

The logistic model describes the growth of biological populations. The standard deterministic model is described by the ordinary differential equation

$$\frac{dP_t}{dt} = rP_t \left(1 - \frac{P_t}{K}\right)$$

where  $P_t$  denotes the population size at time  $t$ ,  $r$  is the growth rate and  $K$  is the carrying capacity, the maximum population size that the environment can sustain.

The solution of the deterministic equation is  $P_t = \frac{KP_0}{P_0 + (K - P_0)e^{-rt}}$

A stochastic logistic equation is described the SDE

$$dP_t = rP_t \left(1 - \frac{P_t}{K}\right) dt + \sigma P_t dB_t$$

where  $\sigma > 0$  is a parameter.

The solution to the logistic SDE is

$$P_t = \frac{P_0 K X_t}{K + P_0 r \int_0^t X_s ds}$$

where  $X_t = e^{\left(r - \frac{\sigma^2}{2}\right)t + \sigma B_t}$ , geometric brownian motion.

```
In [3]: using Plots,Distributions
        plotlyjs();
```

```
In [2]: p0 = 2; r = 0.06; k = 300;
        sigma_vals = [0.02,0.05,0.15];
```

```
In [4]: function sol(t)
        num = k*p0
        den = p0 + (k- p0)*exp(-r*t)
        return num/den
    end
```

```
Out[4]: sol (generic function with 1 method)
```

```
In [5]: tvals = range(start =0,stop= 200, length = 20000)
        y = [sol(t) for t in tvals];
```

```
In [6]: sigma = 0.02
        Xt = [exp(r - sigma^2/2)*t + sigma*rand(Normal(0,sqrt(t)))
              for t in tvals]
        numerator = p0*k*Xt;
```

cumsum(Xt[1:length(tvals)] .\* (tvals[2:length(tvals)] .- tvals[1:length(tvals)])) # it gives dimension does not match error

```
In [7]: n = 20_000
        t = 200
```

```
x = rand(Normal(0,sqrt(t/n)),n)
bm = [0;cumsum(x)]
steps = range(start = 0,stop = t, length = n+1)
integral = cumsum(bm[1:n].* (steps[2:n+1] .- steps[1:n]))
```

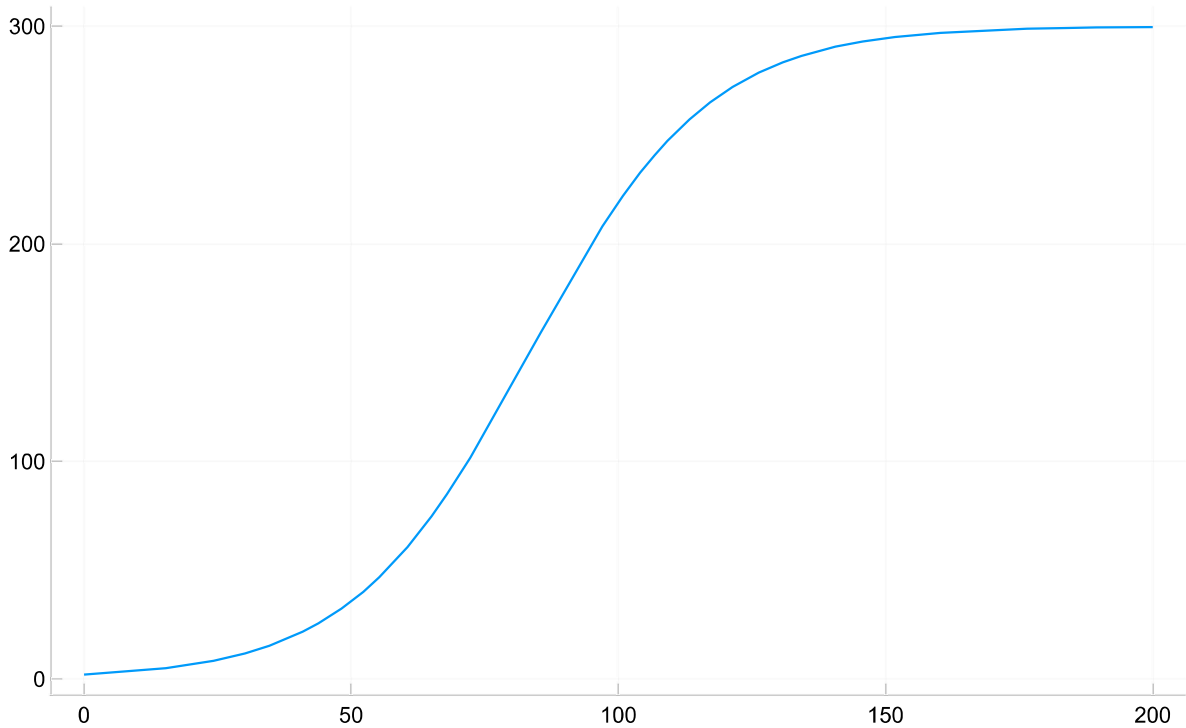
Out[7]: 20000-element Vector{Float64}:

```
0.0
0.0007150772187182679
0.00036007424481368863
-0.0001469485261921814
-0.00169877412205009
-0.0035997842878361906
-0.006052820372833492
-0.007132409254204369
-0.008113862204168416
-0.008812131587017836
-0.008455987964823828
-0.008818829059129916
-0.009164655893670584
⋮
-1020.1362494528001
-1020.2803223625731
-1020.4242557717726
-1020.5677170578697
-1020.7100680207005
-1020.8525894337878
-1020.9961703416889
-1021.1404093210344
-1021.2862463683978
-1021.4304185920092
-1021.575711540347
-1021.7201388789304
```

In [8]: `steps = range(start = 0, stop = 0.01, length = 100);`

In [9]: `plot(tvals,y,label =:none)`

Out[9]:



Let  $B_t$  is a standard brownian motion. Let  $G_0$  is a constant and  $G_t = G_0 e^{\mu t + \sigma B_t}$ . Then  $G_t$  is called geometric brownian motion. The expectation and variance of geometric brownian motion are:

- $E[G_t] = G_0 e^{t(\mu + \frac{\sigma^2}{2})}$
- $Var[G_t] = G_0^2 e^{2t(\mu + \frac{\sigma^2}{2})} (e^{\sigma^2 t} - 1)$

## Computation of Ito Integral:

Compute the integral  $\int_0^t B_s dB_s$ . We compute this integral by the following way:

$$\int_0^t B_s dB_s = \lim_{n \rightarrow \infty} \sum_{k=1}^n B_{t_{k-1}} (B_{t_k} - B_{t_{k-1}})$$

where we partition the interval  $[0, t]$  into  $n$  many parts that is  $0 = t_0 < t_1 < \dots < t_n = t$ . This sum will give us approximate integral to the Stochastic Integral. The exact integral will be  $\frac{1}{2} (B_t^2 - t)$ .

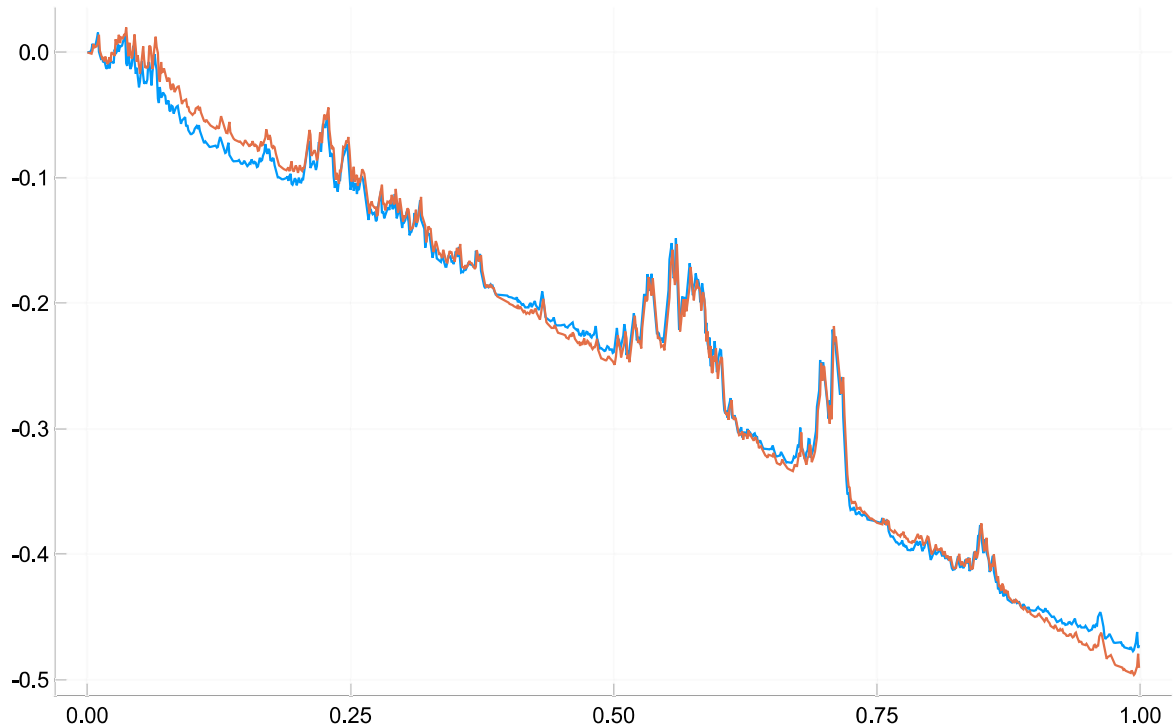
```
In [10]: n = 1000
t = 1
x = rand(Normal(0, sqrt(t/n)), n)
bm = [0; cumsum(x)]
steps = range(start = 0, stop = t, length = n+1)
```

Out[10]: 0.0:0.001:1.0

In [11]: `int = cumsum(bm[1:n].* (bm[2:n+1] .- bm[1:n]));`

In [12]: `plot(steps[1:n],int,label =:none) #,label="approx_integral",lw = 2,fg_legend = :none)`  
`plot!(steps[1:n],(0.5*(bm.^2 - steps))[1:n],label =:none)`

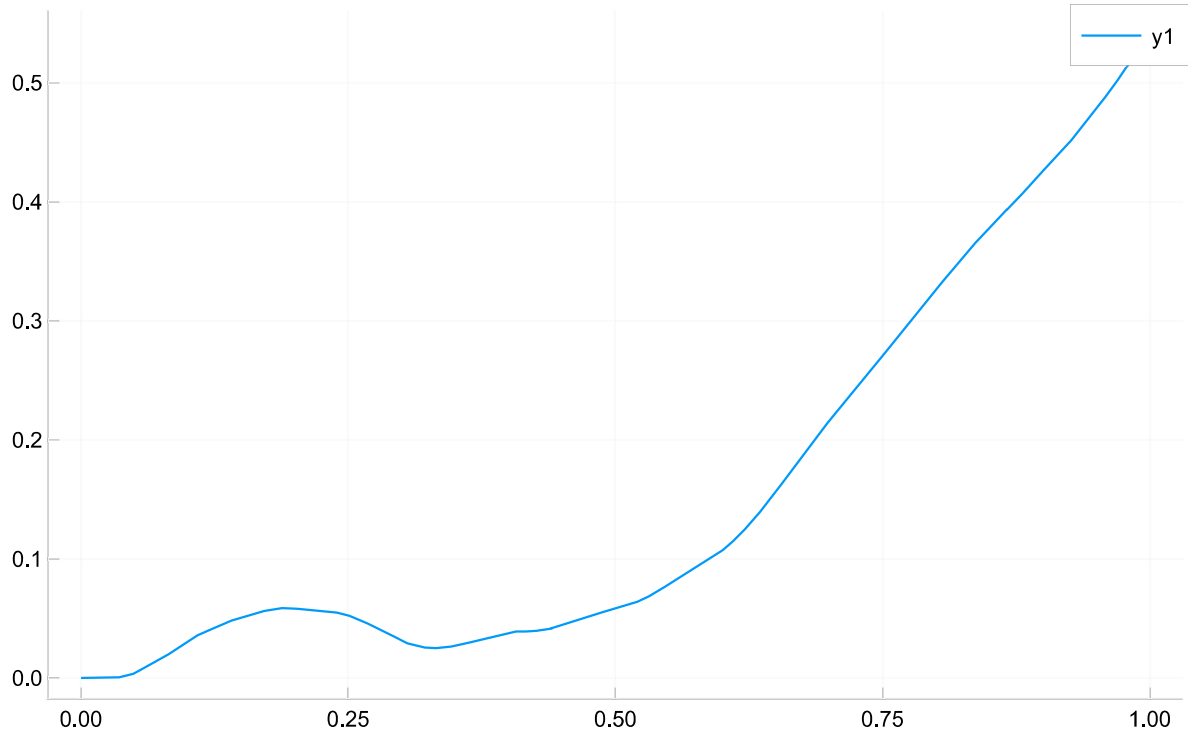
Out[12]:



Consider the integral  $\int_0^t B_s ds$ .

In [13]: `n = 10_00_000`  
`t = 1`  
`x = rand(Normal(0,sqrt(t/n)),n)`  
`bm = [0;cumsum(x)]`  
`steps = range(start = 0,stop = t, length = n+1)`  
`integral = cumsum(bm[1:n].* (steps[2:n+1] .- steps[1:n]))`  
`plot(steps[1:n],integral)`

Out[13]:

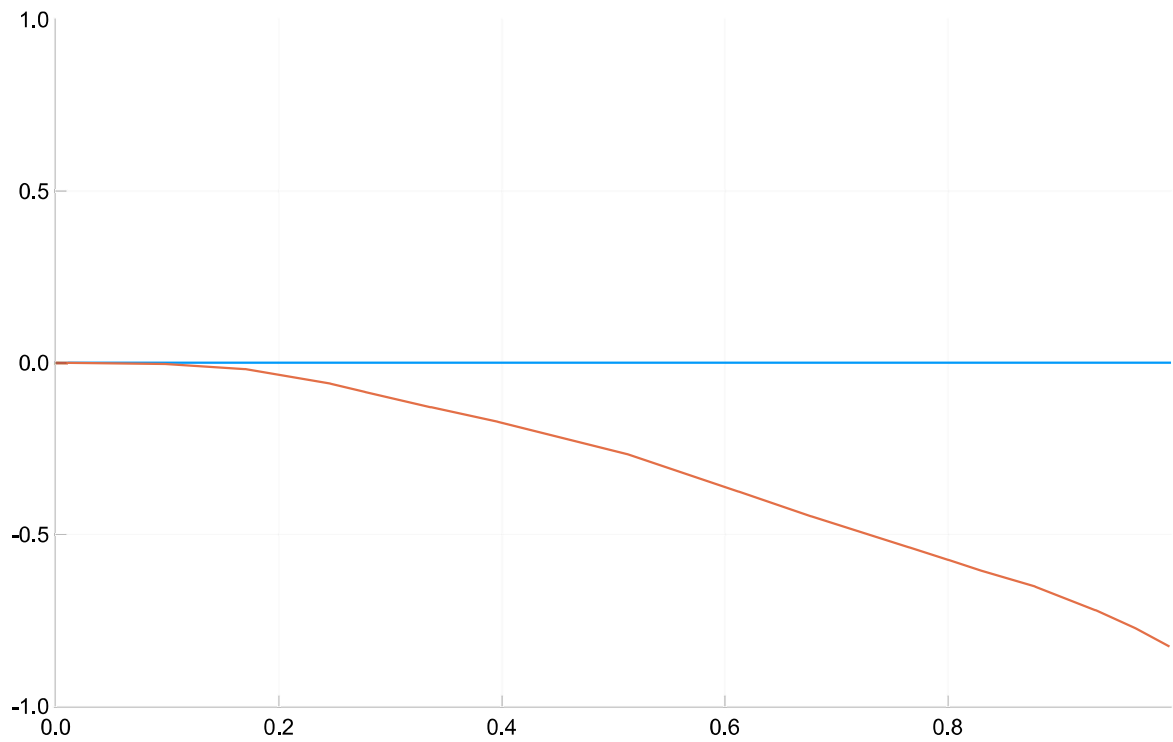
In [14]: `mean(integral)`

Out[14]: 0.15488185830666562

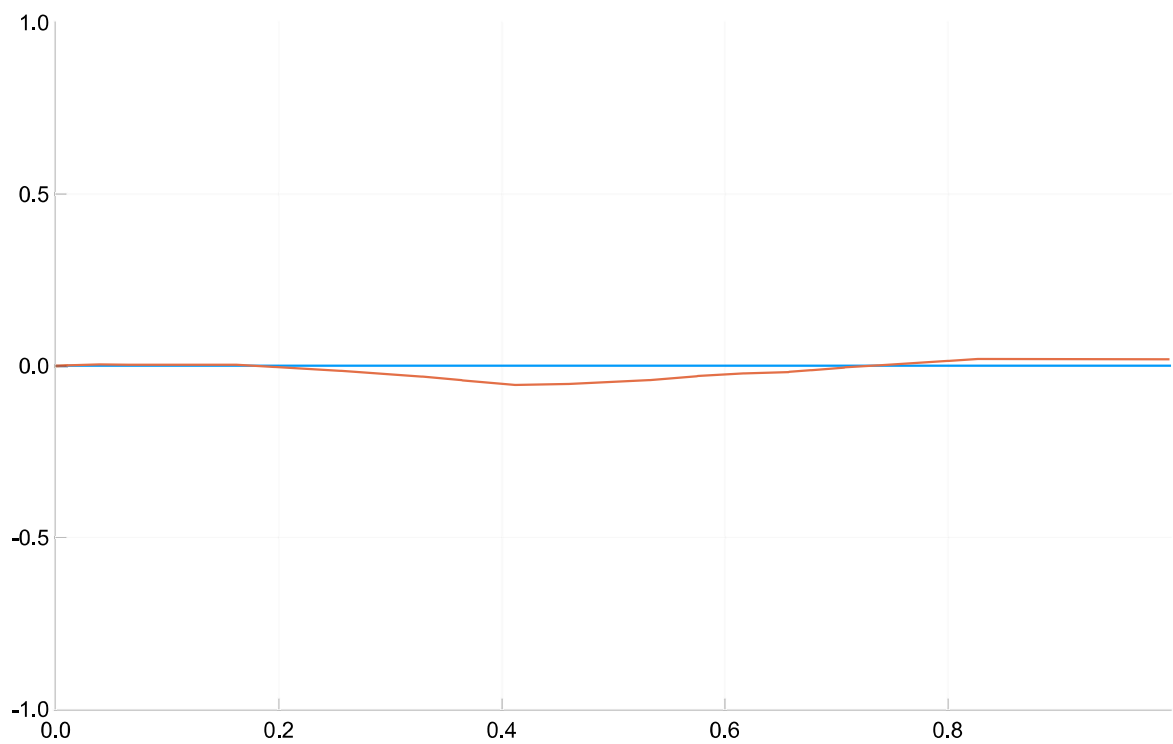
```

In [24]: for i in range(start = 1, step = 1, stop = 9)
          p = plot(xlims = (0,1), label = :none, ylims = (-1.0,1.0))
          hline!([0], label = :none)
          n = 1000
          t = 1
          x = rand(Normal(0, sqrt(t/n)), n)
          bm = [0; cumsum(x)]
          steps = range(start = 0, stop = t, length = n+1)
          integral = cumsum(bm[1:n].* (steps[2:n+1] .- steps[1:n]))
          p = plot!(steps[1:n], integral, label = :none)
          display(p)
          println("Area : ", mean(integral))
        end

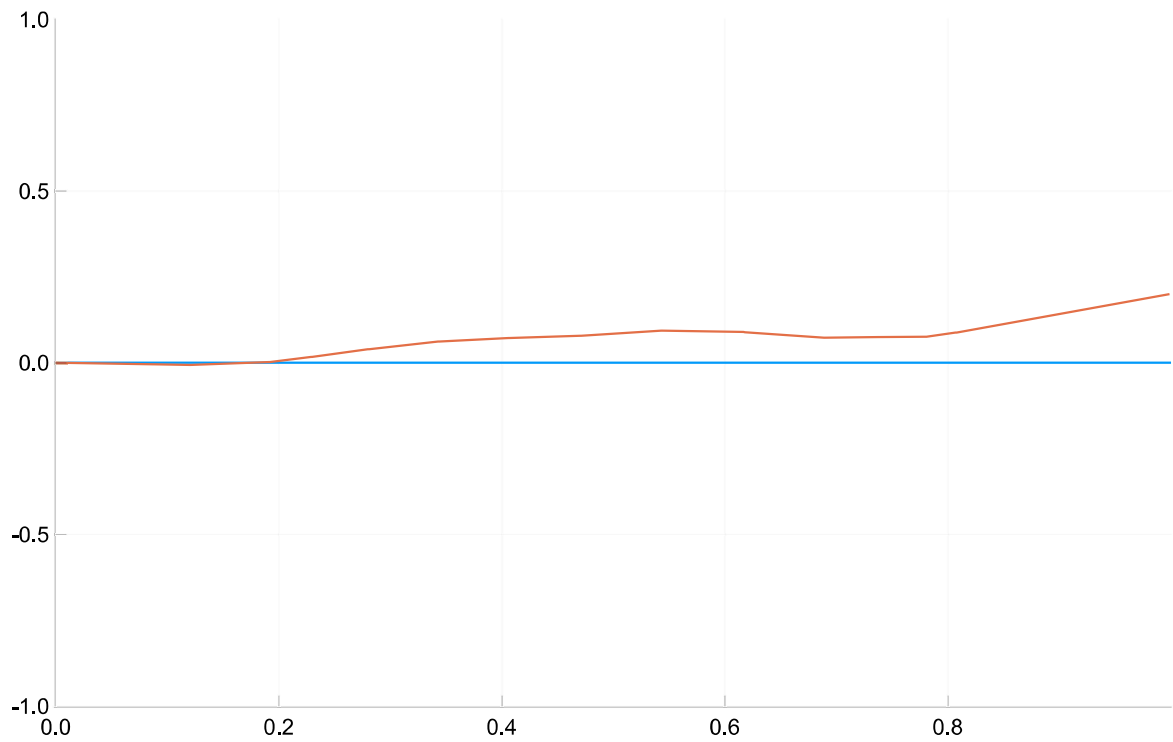
```



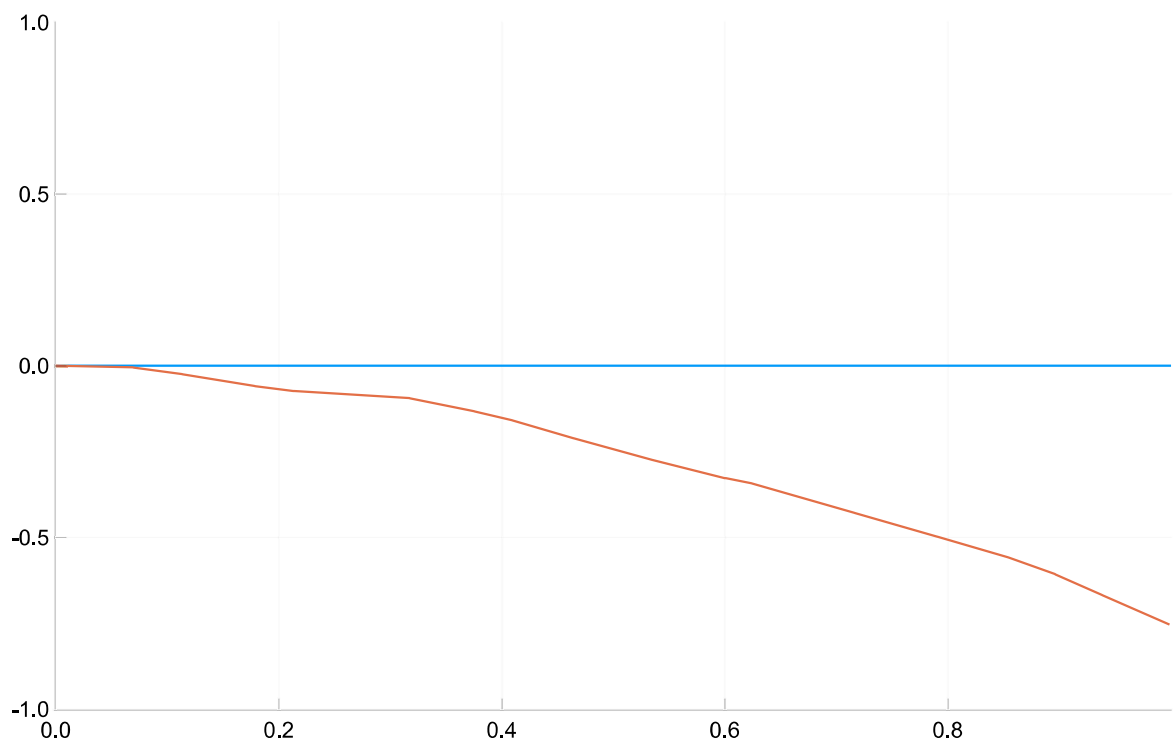
Area : -0.3044446120029065



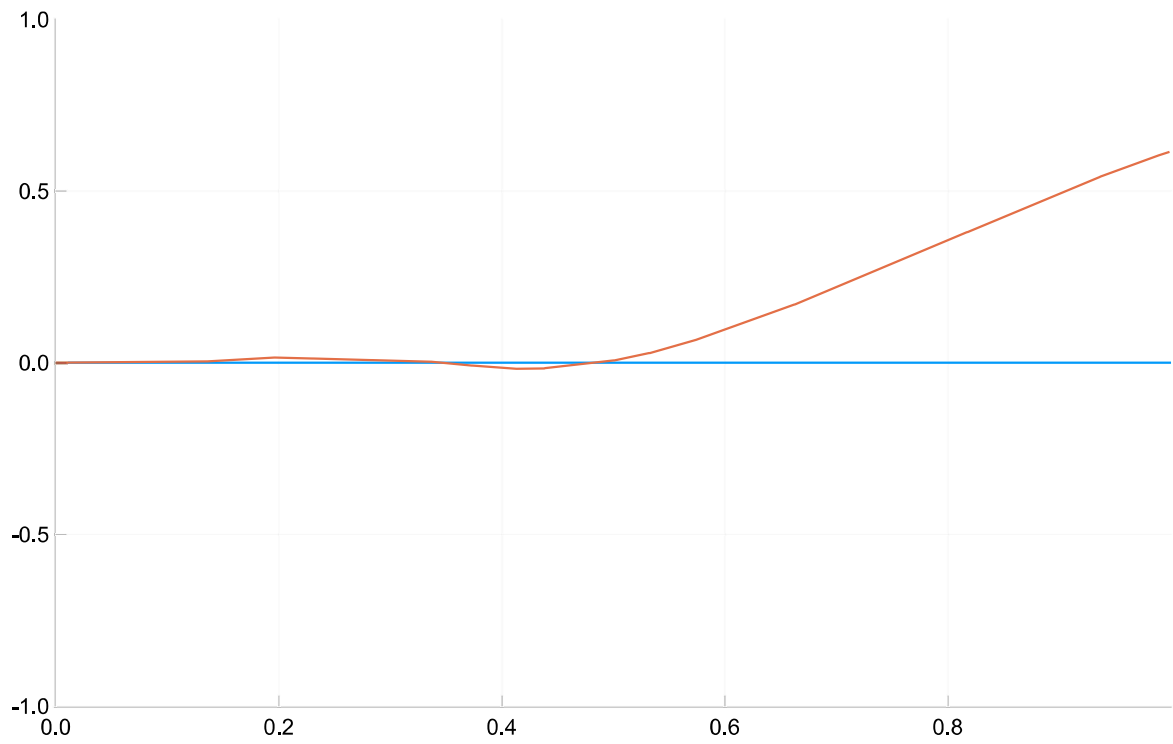
Area : -0.011310643774664415



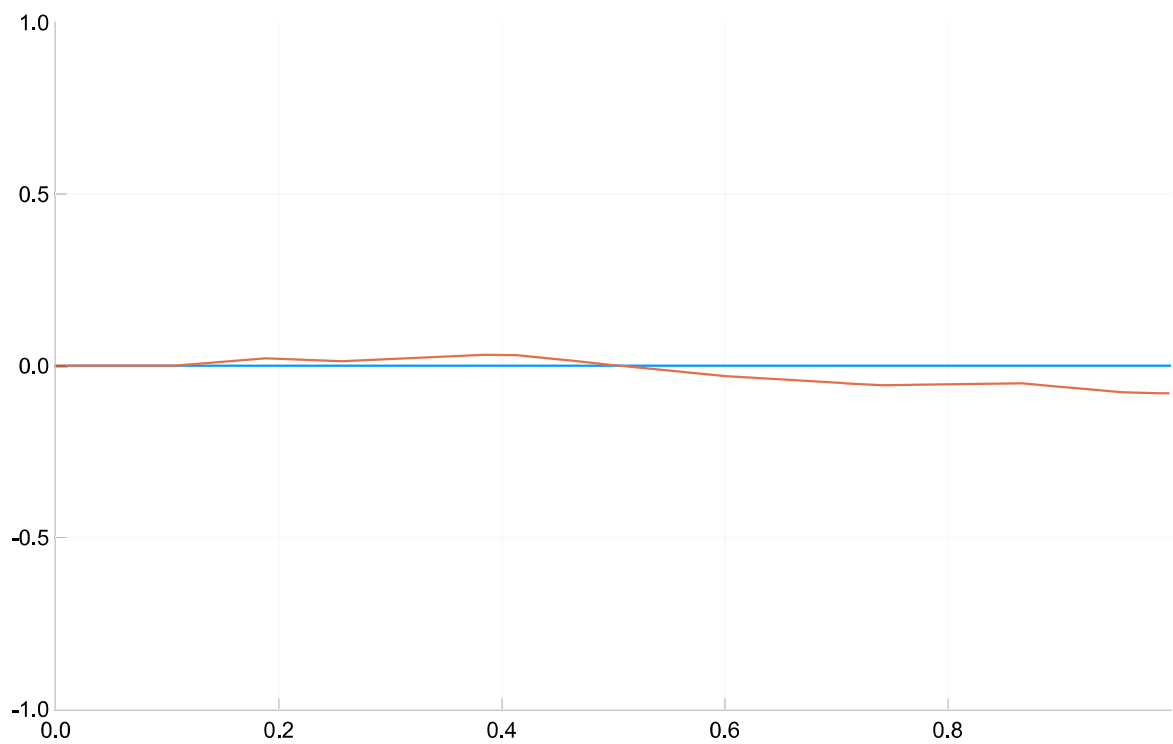
Area : 0.06922130678983564



Area : -0.2791050927104301

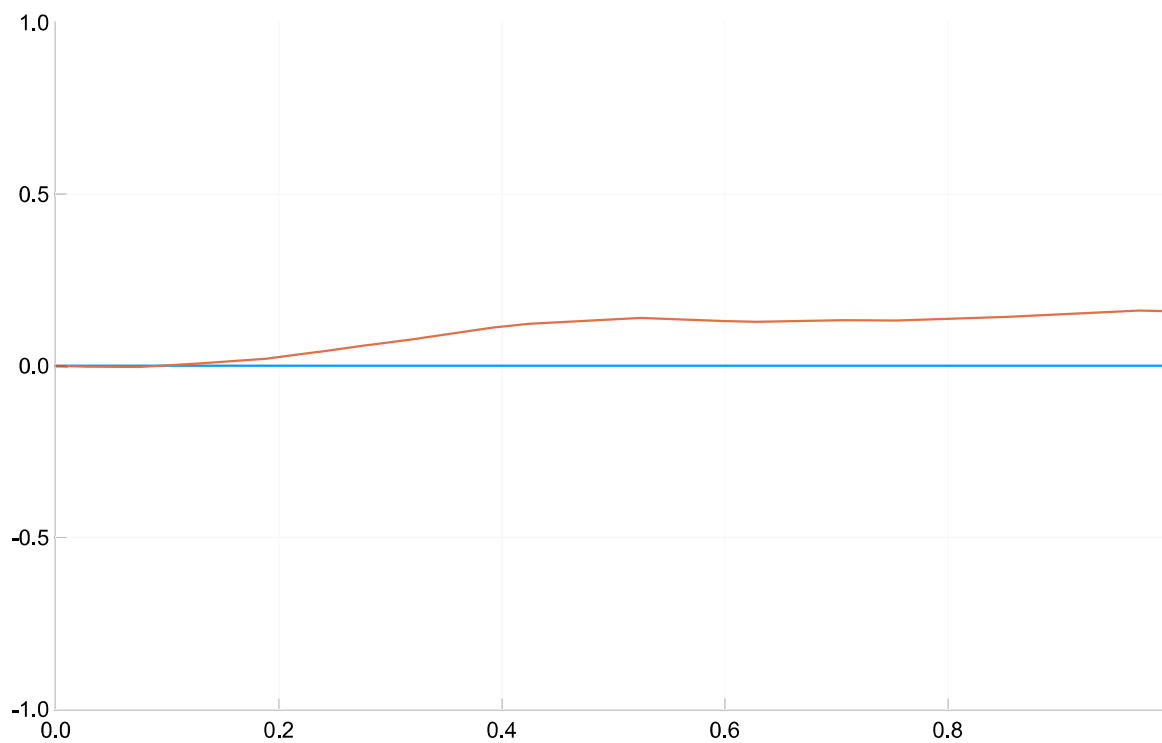


Area : 0.14751287719392545

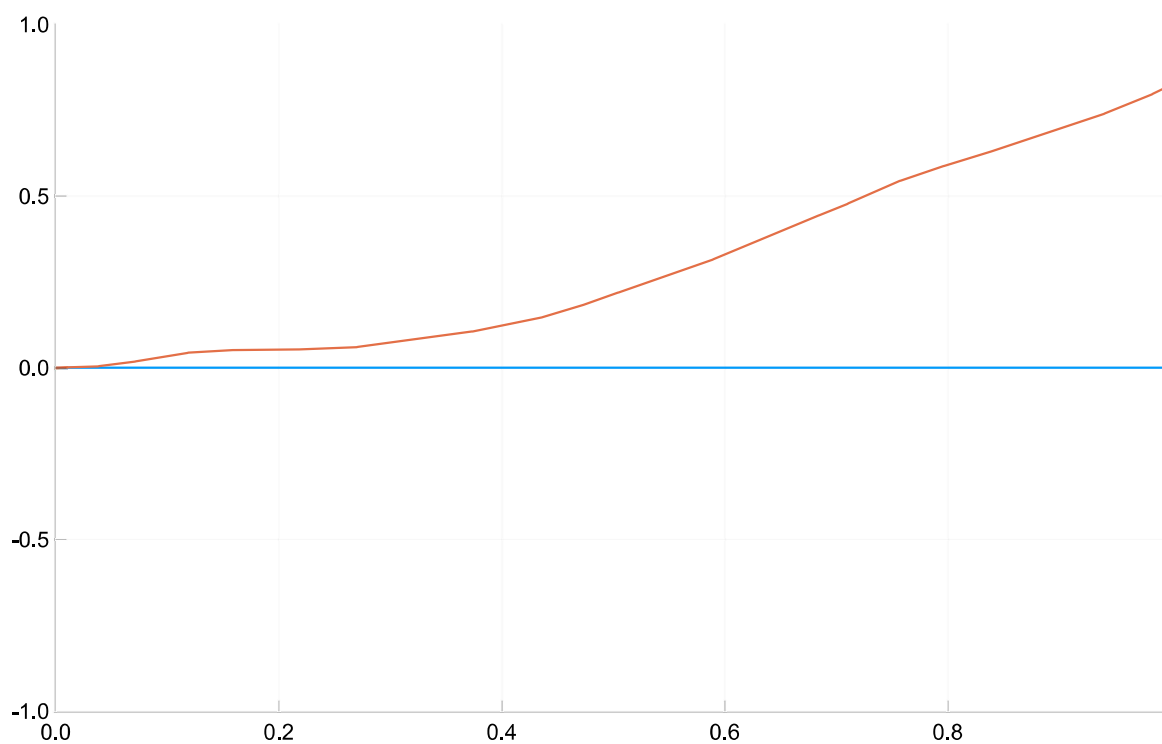


Area : -0.015789583592238993

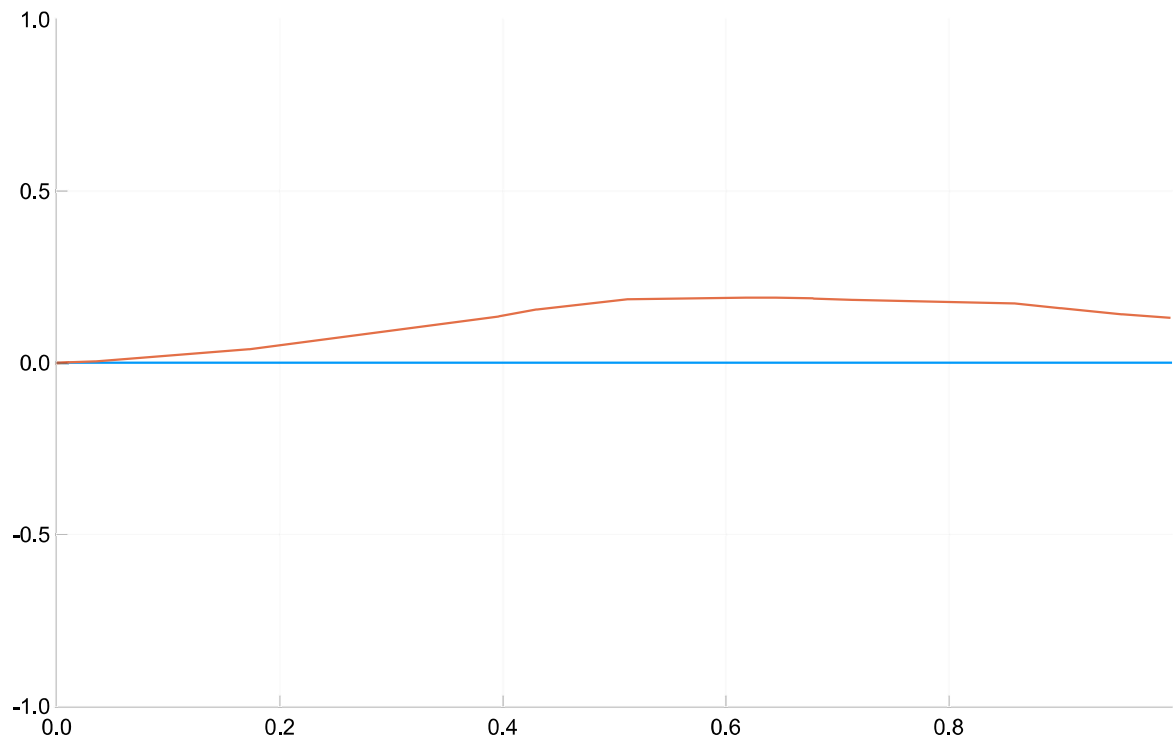




Area : 0.0973206529210069



Area : 0.29717777526685113



Compute the integral  $\int_0^t s dB_s$

```
In [16]: n = 10_000
t = 5
x = rand(Normal(0,sqrt(t/n)),n)
bm = [0;cumsum(x)]
steps = range(start = 0,stop = t, length = n+1)
integral = cumsum(steps[1:n].* (bm[2:n+1] .- bm[1:n]))
plot(steps[1:n],integral,label = :none)
```

Out[16]:



In [ ]: