```
    using Markdown
```

1D Gaussians [10 pts]

Let X be a univariate random variable distributed according to a Gaussian distribution with mean μ and variance σ^2

```
md"""
# 1D Gaussians [10 pts]
Let $X$ be a univariate random variable distributed according to a Gaussian distribution with mean $\mu$ and variance $\sigma^2$
"""
```

Can the probability density function (pdf) of X ever take values greater than 1?

Answer:

```
md"""
### Can the probability density function (pdf) of $X$ ever take values greater than $1$?
Answer:
"""
```

Write the expression for the pdf of a univariate gaussian:

Answer:

Write the code for the function that computes the pdf at x.

gaussian_pdf (generic function with 1 method)

```
function gaussian_pdf(x; mean=0., variance=0.01)
#default variables mean and variance
#set with keyword arguments
return #TODO: implement pdf at x
end
```

Test your implementation against a standard implementation

E.g. from a library, e.g. Distributions.jl.

```
• Enter cell code...

• using Test

• using Distributions: pdf, Normal
• # Note Normal uses N(mean, stddev) for parameters

Some tests did not pass: 0 passed, 0 failed, 2 errored, 0 broken.

1. finish(::Test.DefaultTestSet) @ Test.jl:879
2. top-level scope @ Test.jl:1125
3. top-level scope @ [Local: 2]

• @testset "Implementation of Gaussian pdf" begin
• x = randn()
• @test gaussian_pdf(x) ≈ pdf.(Normal(0.,sqrt(0.01)),x)
• # ≈ is syntax sugar for isapprox, typed with '\approx <TAB>'
• # or use the full function, like below
• @test isapprox(gaussian_pdf(x,mean=10., variance=1) , pdf.(Normal(10., sqrt(1)),x))
• end
```

What is the value of the pdf at x = 0? What is probability that x = 0?

```
md"""### What is the value of the pdf at $x=0$? What is probability that $x=0$?"""
```

```
• Enter cell code...
```

A Write the transformation that takes $x \sim \mathcal{N}(0., 1.)$ to $z \sim \mathcal{N}(\mu, \sigma^2)$

A Gaussian with mean μ and variance σ^2 can be written as a simple transformation of the standard Gaussian with mean 0, and variance 1.

Answer:

```
md"""
### A Write the transformation that takes $x \sim \mathcal{N}(0.,1.)$ to $z \sim \mathcal{N}(\mu, \sigma^2)$
A Gaussian with mean $\mu$ and variance $\sigma^2$ can be written as a simple transformation of the standard Gaussian with mean $0.$ and variance $1.$.
Answer:
"""
```

Enter cell code...

Write a code to sample from $\mathcal{N}(\mu, \sigma^2)$

Implement function returning n independent samples from $\mathcal{N}(\mu, \sigma^2)$ by transforming n samples from $\mathcal{N}(0., 1.)$

```
function sample_gaussian(n; mean=0., variance=0.01)
  # n samples from standard gaussian
  x = #TODO

# TODO: transform x to sample z from N(mean, variance)
  z =
  return z
end;
```

Test your implementation by computing statistics on the samples

```
• using Statistics: mean, var

• @testset "Numerically testing Gaussian Sample Statistics" begin
• #TODO: choose some values of mean and variance to test
• #TODO: Sample 100000 samples with sample_gaussian
• #TODO: Use 'mean' and 'var' to compute statistics
• #TODO: test statistics against true values
• # hint: use isapprox with keyword argument atol=1e-2
• end;
```

Plot pdf and normalized histogram of samples

Sample 10000 samples from a Gaussian with mean 10. and variance 2.0.

- 1. Plot the **normalized** histogram of these samples.
- 2. On the same axes plot! the pdf of this distribution.

Confirm that the histogram approximates the pdf.

(Note: with Plots.jl the function plot! will add to the existing axes.)

```
"""
### Plot pdf and normalized histogram of samples

Sample $10000$ samples from a Gaussian with mean $10.$ and variance $2.0$.

1. Plot the **normalized** 'histogram' of these samples.
2. On the same axes 'plot!' the pdf of this distribution.
Confirm that the histogram approximates the pdf.

(Note: with 'Plots.jl' the function 'plot!' will add to the existing axes.)
"""
```

using Plots

- #histogram() #TODO#plot!() #TODO