = \cos \theta$, $x_\theta = -r \sin \theta$, $y_r = \sin \theta$, $y_\theta = r \cos \theta$

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-k(x^2+y^2)} dx dy = \iint e^{-kr^2} J dr d\theta \quad J = \begin{vmatrix} x_r & x_\theta \\ y_r & y_\theta \end{vmatrix} = \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix} = r$$

Therefore

$$\int_{-\infty}^{\infty} e^{-kx^2} dx = \left(\int_0^{2\pi} \int_0^{\infty} e^{-kr^2} r dr d\theta \right)^{\frac{1}{2}} = \left(\underbrace{\int_0^{2\pi} d\theta}_{2\pi} \frac{-1}{2k} \underbrace{\int_0^{\infty} e^{-kr^2} d(-kr^2)}_{-1} \right)^{\frac{1}{2}} = \sqrt{\frac{\pi}{k}}$$

3.3 Sum of PDF = 1

$$\int_{-\infty}^{\infty} \left(\frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \right) dx$$

Let $t = \frac{x-\mu}{\sigma}$, so $dx = \sigma dt$

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{t^2}{2}} dt$$

Apply pervious part with $k = \frac{1}{2}$, the integral is thus

$$\frac{1}{\sqrt{2\pi}} \sqrt{\frac{\pi}{\frac{1}{2}}} = 1$$

3.4 μ

$$\mu = E[X] = \int_{-\infty}^{\infty} x \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

Let $t = \frac{x-\mu}{\sigma}$, $dx = \sigma dt$, $x = t\sigma + \mu$

$$\int_{-\infty}^{\infty} (t\sigma + \mu) \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt = \int_{-\infty}^{\infty} \frac{t\sigma}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt + \int_{-\infty}^{\infty} \frac{\mu}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$$

The first integral is zero, since $te^{-\frac{t^2}{2}}$ is an odd function

The second integral is the type in previous part with $k = \frac{1}{2}$, thus the mean of Gaussian Distribution is

$$\underbrace{\mu \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt}_{1} = \mu$$

3.5 σ^2

$$\sigma^2 = E[X^2] - (E[X])^2$$

$$E[X^2] = \int_{-\infty}^{\infty} x^2 \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

Let $t = \frac{x-\mu}{\sigma}$, $dx = \sigma dt$, $x = t\sigma + \mu$

$$\begin{aligned} & \int_{-\infty}^{\infty} (t^2\sigma^2 + 2t\sigma\mu + \mu^2) \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt \\ & \underbrace{\int_{-\infty}^{\infty} \frac{t^2\sigma^2}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt}_{\text{By-parts}} + \underbrace{\int_{-\infty}^{\infty} \frac{2t\sigma\mu}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt}_{\text{Zero, odd function}} + \underbrace{\int_{-\infty}^{\infty} \frac{\mu^2}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt}_{\mu^2} \end{aligned}$$

$$\frac{\sigma^2}{2} \int_{-\infty}^{\infty} \frac{t\sigma^2}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt + \mu^2 = -\sigma^2 \int_{-\infty}^{\infty} \frac{t}{\sqrt{2\pi}} de^{-\frac{t^2}{2}} + \mu^2 = \underbrace{\left[-\sigma^2 \frac{t}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} \right]_{-\infty}^{\infty}}_0 + \sigma^2 \underbrace{\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt}_1 + \mu^2 = \sigma^2 + \mu^2$$

Therefore

$$\text{Var}(X) = E[X^2] - \mu^2 = \sigma^2 + \mu^2 - \mu^2 = \sigma^2$$

Remark. $\left[-\sigma^2 \frac{t}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} \right]_{-\infty}^{\infty} = \frac{-\sigma^2}{\sqrt{2\pi}} \left[\frac{t}{e^{-\frac{t^2}{2}}} \right]_{-\infty}^{\infty} = \frac{-\sigma^2}{\sqrt{2\pi}} \left[\lim_{t \rightarrow \infty} \frac{t}{e^{-\frac{t^2}{2}}} - \lim_{t \rightarrow -\infty} \frac{t}{e^{-\frac{t^2}{2}}} \right]$

L'Hospital $= \frac{-\sigma^2}{\sqrt{2\pi}} \left[\lim_{t \rightarrow \infty} \frac{1}{-te^{-\frac{t^2}{2}}} - \lim_{t \rightarrow -\infty} \frac{1}{-te^{-\frac{t^2}{2}}} \right] = 0$

3.6 MGF

$$M(t) = E[e^{tx}] = \int_{-\infty}^{\infty} e^{tx} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp \left[tx - \frac{(x-\mu)^2}{2\sigma^2} \right] dx$$

Let $z = \frac{x-\mu}{\sigma}$, $dx = \sigma dz$, $x = z\sigma + \mu$

$$\begin{aligned} & \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp \left[t\sigma z + t\mu - \frac{z^2}{2} \right] dz \\ & = (\exp t\mu) \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp - \left[\frac{z^2}{2} - t\sigma z \right] dz \\ & = (\exp t\mu) \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp - \left[\frac{z^2}{2} - t\sigma z + \frac{t^2\sigma^2}{2} - \frac{t^2\sigma^2}{2} \right] dz \\ & = (\exp t\mu) \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp - \left[\frac{(z-t\sigma)^2}{2} - \frac{t^2\sigma^2}{2} \right] dz \end{aligned}$$

$$= \exp\left(t\mu + \frac{t^2\sigma^2}{2}\right) \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp - \left[\frac{(z - t\sigma)^2}{2} \right] dz$$

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp - \left[\frac{(z - t\sigma)^2}{2} \right] dz = 1 \quad \text{By applying previous integral with } k = \frac{1}{2}$$

Therefore

$$M_X(t) = \exp\left(t\mu + \frac{t^2\sigma^2}{2}\right)$$

3.7 CDF

$$CDF = \int_{-\infty}^x \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

Which is the standard Error Function

$$\frac{1}{2} \left[1 + \operatorname{erf} \left(\frac{x - \mu}{\sqrt{2\sigma^2}} \right) \right]$$

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