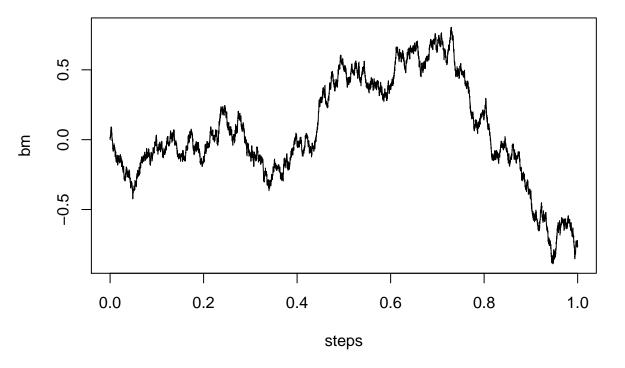
Brownian Motion & Option Pricing

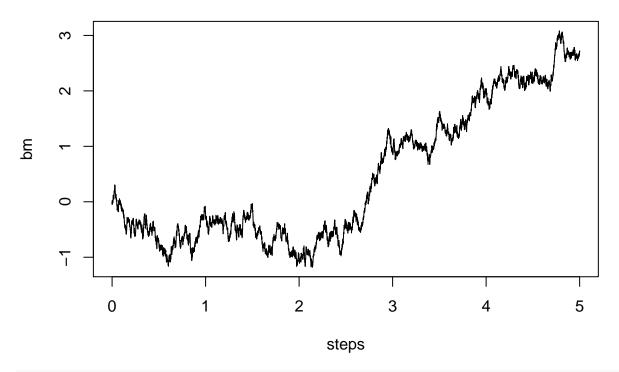
Aymen Rumi

Functions to Simulate Brownian Motion & Brownian Motion with Drift

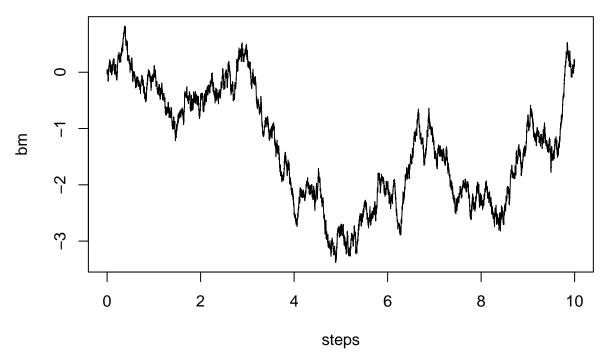
```
# Brownian Motion
SimulateBrownianMotion<-function(t,n)
{
    bm<-c(0,cumsum(rnorm(n,0,sqrt(t/n))))
    steps<-seq(0,t,length=n+1)
    plot(steps,bm,type="l")
}
# Simulations 1: t=1
SimulateBrownianMotion(1,10000)</pre>
```



```
# Simulations 2: t=5
SimulateBrownianMotion(5,10000)
```

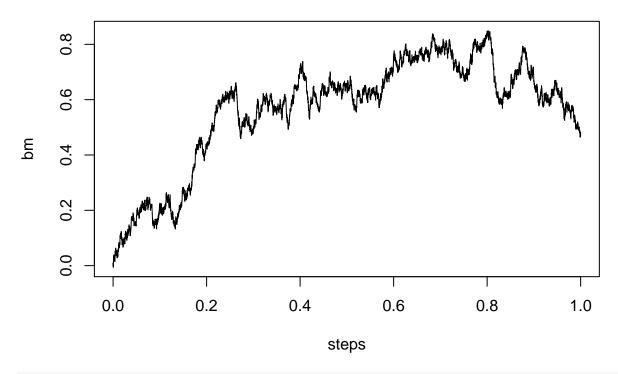


Simulations 3: t=10 SimulateBrownianMotion(10,10000)

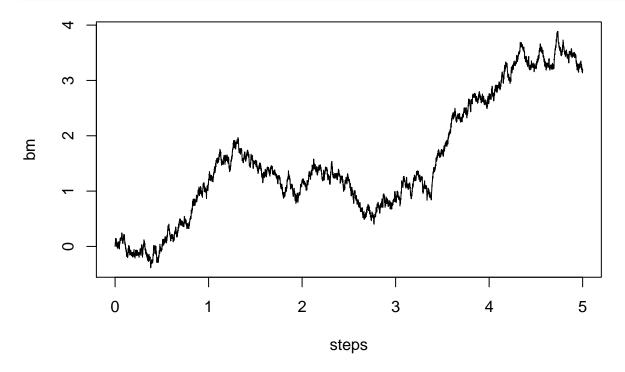


```
# Brownian Motion with Drift
SimulateBrownianMotionWithDrift<-function(t,n,mu,sigma)
{
    bm<-c(0,cumsum(rnorm(n,(mu*(t/n)),sqrt(sigma*(t/n)))))
    steps<-seq(0,t,length=n+1)
    plot(steps,bm,type="l")
}</pre>
```

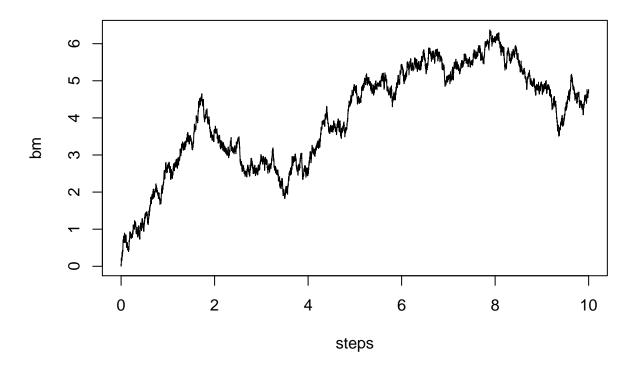
Simulations 1: t=1, mean=0.1, sigma=0.25
SimulateBrownianMotionWithDrift(1,10000,0.1,0.25)



Simulations 2: t=5, mean=0.25, sigma=0.75
SimulateBrownianMotionWithDrift(5,10000,0.25,0.75)



Simulations 2: t=10, mean=0.5, sigma=1.25
SimulateBrownianMotionWithDrift(10,10000,0.5,1.25)



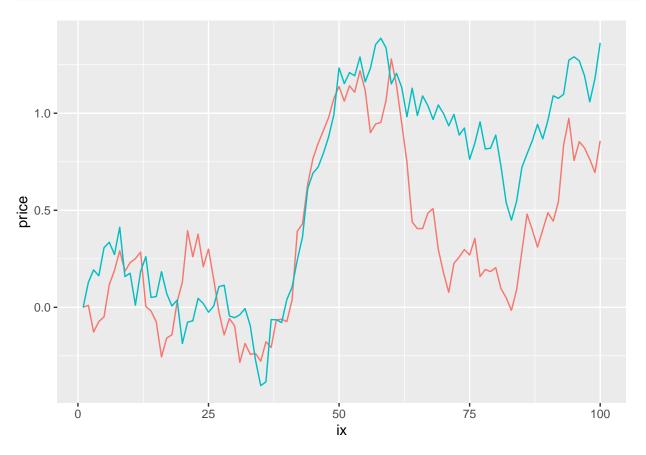
Repeated Brownian Motion Simulations: Geometric & BM with Drift

You can also embed plots, for example:

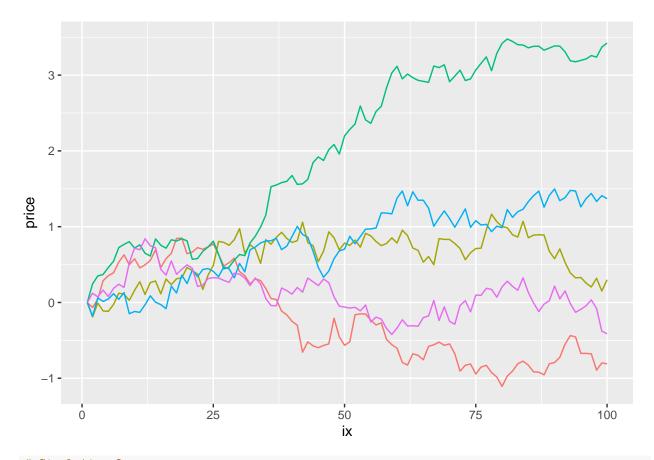
```
# Brownian Motion with Drift
BrownianMotionWithDrift_Simulation<-function(nsim,t,n,mu,sigma)
{
    gbm <- matrix(ncol = nsim, nrow = n)
    for (simu in 1:nsim) {
        gbm[1, simu] <- 0
        for (day in 2:n) {
            gbm[day, simu] <- gbm[(day-1), simu] + rnorm(1,(mu*(t/n)),sqrt(sigma*(t/n)))
            }
    }
    gbm_df <- as.data.frame(gbm) %>%
    mutate(ix = 1:nrow(gbm)) %>%
    pivot_longer(-ix, names_to = 'sim', values_to = 'price')
    ggplot(data=gbm_df,aes(x=ix, y=price, color=sim)) +
```

```
geom_line() +
   theme(legend.position = 'none')

# Simulation 1
BrownianMotionWithDrift_Simulation(2,0.5,100,2,3)
```



Simulation 2
BrownianMotionWithDrift_Simulation(5,0.5,100,2,3)

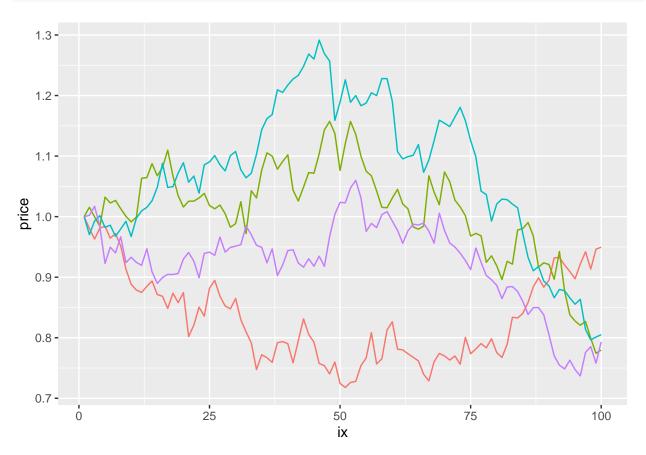


Simulation 3
BrownianMotionWithDrift_Simulation(100,0.5,100,0,0.5)

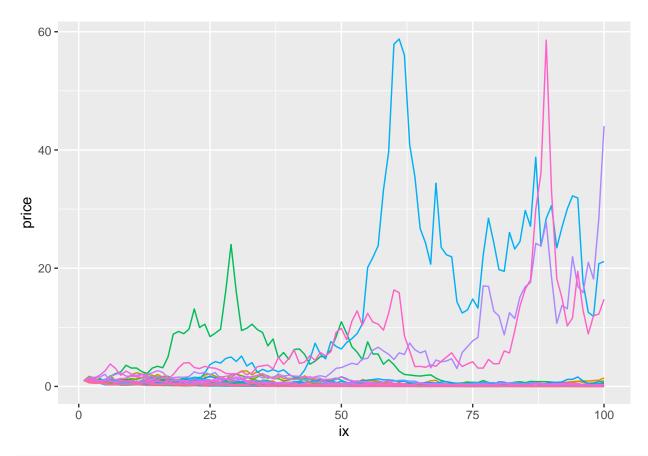


```
# Geometric Brownian Motion: Simulation of Stock Prices
GeometricBrownianMotion_Simulation<-function(nsim,t,mu,sigma,S0)</pre>
{
    gbm <- matrix(ncol = nsim, nrow = t)</pre>
    for (simu in 1:nsim) {
    gbm[1, simu] <- S0
    for (day in 2:t) {
      epsilon <- rnorm(1)</pre>
      dt = 1 / 365
      gbm[day, simu] <- gbm[(day-1), simu] * exp((mu - sigma * sigma / 2) * dt + sigma * epsilon * sqrt
    }
    gbm_df <- as.data.frame(gbm) %>%
    mutate(ix = 1:nrow(gbm)) %>%
    pivot_longer(-ix, names_to = 'sim', values_to = 'price')
    ggplot(data=gbm_df,aes(x=ix, y=price, color=sim)) +
    geom_line() +
    theme(legend.position = 'none')
```

```
# Sample 1: mean=0, sigma=0.5
GeometricBrownianMotion_Simulation(4,100,0,0.5,1)
```



Sample 2: mean=1, sigma=5
GeometricBrownianMotion_Simulation(25,100,1,5,1)



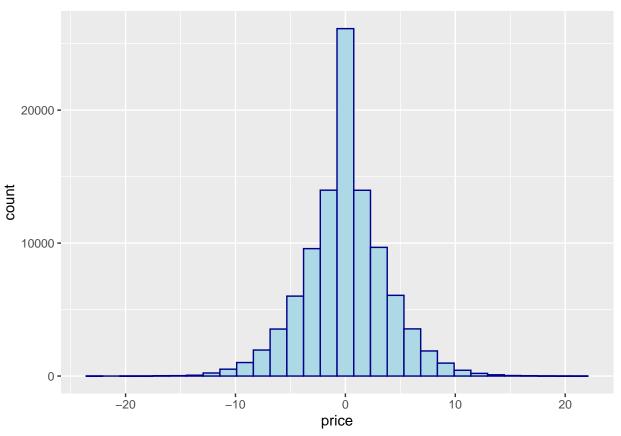
Sample 3: mean=0, sigma=1
GeometricBrownianMotion_Simulation(100,100,0,1,1)



Plotting Brownian Motion Simulation Histograms

```
# Function for Simulating Brownian Motion & Plotting Histogram
BrownianMotionWithDrift_Histogram<-function(nsim,t,n,mu,sigma)
{
    gbm <- matrix(ncol = nsim, nrow = n)
    for (simu in 1:nsim) {
        gbm[1, simu] <- 0
        for (day in 2:n) {
            gbm[day, simu] <- gbm[(day-1), simu] + rnorm(1,(mu*(t/n)),sqrt(sigma*(t/n)))
        }
    }
    gbm_df <- as.data.frame(gbm) %>%
    mutate(ix = 1:nrow(gbm)) %>%
    mutate(ix = 1:nrow(gbm)) %>%
    pivot_longer(-ix, names_to = 'sim', values_to = 'price')
```

```
ggplot(data=gbm_df,aes(x=price, color=sim)) +
   geom_histogram(color="darkblue", fill="lightblue",bins=30)+
   theme(legend.position = 'none')
}
# Plotting Histogram
BrownianMotionWithDrift_Histogram(10000,10,0,3)
```

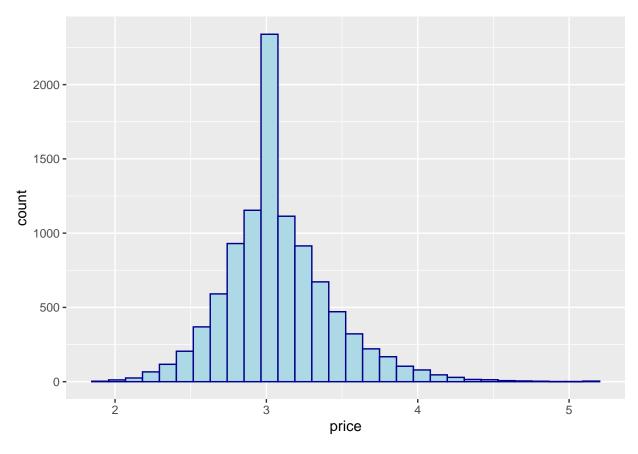


```
# Plotting Histogram for Geometric Brownian Motion
GeometricBrownianMotion_Histogram<-function(nsim,t,mu,sigma,S0)
{
    gbm <- matrix(ncol = nsim, nrow = t)
    for (simu in 1:nsim) {
        gbm[1, simu] <- S0
        for (day in 2:t) {
            epsilon <- rnorm(1)
            dt = 1 / 365
            gbm[day, simu] <- gbm[(day-1), simu] * exp((mu - sigma * sigma / 2) * dt + sigma * epsilon * sqrt
            }
        }
    }
}</pre>
```

```
gbm_df <- as.data.frame(gbm) %>%
  mutate(ix = 1:nrow(gbm)) %>%
  pivot_longer(-ix, names_to = 'sim', values_to = 'price')

ggplot(gbm_df, aes(x=price))+
  geom_histogram(color="darkblue", fill="lightblue",bins=30)+
  theme(legend.position = 'none')

GeometricBrownianMotion_Histogram(1000,10,2,1,3)
```



Financial Options & Black Scholes Pricing Model

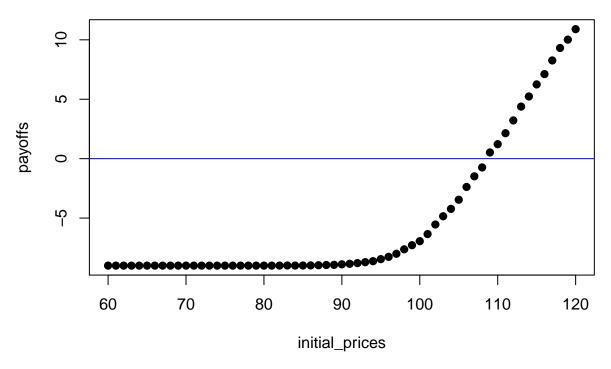
```
# Calculates mean expected payoff given initial stock price, premium, expiration...

# Finds Payoff by Simulating Brownian Motion

FinancialOptions<-function(initial_price, premium, expiration_date, strike_price, mu, sigma)
{</pre>
```

```
nsim<-1000
    t<-expiration_date
    gbm <- matrix(ncol = nsim, nrow = t)</pre>
    for (simu in 1:nsim) {
    gbm[1, simu] <- initial_price</pre>
    for (day in 2:t) {
      epsilon <- rnorm(1)</pre>
      dt = 1 / 365
      gbm[day, simu] <- gbm[(day-1), simu] * exp((mu - sigma * sigma / 2) * dt + sigma * epsilon * sqrt
        }
    }
    gbm_df <- as.data.frame(gbm) %>%
    mutate(ix = 1:nrow(gbm)) %>%
    pivot_longer(-ix, names_to = 'sim', values_to = 'price')
    X<-seq(0,length(gbm_df$price))</pre>
    for(i in 1:length(gbm_df$price))
        X[i] <-max(gbm_df$price[i]-strike_price,0)</pre>
    }
    return(mean(X)-premium)
}
# Simulating Payoffs for Different initial prices
initial_prices<-60:120
payoffs<-c()</pre>
for(i in initial_prices){
    payoffs<-c(payoffs,FinancialOptions(i,10,10,100,0.1,0.5))</pre>
plot(x=initial_prices,y=payoffs,pch=19,main="Initial Price vs Payoffs")
abline(h=0, col="blue")
abline(h=100+10, col="blue")
```

Initial Price vs Payoffs

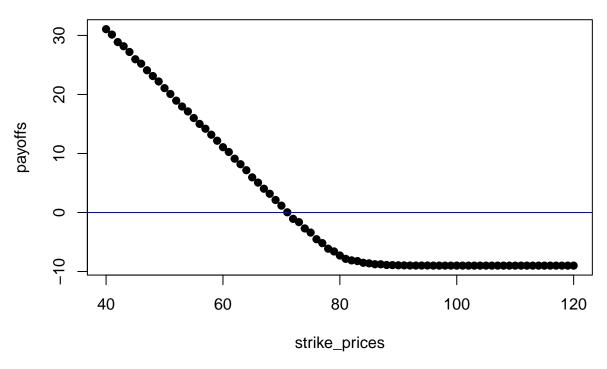


```
# Simulating Payoffs for Different strike prices
strike_prices<-40:120
payoffs<-c()

for(i in strike_prices){
    payoffs<-c(payoffs,FinancialOptions(80,10,10,i,0.1,0.5))
}

plot(x=strike_prices,y=payoffs,pch=19,main="Strike Price vs Payoff")
abline(h=0, col="blue")</pre>
```

Strike Price vs Payoff



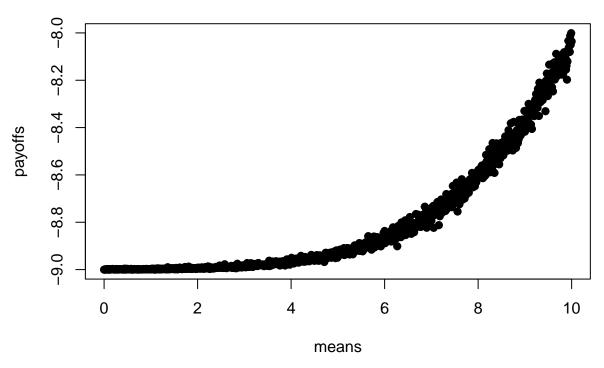
```
#Simulating Payoffs for Different means

means<-seq(0,10,0.01)
payoffs<-c()

for(i in means){
    payoffs<-c(payoffs,FinancialOptions(80,10,10,100,i,0.5))
}

plot(x=means,y=payoffs,pch=19,main="Means vs Payoffs")</pre>
```

Means vs Payoffs



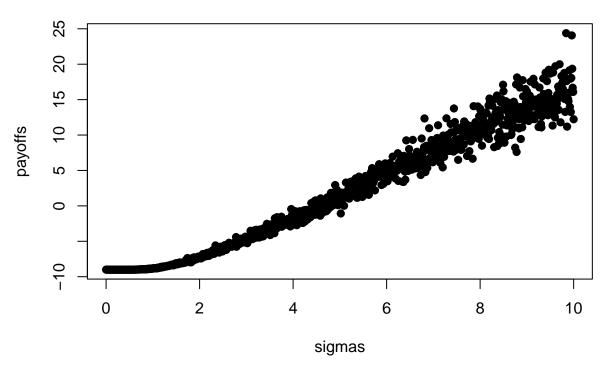
```
#Simulating Payoffs for Different volatilities

sigmas<-seq(0,10,0.01)
payoffs<-c()

for(i in sigmas){
    payoffs<-c(payoffs,FinancialOptions(80,10,10,100,0.1,i))
}

plot(x=sigmas,y=payoffs,pch=19,main="Volatilities vs Payoffs")</pre>
```

Volatilities vs Payoffs



```
# Black Scholes Option Pricing
# Gives Optimal Price to Pay for Given Option

BlackScholesOptionPricing<-function(initial_price,strike_price,expiration_date,interest_rate,volatility)
{
    t<-expiration_date/365
    alpha<-(log(strike_price/initial_price)-(interest_rate-volatility/2)*t)/sqrt(volatility)

    (initial_price*pnorm((alpha-sqrt(volatility)*t)/sqrt(t), mean = 0, sd = 1, lower.tail=FALSE))-
    (exp(-interest_rate*t)*strike_price*pnorm(alpha/sqrt(t),mean=0,sd=1,lower.tail=FALSE))
}

# Optimal price for premium for stock option priced at 80$ with strike price $100, 2% Interest Rate & s
BlackScholesOptionPricing(80,100,90,0.002,0.5)</pre>
```

[1] 4.966207