

uniform distribution

$$U \sim \text{Unif}(0, 1)$$

$$X \sim F(x)$$

$$\text{let } x = F^{-1}(U).$$

$$\text{then } X \sim F(x)$$

Useful when we want to generate random numbers with a particular distribution.

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if we have a complicated function  $g(x)$   
we may not be able to compute

$$E[g(x)] = \int g(x) f(x) dx \quad \text{analytically.}$$

can compute the expectation by

generating ~~and~~ realizations  $x_i$  of  $F(x)$

computing  $g(x_i)$

averaging the  $g(x_i)$  values.

Monte Carlo Method.

$$F(x) = 1 - e^{-x} \quad x \geq 0$$

$$F^{-1}(u) = -\log(1-u)$$

Simulate  $u \sim \text{Unif}(0,1)$

$$x = -\log(1-u) \quad \leftarrow \text{these values are distributed as}$$

$\text{Exp}(1)$   
exponential distribution with  $\lambda=1$

Independence.

$X_1, \dots, X_n$  are independent if.

$$P(X_1 \leq x_1, X_2 \leq x_2, \dots, X_n \leq x_n) \\ = P(X_1 \leq x_1) P(X_2 \leq x_2) \dots P(X_n \leq x_n).$$

joint CDF.

product of marginal CDFs

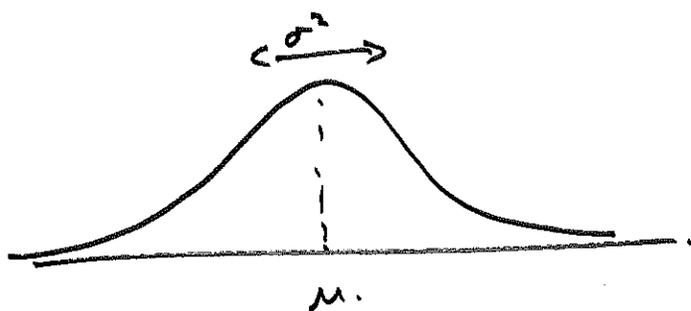
for all  $x_1, x_2, \dots, x_n$

The pdf will factorize in a similar manner.

# Normal Distribution.

- Important due to the Central Limit Theorem

CLT - the sum of a large number of iid random variables with any (finite variance) distribution will be approximately Normal.



$N(\mu, \sigma^2)$   
|                    |  
Mean                Variance

Standard Normal  $N(0, 1)$

pdf  $f(z) = C e^{-z^2/2}$   
 $= C \exp(-z^2/2)$

— Symmetric about 0  
tends to 0 as  $z \rightarrow \pm\infty$

What is  $c$ ?

Recall - to be a valid pdf

$$\int_{-\infty}^{\infty} f(z) dz = 1$$

$$\Rightarrow c = \frac{1}{\int_{-\infty}^{\infty} \exp(-z^2/2) dz}$$

$$\frac{1}{c} = \int_{-\infty}^{\infty} e^{-z^2/2} dz$$

substitution X  
by parts X

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$$I = \int_{-\infty}^{\infty} e^{-z^2/2} dz$$

$$I^2 = \int_{-\infty}^{\infty} e^{-z^2/2} dz \times \int_{-\infty}^{\infty} e^{-z^2/2} dz$$

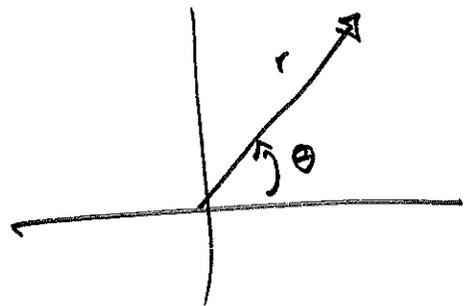
$$= \int_{-\infty}^{\infty} e^{-x^2/2} dx \int_{-\infty}^{\infty} e^{-y^2/2} dy$$

$$= \int_{y=-\infty}^{\infty} \int_{x=-\infty}^{\infty} e^{-(x^2+y^2)/2} dx dy$$

Trans form to polar coordinates.

$$x = r \cos \theta$$

$$y = r \sin \theta$$



The Jacobian of the transformation  
relates  $\underline{dx dy}$  to  $dr d\theta$

small area  
in  $x-y$   
coordinates

$$\begin{aligned} \text{Jacobian} \quad \left| \frac{d(x, y)}{d(r, \theta)} \right| &= \begin{vmatrix} \frac{dx}{dr} & \frac{dx}{d\theta} \\ \frac{dy}{dr} & \frac{dy}{d\theta} \end{vmatrix} \\ &= \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix} \\ &= r \cos^2 \theta - (-r) \sin^2 \theta \\ &= r (\cos^2 \theta + \sin^2 \theta) \\ &= r \end{aligned}$$

hence  $dx dy \rightarrow r dr d\theta$

$$I^2 = \int_{\theta=0}^{2\pi} \int_{r=0}^{\infty} e^{-r^2/2} r dr d\theta.$$

$$\text{let } u = \frac{r^2}{2}$$

$$du = r dr$$

$$I^2 = \int_{\theta=0}^{2\pi} \int_{u=0}^{\infty} e^{-u} du d\theta$$

$$= \int_{\theta=0}^{2\pi} \left( -e^{-u} \Big|_0^{\infty} \right) d\theta$$

$$= \int_{\theta=0}^{2\pi} 1 d\theta = \underline{\underline{2\pi}}$$

$$I^2 = 2\pi$$

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

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right)$$

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$$\begin{aligned}
 E[z^2] &= \frac{2}{\sqrt{2\pi}} \left[ uv \Big|_0^\infty - \int_0^\infty v \, du \right] \\
 &= \frac{2}{\sqrt{2\pi}} \left[ -z e^{-z^2/2} \Big|_0^\infty - \int_0^\infty e^{-z^2/2} \, dz \right] \\
 &= \frac{2}{\sqrt{2\pi}} \left[ 0 + \frac{1}{2} \sqrt{2\pi} \right]
 \end{aligned}$$

$$\Rightarrow E[z^2] = \frac{2}{\sqrt{2\pi}} \times \frac{\sqrt{2\pi}}{2} = 1$$

Hence  $\text{Var}[z] = 1$ .

can't do the definite integrals (eg  $\int_{-\infty}^{\infty} \exp(-x^2/2) \, dx$ ) analytically.

They can be computed accurately.

$\Phi(z)$  = CDF of standard Normal.

$$\Phi(-z) = 1 - \Phi(z)$$