

Probabilités et Statistique II

Chapitre 6. Lois continues de probabilités

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Uniform Distribution

`In[*]:= PDF[UniformDistribution[{a, b}], x]`

`Out[*]=`

$$\begin{cases} \frac{1}{-a+b} & a \leq x \leq b \\ 0 & \text{True} \end{cases}$$

`In[*]:= PiecewiseExpand[Piecewise[{{ $\frac{1}{-a+b}$, $a \leq x \leq b$ }}, 0]]`

`Out[*]=`

$$\begin{cases} \frac{1}{-a+b} & a - x \leq 0 \&\& b - x \geq 0 \\ 0 & \text{True} \end{cases}$$

`In[*]:= Simplify[Piecewise[{{ $\frac{1}{-a+b}$, $a \leq x \leq b$ }}, 0]]`

`Out[*]=`

$$\begin{cases} \frac{1}{-a+b} & a \leq x \leq b \\ 0 & \text{True} \end{cases}$$

```
In[*]:= d = UniformDistribution[{1, 11}]
```

```
Out[*]=  
UniformDistribution[{1, 11}]
```

```
In[*]:= PDF[d, x]
```

```
Out[*]=  

$$\begin{cases} \frac{1}{10} & 1 \leq x \leq 11 \\ 0 & \text{True} \end{cases}$$

```

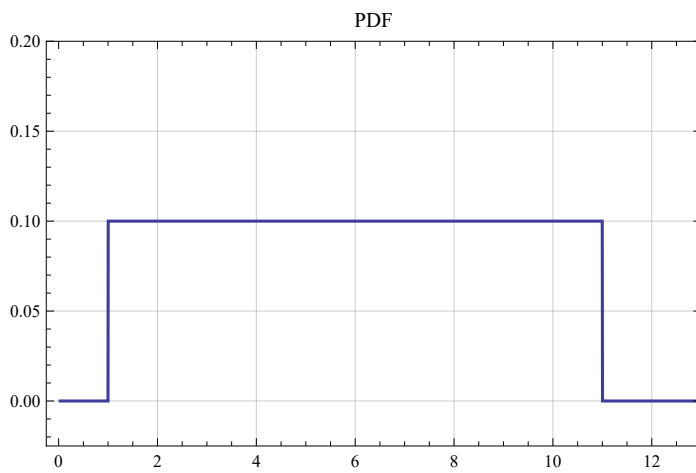
```
In[ ]:= PiecewiseExpand[Piecewise[{{ $\frac{1}{10}$ ,  $1 \leq x \leq 11$ }}, 0]]
```

```
Out[ ]:=
```

$$\begin{cases} \frac{1}{10} & 1 \leq x \leq 11 \\ 0 & \text{True} \end{cases}$$

```
In[ ]:= pdf = Plot[PiecewiseExpand[Piecewise[{{ $\frac{1}{10}$ ,  $1 \leq x \leq 11$ }}, 0]],  
  {x, 0, 15}, PlotLabel -> "PDF", PlotStyle -> ColorData[1, 1],  
  GridLines -> Automatic, PlotTheme -> "Scientific",  
  ImageSize -> Medium, PlotRange -> {{-0.25, 13}, {-0.025, 0.2}}]
```

```
Out[ ]:=
```



In[*]:= CDF[d, x]

Out[*]=

$$\begin{cases} \frac{1}{10} (-1 + x) & 1 \leq x \leq 11 \\ 1 & x > 11 \\ 0 & \text{True} \end{cases}$$

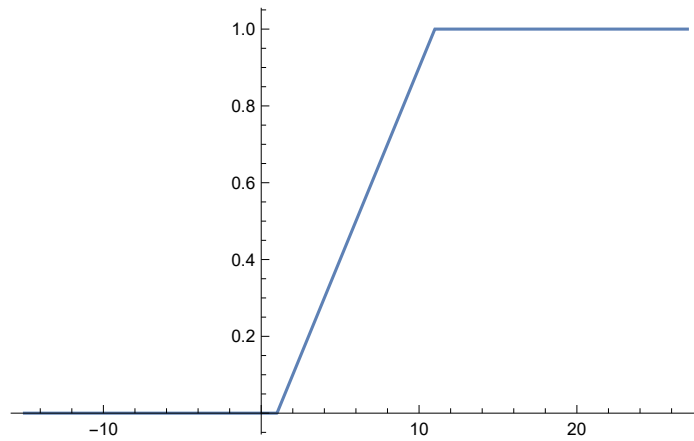
In[*]:= PiecewiseExpand[Piecewise[{{ $\frac{1}{10} (-1 + x)$, $1 \leq x \leq 11$ }, {1, $x > 11$ }}, 0]]

Out[*]=

$$\begin{cases} 1 & x > 11 \\ \frac{1}{10} (-1 + x) & 1 \leq x \leq 11 \\ 0 & \text{True} \end{cases}$$

In[*]:= Plot[Piecewise[{{1, $x > 11$ }, { $\frac{1}{10} (-1 + x)$, $1 \leq x \leq 11$ }}, 0], {x, -15., 27.}]

Out[*]=

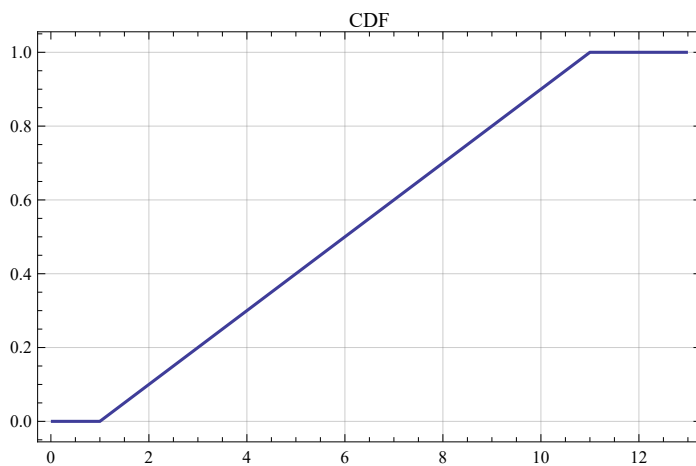


```

In[ ]:= cdf = Plot[PiecewiseExpand[
    Piecewise[{{1, x > 11}, { $\frac{1}{10}(-1 + x)$ , 1 ≤ x ≤ 11}}, 0]],
    {x, 0, 13}, PlotLabel → "CDF", PlotStyle → ColorData[1, 1],
    GridLines → Automatic,
    PlotTheme → "Scientific", ImageSize → Medium]

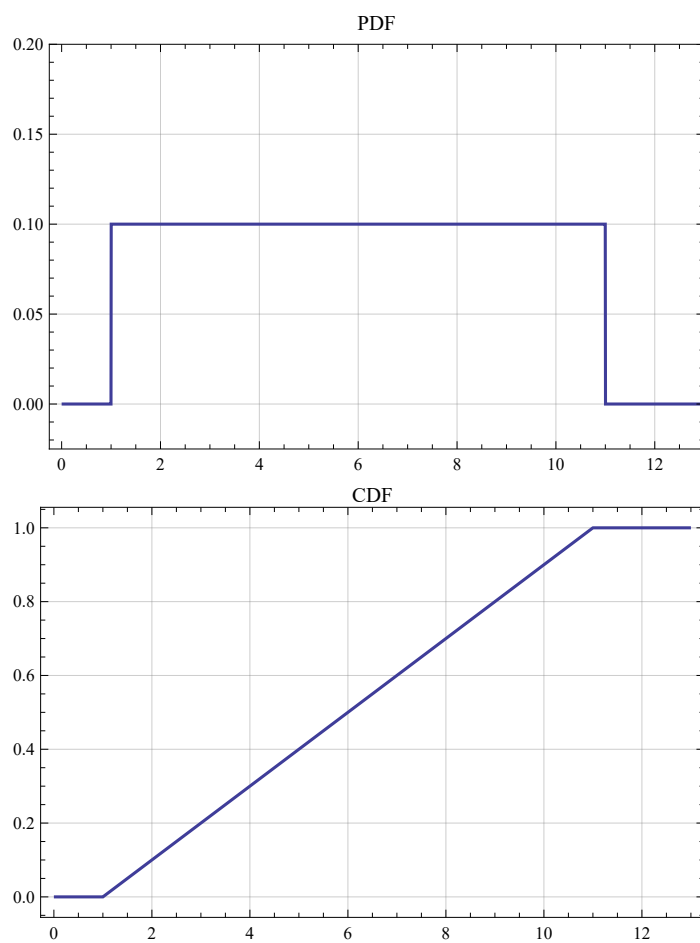
```

Out[]=



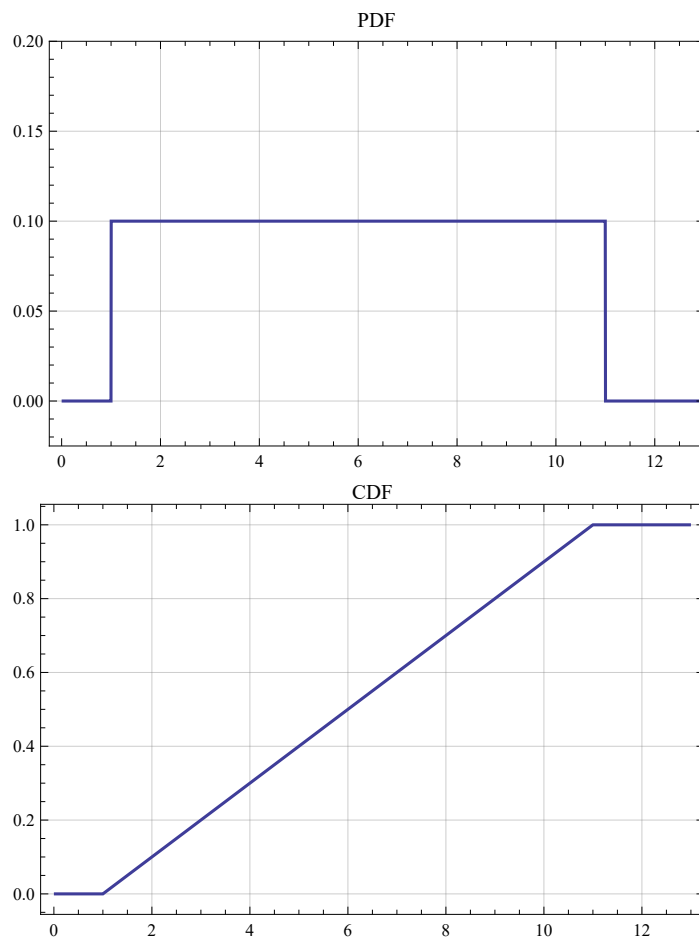
```
In[ ]:= Grid[{{pdf}, {cdf}}]
```

```
Out[ ]:=
```



```
In[ ]:= Insert[%40, Alignment → Right, 2]
```

```
Out[ ]:=
```



Mean Uniform Distribution

The aim is to replicate the demonstration than can be found here :

<https://en.wikibooks.org/wiki/Statistics/Distributions/Uniform>

`In[*]:= Mean[UniformDistribution[{a, b}]]`

`Out[*]=`

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

We derive the mean as follows.

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx$$

As the uniform distribution is 0 everywhere but $[a, b]$ we can restrict ourselves that interval

$$E[X] = \int_a^b \frac{1}{b-a} x dx$$

$$E[X] = \frac{1}{(b-a)} \frac{1}{2} x^2 \Big|_a^b$$

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

$$E[X] = \frac{b+a}{2}$$

Variance Uniform Distribution

The aim is to replicate the demonstration than can be found here :

https://en.wikibooks.org/wiki/Statistics/Distributions/Discrete_Uniform

In[]:=

Variance[UniformDistribution[{a, b}]]

Out[]:=

$$\frac{1}{12} (-a + b)^2$$

Variance [[edit](#) | [edit source](#)]

We use the following formula for the variance.

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

$$\text{Var}(X) = \left[\int_{-\infty}^{\infty} f(x) \cdot x^2 dx \right] - \left(\frac{b+a}{2} \right)^2$$

$$\text{Var}(X) = \left[\int_a^b \frac{1}{b-a} x^2 dx \right] - \frac{(b+a)^2}{4}$$

$$\text{Var}(X) = \frac{1}{b-a} \frac{1}{3} x^3 \Big|_a^b - \frac{(b+a)^2}{4}$$

$$\text{Var}(X) = \frac{1}{3(b-a)} [b^3 - a^3] - \frac{(b+a)^2}{4}$$

$$\text{Var}(X) = \frac{4(b^3 - a^3) - 3(b+a)^2(b-a)}{12(b-a)}$$

$$\text{Var}(X) = \frac{(b-a)^3}{12(b-a)}$$

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

Exponential Distribution

In[*]:= **exp = ExponentialDistribution[α]**

Out[*]=
ExponentialDistribution[α]

In[*]:= **PDF[exp, x]**

Out[*]=

$$\begin{cases} e^{-x\alpha} \alpha & x \geq 0 \\ 0 & \text{True} \end{cases}$$

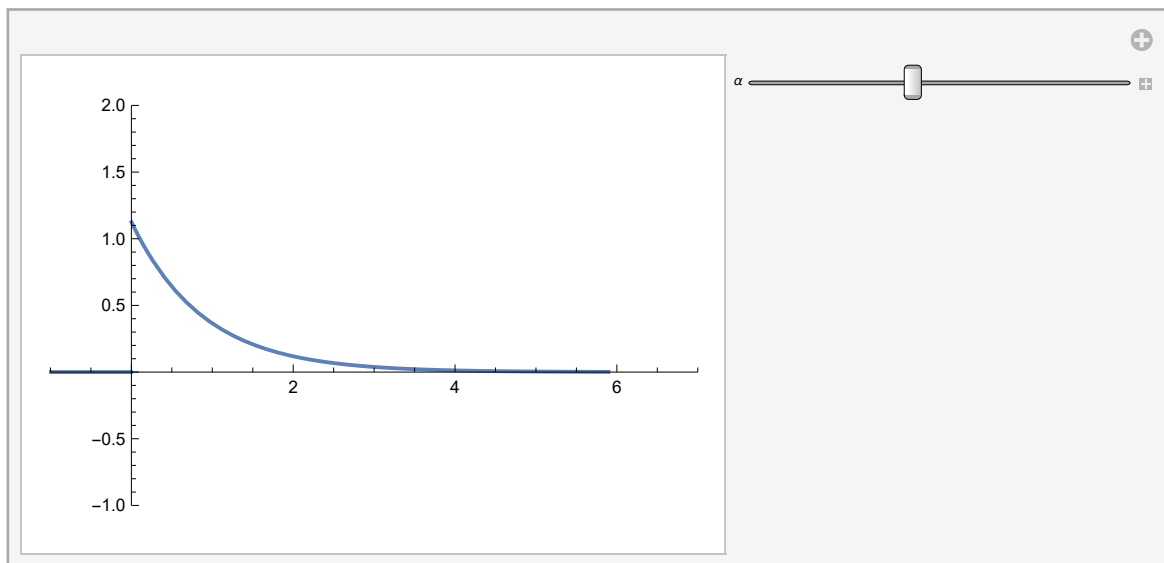
In[*]:= **PiecewiseExpand[Piecewise[{{ $e^{-x\alpha} \alpha$, $x \geq 0$ }}, 0]]**

Out[*]=

$$\begin{cases} e^{-x\alpha} \alpha & x \geq 0 \\ 0 & \text{True} \end{cases}$$

In[*]:= **Manipulate[Plot[PiecewiseExpand[Piecewise[{{ $e^{-x\alpha} \alpha$, $x \geq 0$ }}, 0]],
{x, -1, 5.89588}, PlotRange -> {{-1, 7}, {-1, 2}}], { α , 0.5, 2}]**

Out[*]=



`In[*]:= CDF[exp, x]`

`Out[*]=`

$$\begin{cases} 1 - e^{-x^\alpha} & x \geq 0 \\ 0 & \text{True} \end{cases}$$

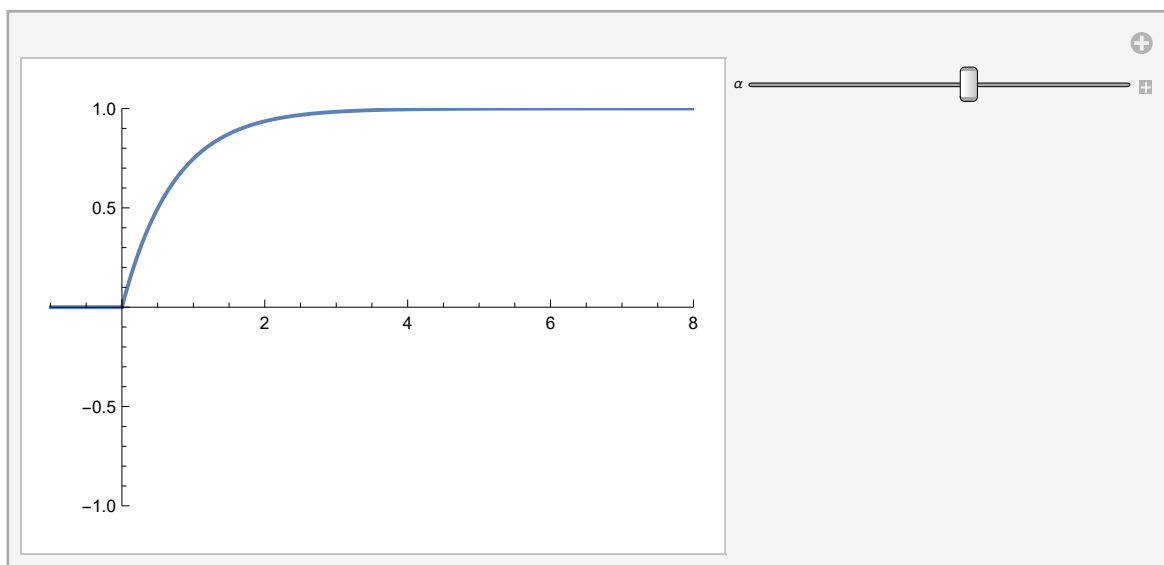
`In[*]:= PiecewiseExpand[Piecewise[{{1 - e-xα, x ≥ 0}}, 0]]`

`Out[*]=`

$$\begin{cases} e^{-x^\alpha} (-1 + e^{x^\alpha}) & x \geq 0 \\ 0 & \text{True} \end{cases}$$

`In[*]:= Manipulate[Plot[PiecewiseExpand[Piecewise[{{1 - e-xα, x ≥ 0}}, 0]],
{x, -1, 10}, PlotRange → {{-1, 8}, {-1, 1}}, {α, 0.5, 2}]`

`Out[*]=`

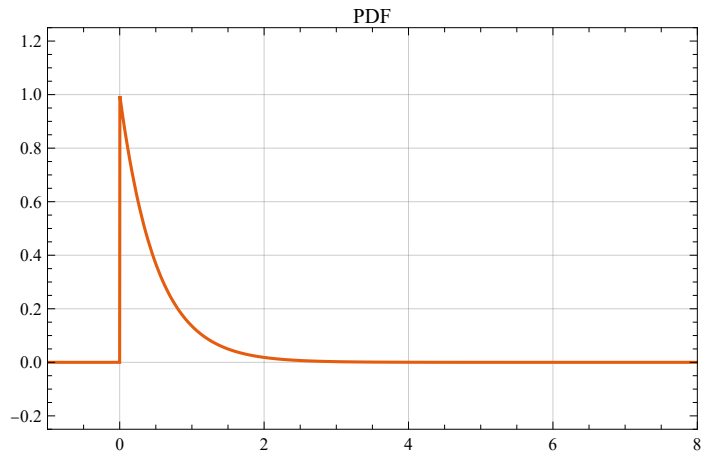


```

In[*]:= pdf = Plot[PiecewiseExpand[Piecewise[{{e-x2, x ≥ 0}}, 0]],
  {x, -1, 10}, PlotRange → {{-1, 8}, {-0.25, 1.25}},
  GridLines → Automatic, PlotLabel → "PDF",
  PlotTheme → "Scientific", ImageSize → Medium]

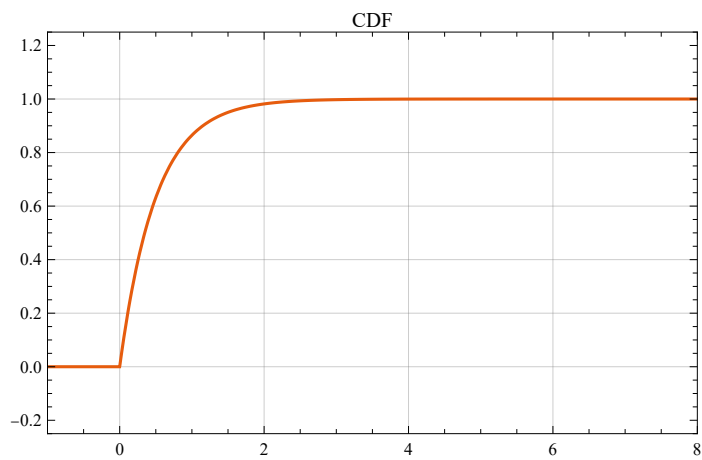
```

Out[*]=

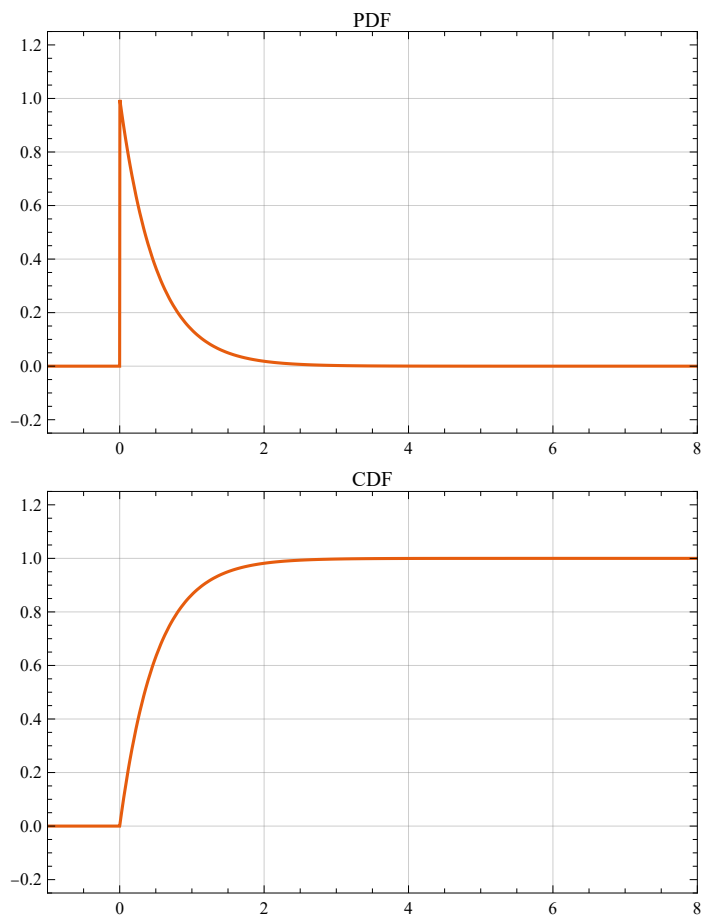


```
In[ ]:= cdf = Plot[PiecewiseExpand[Piecewise[{{1 - e-x2, x ≥ 0}}, 0]],
  {x, -1, 10}, PlotRange → {{-1, 8}, {-0.25, 1.25}}, GridLines → Automatic,
  PlotLabel → "CDF", PlotTheme → "Scientific", ImageSize → Medium]
```

Out[]:=



```
In[ ]:= Grid[{{pdf}, {cdf}}]  
Out[ ]=
```



Demonstration (Mean)

<https://en.wikibooks.org/wiki/Statistics/Distributions/Exponential>

`In[]:= Mean[ExponentialDistribution[λ]]`

`Out[]:=`

$$\frac{1}{\lambda}$$

Mean [[edit](#) | [edit source](#)]

We derive the mean as follows.

$$E[X] = \int_{-\infty}^{\infty} x \cdot f(x) dx$$

$$E[X] = \int_0^{\infty} x \lambda e^{-\lambda x} dx$$

$$E[X] = \int_0^{\infty} (-x)(-\lambda e^{-\lambda x}) dx$$

We will use **integration by parts** with $u=-x$ and $v=e^{-\lambda x}$. We see that $du=-1$ and $dv=-\lambda e^{-\lambda x}$.

$$E[X] = [-x \cdot e^{-\lambda x}]_0^{\infty} - \int_0^{\infty} (e^{-\lambda x})(-1) dx$$

$$E[X] = [0 - 0] + \left[\frac{-1}{\lambda} (e^{-\lambda x}) \right]_0^{\infty}$$

$$E[X] = \left[0 - \frac{-1}{\lambda} \right]$$

$$E[X] = \frac{1}{\lambda}$$

Demonstration (Variance)

<https://www.statlect.com/probability-distributions/exponential-distribution>

In[*]:= **Variance**[**ExponentialDistribution**[λ]]

Out[*]=

$$\frac{1}{\lambda^2}$$

Variance

The **variance** of an exponential random variable X is

$$\text{Var}[X] = \frac{1}{\lambda^2}$$

Proof

It can be derived thanks to the usual **variance formula** ($\text{Var}[X] = \text{E}[X^2] - \text{E}[X]^2$):

$$\begin{aligned} \text{E}[X^2] &= \int_0^{\infty} x^2 \lambda \exp(-\lambda x) dx \\ &= \left[-x^2 \exp(-\lambda x) \right]_0^{\infty} + \int_0^{\infty} 2x \exp(-\lambda x) dx \quad (\text{integrating by parts}) \\ &= (0 - 0) + \left[-\frac{2}{\lambda} x \exp(-\lambda x) \right]_0^{\infty} + \frac{2}{\lambda} \int_0^{\infty} \exp(-\lambda x) dx \quad (\text{integrating by parts again}) \\ &= (0 - 0) + \frac{2}{\lambda} \left[-\frac{1}{\lambda} \exp(-\lambda x) \right]_0^{\infty} \\ &= \frac{2}{\lambda^2} \\ \text{E}[X]^2 &= \left(\frac{1}{\lambda} \right)^2 = \frac{1}{\lambda^2} \\ \text{Var}[X] &= \text{E}[X^2] - \text{E}[X]^2 = \frac{2}{\lambda^2} - \frac{1}{\lambda^2} = \frac{1}{\lambda^2} \end{aligned}$$

In[*]:= **Integrate**[$(\lambda \text{Power}[x, 2]) / \text{Power}[E, (\lambda x)]$], {x, 0, Infinity}]

Out[*]=

$$\frac{2}{\lambda^2} \text{ if } \text{Re}[\lambda] > 0$$

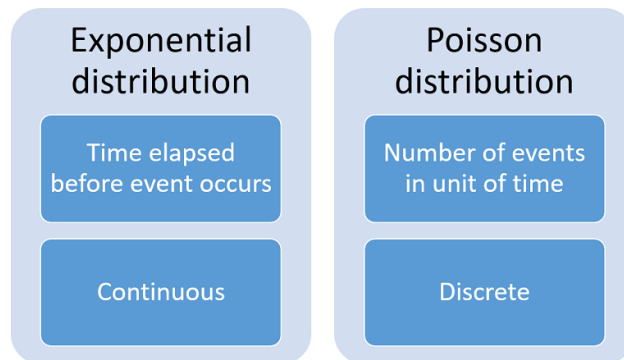
How the distribution is used

The exponential distribution is often used to answer in probabilistic terms questions such as:

- How much time will elapse before an earthquake occurs in a given region?
- How long do we need to wait until a customer enters our shop?
- How long will it take before a call center receives the next phone call?
- How long will a piece of machinery work without breaking down?

All these questions concern the time we need to wait before a given event occurs.

If this waiting time is unknown, it is often appropriate to think of it as a *random variable* having an exponential distribution.



Normal Distribution

In[]:= PDF[NormalDistribution[μ , σ]]

Out[]:=

$$\text{Function}\left[x, \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi}\sigma}\right]$$

In[]:= d = NormalDistribution[μ , σ]

Out[]:=

NormalDistribution[μ , σ]

In[]:= PDF[d, x]

Out[]:=

$$\frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi}\sigma}$$

In[]:= PDF[d, x]

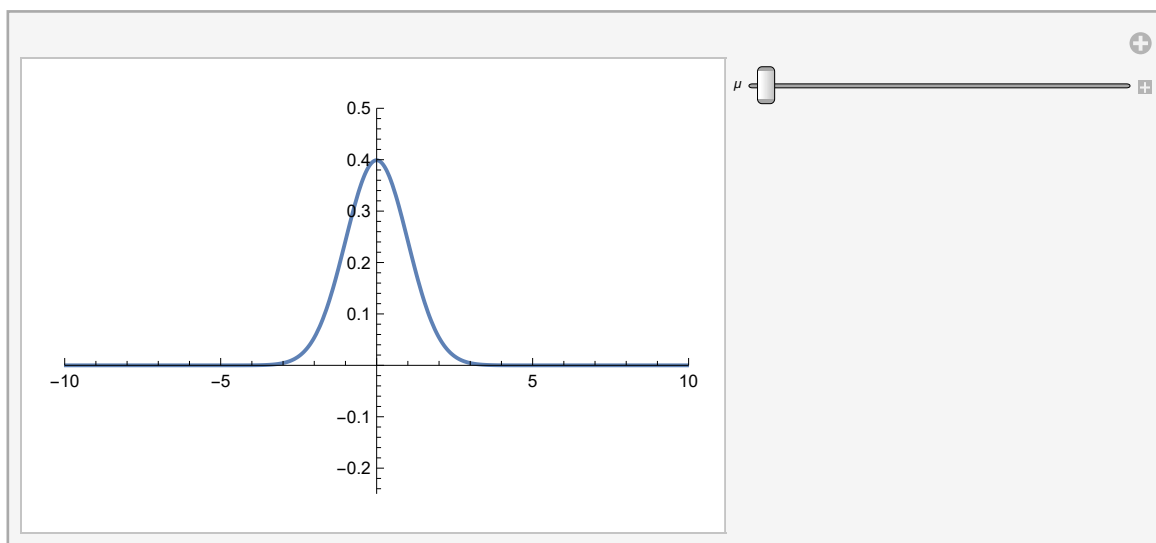
Out[]:=

$$\frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi}\sigma}$$

In[]:= Manipulate[

Plot $\left[\frac{e^{-\frac{(x-\mu)^2}{2\cdot 1^2}}}{\sqrt{2\pi}\cdot 1}, \{x, -10, 10\}, \text{PlotRange} \rightarrow \{\{-10, 10\}, \{-0.25, 0.5\}\}\right], \{\mu, 0, 5\}$

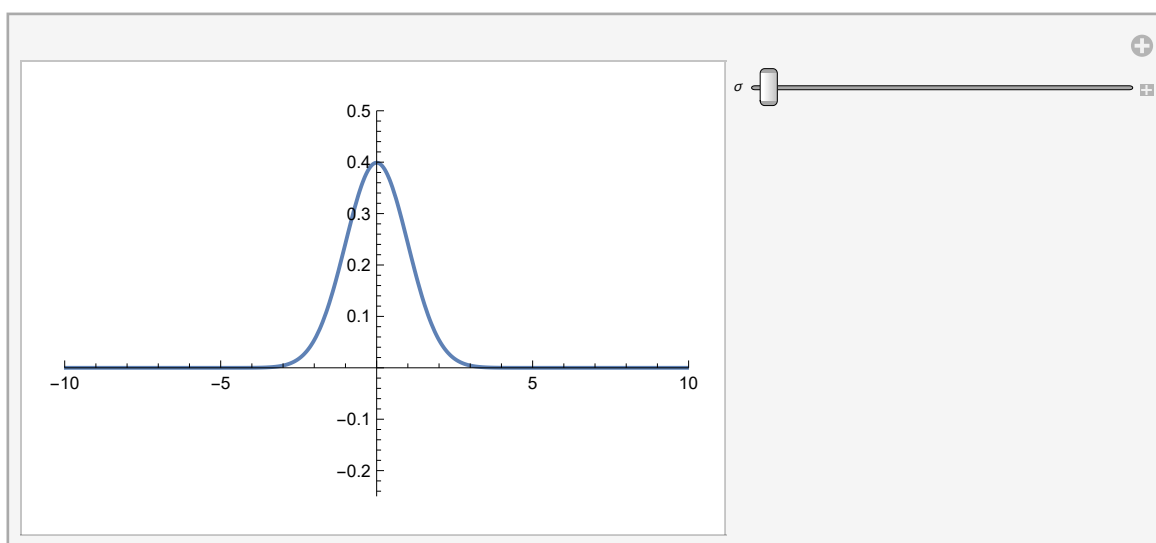
Out[]:=



In[]:= Manipulate[

Plot $\left[\frac{e^{-\frac{(x-0)^2}{2\sigma^2}}}{\sqrt{2\pi}\sigma}, \{x, -10, 10\}, \text{PlotRange} \rightarrow \{\{-10, 10\}, \{-0.25, 0.5\}\}\right], \{\sigma, 1, 5\}$

Out[]:=

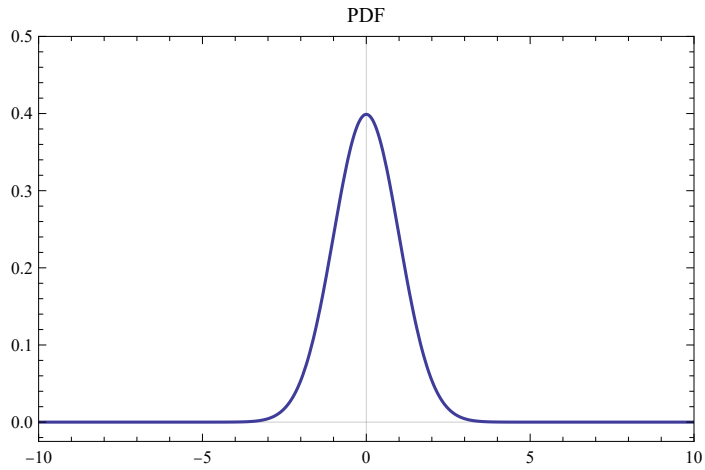


```

In[ ]:= pdf = Plot[ $\frac{e^{-\frac{(x-0)^2}{2 \cdot 1^2}}}{\sqrt{2 \pi} \cdot 1}$ , {x, -10, 10}, PlotLabel -> "PDF", PlotStyle -> ColorData[1, 1],
  PlotTheme -> "Scientific", ImageSize -> Medium, PlotRange -> {{-10, 10}, {-0.025, 0.5}}]

```

Out[]:=



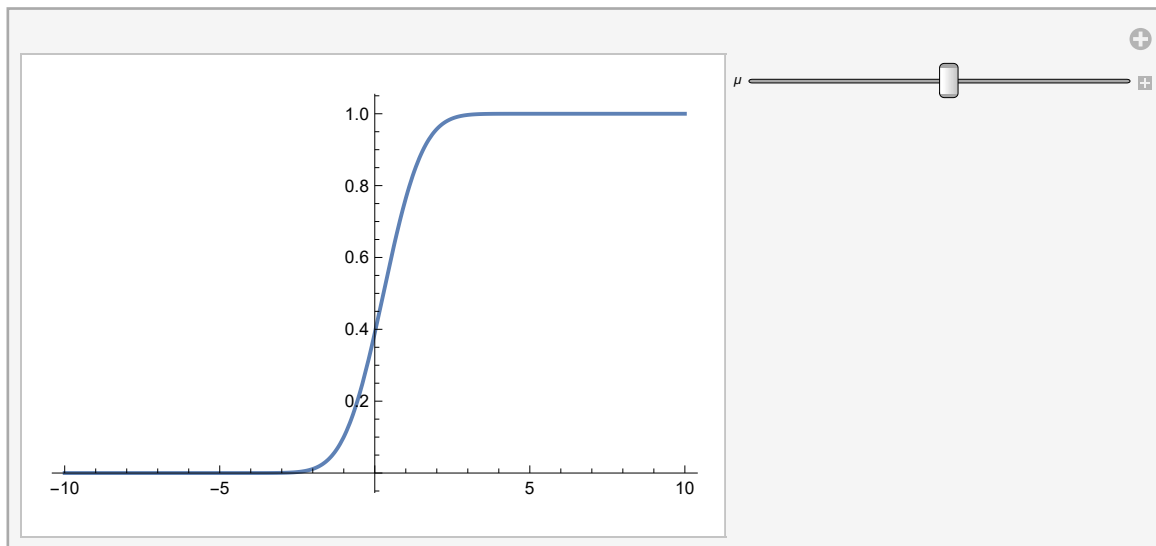
In[]:= CDF[d, x]

Out[]:=

$$\frac{1}{2} \operatorname{Erfc}\left[\frac{-x + \mu}{\sqrt{2} \sigma}\right]$$

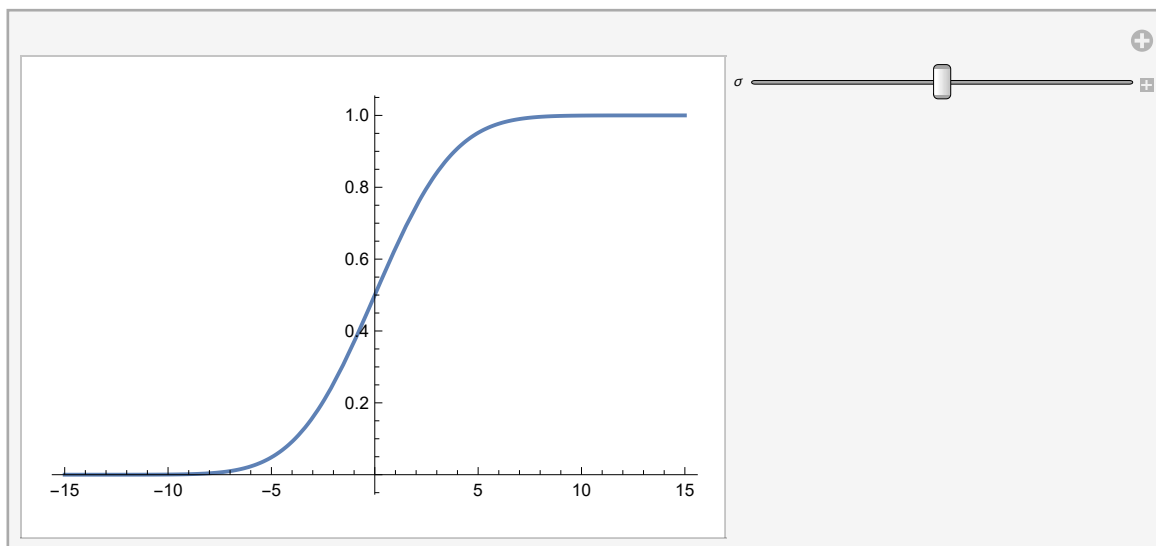
In[]:= Manipulate[Plot[$\frac{1}{2} \operatorname{Erfc}\left[\frac{-x + \mu}{\sqrt{2} \sigma}\right]$, {x, -10, 10}], {μ, -5.31623, 5.31623}]

Out[]:=



In[]:= Manipulate[Plot[$\frac{1}{2} \operatorname{Erfc}\left[\frac{-x + 0}{\sqrt{2} \sigma}\right]$, {x, -15, 15}], {σ, 1, 5}]

Out[]:=

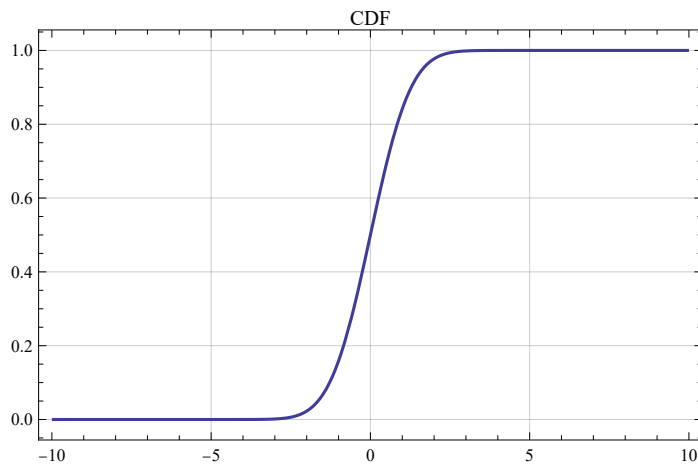


```

In[ ]:= cdf = Plot[ $\frac{1}{2} \operatorname{Erfc}\left[\frac{-x+\theta}{\sqrt{2}}$ ], {x, -10, 10}, PlotStyle -> ColorData[1, 1], PlotLabel -> "CDF",
GridLines -> Automatic, PlotTheme -> "Scientific", ImageSize -> Medium]

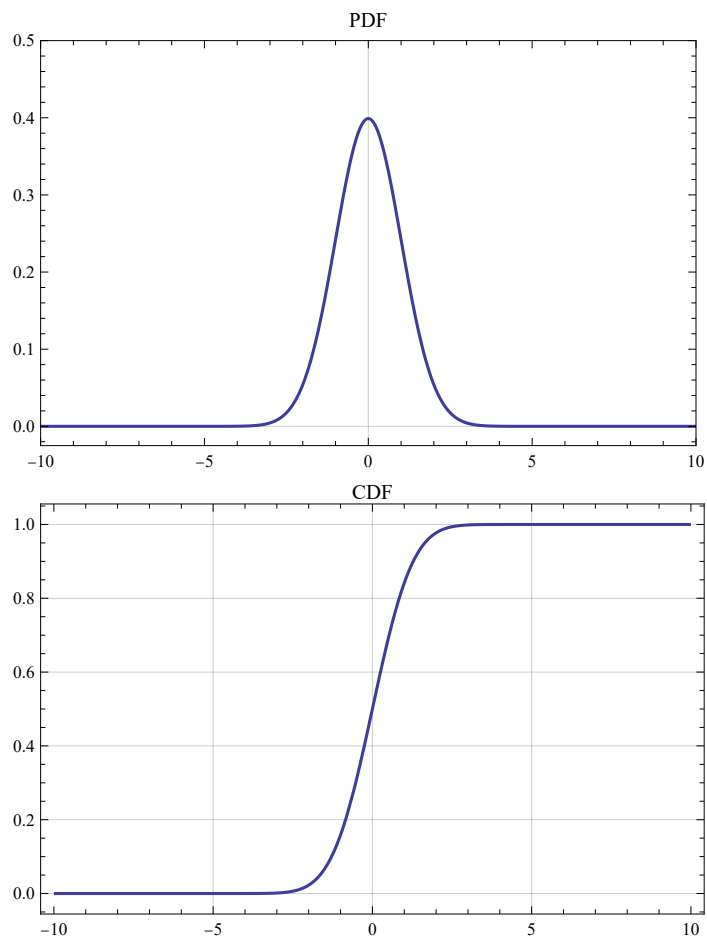
```

Out[]=



```
In[ ]:= Grid[{{pdf}, {cdf}}]
```

Out[]=



Demonstration (Mean)

(* (1) Write the expression for the Mean: $\sum_{i=0}^n f(x_i) \times (x_i)$ *)

$$\begin{aligned}
 E[X] &= \int_{-\infty}^{\infty} xf_X(x)dx \\
 &= (2\pi)^{-1/2} \int_{-\infty}^{\infty} x \exp\left(-\frac{1}{2}x^2\right)dx \\
 &= (2\pi)^{-1/2} \int_{-\infty}^0 x \exp\left(-\frac{1}{2}x^2\right)dx + (2\pi)^{-1/2} \int_0^{\infty} x \exp\left(-\frac{1}{2}x^2\right)dx \\
 &= (2\pi)^{-1/2} \left[-\exp\left(-\frac{1}{2}x^2\right)\right]_{-\infty}^0 + (2\pi)^{-1/2} \left[-\exp\left(-\frac{1}{2}x^2\right)\right]_0^{\infty} \\
 &= (2\pi)^{-1/2}[-1 + 0] + (2\pi)^{-1/2}[0 + 1] \\
 &= (2\pi)^{-1/2} - (2\pi)^{-1/2} \\
 &= 0
 \end{aligned}$$

<https://www.statlect.com/probability-distributions/normal-distribution>

Demonstration (Variance)

(* (1) Write the expression for the "Squared Mean": $\sum_{i=0}^n f(x_i) \times (x_i)^2$ *)

$$\begin{aligned}
 E[X^2] &= \int_{-\infty}^{\infty} x^2 f_X(x) dx \\
 &= (2\pi)^{-1/2} \int_{-\infty}^{\infty} x^2 \exp\left(-\frac{1}{2}x^2\right) dx \\
 &= (2\pi)^{-1/2} \left\{ \int_{-\infty}^0 x \exp\left(-\frac{1}{2}x^2\right) dx + \int_0^{\infty} x \exp\left(-\frac{1}{2}x^2\right) dx \right\} \\
 &= (2\pi)^{-1/2} \left\{ \left[-x \exp\left(-\frac{1}{2}x^2\right) \right]_{-\infty}^0 + \int_{-\infty}^0 \exp\left(-\frac{1}{2}x^2\right) dx + \left[-x \exp\left(-\frac{1}{2}x^2\right) \right]_0^{\infty} \right. \\
 &\quad \left. + \int_0^{\infty} \exp\left(-\frac{1}{2}x^2\right) dx \right\} \quad (\text{integrating by parts}) \\
 &= (2\pi)^{-1/2} \left\{ (0 - 0) + (0 - 0) + \int_{-\infty}^0 \exp\left(-\frac{1}{2}x^2\right) dx + \int_0^{\infty} \exp\left(-\frac{1}{2}x^2\right) dx \right\} \\
 &= (2\pi)^{-1/2} \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2}x^2\right) dx \\
 &= \int_{-\infty}^{\infty} f_X(x) dx = 1 \quad (\text{the integral of a pdf over its support is equal to 1}) \\
 E[X]^2 &= 0^2 = 0 \\
 \text{Var}[X] &= E[X^2] - E[X]^2 = 1 - 0 = 1
 \end{aligned}$$

<https://www.statlect.com/probability-distributions/normal-distribution>

Properties

<https://www.ztable.net/>

Properties

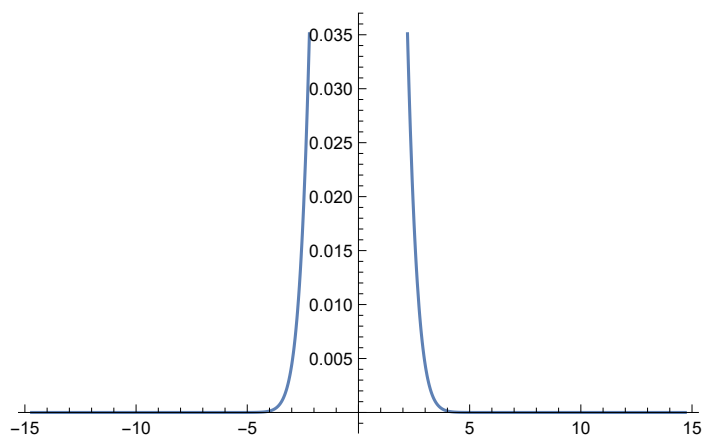
`In[*]:= PDF[NormalDistribution[0, 1], x]`

`Out[*]=`

$$\frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}}$$

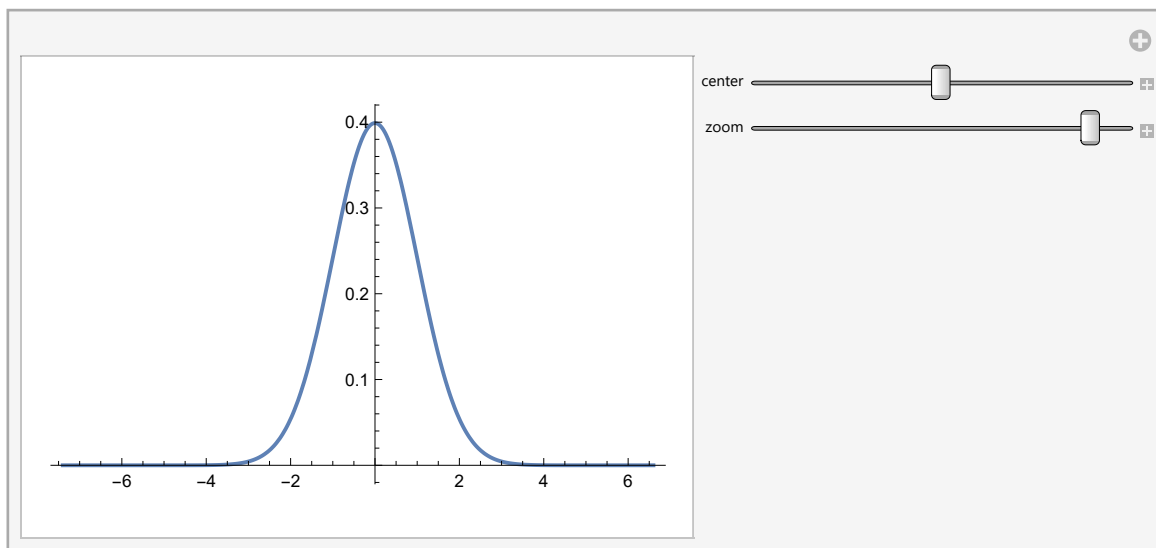
`In[*]:= Plot[$\frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}}$, {x, -14.6969, 14.6969}]`

`Out[*]=`



`In[*]:=`  **interactive plot %12**

`Out[*]=`



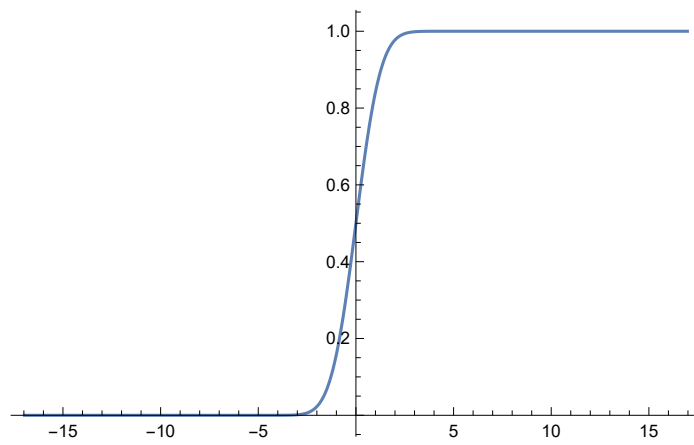
`In[*]:= CDF[NormalDistribution[0, 1], x]`

`Out[*]=`

$$\frac{1}{2} \operatorname{Erfc}\left[-\frac{x}{\sqrt{2}}\right]$$

```
In[*]:= Plot[ $\frac{1}{2} \operatorname{Erfc}\left[-\frac{x}{\sqrt{2}}\right]$ , {x, -16.9706, 16.9706}]
```

Out[*]=



```
In[*]:= CDF[NormalDistribution[0, 1], 1]
```

```
Out[*]=
```

$$\frac{1}{2} \operatorname{Erfc}\left[-\frac{1}{\sqrt{2}}\right]$$

```
In[*]:= N\left[\frac{1}{2} \operatorname{Erfc}\left[-\frac{1}{\sqrt{2}}\right]\right]
```

```
Out[*]=
```

```
0.841345
```

```
In[*]:= CDF[NormalDistribution[0, 1], 1.5]
```

```
Out[*]=
```

```
0.933193
```

```
In[*]:= CDF[NormalDistribution[0, 1], 1.54]
```

```
Out[*]=
```

```
0.93822
```

```
In[*]:= CDF[NormalDistribution[0, 1], -1.92]
```

```
Out[*]=
```

```
0.0274289
```

```
In[*]:= 1 - CDF[NormalDistribution[0, 1], 1.92]
```

```
Out[*]=
```

```
0.0274289
```

```
In[*]:= CDF[NormalDistribution[0, 1], 2] - CDF[NormalDistribution[0, 1], -1]
```

```
Out[*]=
```

$$-\frac{1}{2} \operatorname{Erfc}\left[\frac{1}{\sqrt{2}}\right] + \frac{1}{2} \operatorname{Erfc}\left[-\sqrt{2}\right]$$

```
In[*]:= N\left[-\frac{1}{2} \operatorname{Erfc}\left[\frac{1}{\sqrt{2}}\right] + \frac{1}{2} \operatorname{Erfc}\left[-\sqrt{2}\right]\right]
```

<https://www.wolframalpha.com/input?i=normal+distribution+calculator&assumption=%7B%22F%22%2C+%22NormalProbabilities%22%2C+%22mu%22%7D+-%3E%220%22&assumption=%7B%22F%22%2C+%22NormalProbabilities%22%2C+%22sigma%22%7D+-%3E%221%22&assumption=%7B%22F%22%2C+%22NormalProbabilities%22%2C+%22pr%22%7D+-%3E%220.95%22>