Oblique Reflected Diffusion Simulation with Penalty Method

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Introduction

Given a stochastic differential equation (SDE) with oblique reflection:

$$dX(t) = g(X(t))dt + \sigma(X(t))dW(t) + r(X(t))d\ell(t)$$
(1)

with $\ell(t)$ is a continuous nondecreasing process with $\ell(0) = 0$ which can increase only when it inside it domain and reflected back inside its domain in the direction of r(z) as X hits the domain at a point z. Once the process is in it domain, X(t) behaves as a solution to the SDE below.

$$dX(t) = g(X(t))dt + \sigma(X(t))dW(t). \tag{2}$$

We implement the numerical approximation of solutions of stochastic differential equations driven by brownian motion which are moving inside a domain and obliquely reflected at the boundary as follows:

$$z_{k+1} - z_k = [g(z_k) + \varphi(d(z_k)) \times r(y(z_k))]\epsilon + \sigma(z_k)\epsilon_{k+1}; \tag{3}$$

where $\epsilon_1, \epsilon_2, ... \sim N(0; \epsilon I_d)$. We define this function Z_{ϵ} as piecewise constant: $Z_{\epsilon}(t) := Z_{\epsilon}(k\epsilon) = z_k, t \in [k\epsilon; (k+1)\epsilon), k = 0, 1, ...$

${\it Half-line}$

Obliquely reflected brownian motion with penalty term

We simulate a brownian motion with a penalty term for half-line.

$$z(t+\epsilon) \approx z(t) + [\varphi(d(z(t))) \times r(y(z(t)))]\epsilon + \varepsilon_{t+\epsilon}$$

where Z(0) = 1, $\varepsilon_{t+\epsilon} \sim \mathcal{N}(0, \epsilon)$, $\epsilon = 0.01$, time horizon \$t=1, a=100,p=0.1 \$ and $\varepsilon_{t+\epsilon} \sim \mathcal{N}(0, \epsilon)$, $\varepsilon_{t+\epsilon} \sim$

$$\varphi(d(z(t))) \times r(y(z(t))) = \text{ penalty term with reflection} = \begin{cases} 0, & \text{if } z(t) \geq 0 \\ a[z(t)]^p, & \text{if } z(t) < 0 \end{cases}$$

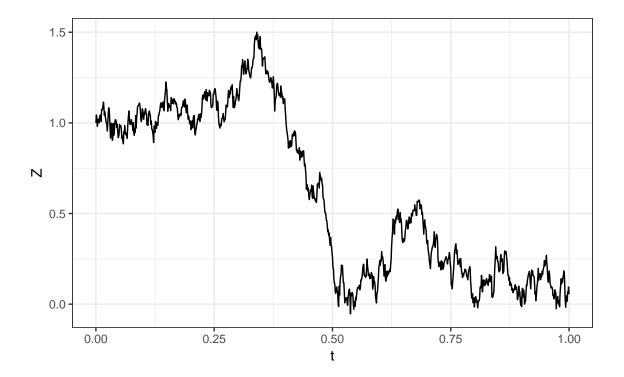


Figure 1: Obliquely reflected brownian motion with penalty

Convergence on half-line

A reflected brownian motion for unit negative drift has a stationary distribution, which is exponential with mean $\frac{1}{2}$, which serves as a limiting distribution for a long-time limit. See figure (1) below for a=100

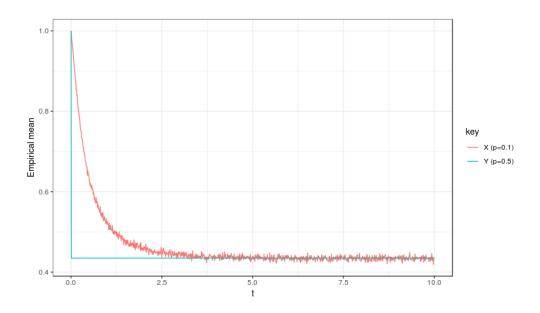
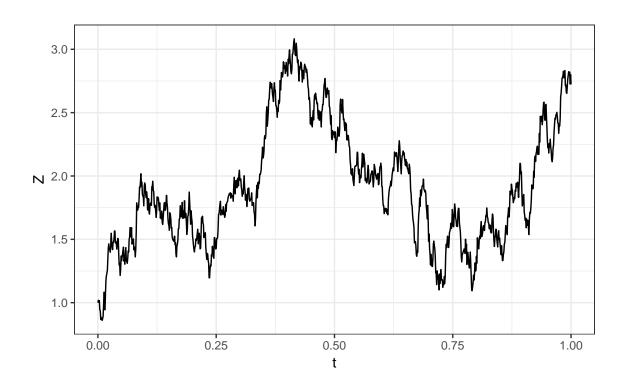


Figure 2: Stationary distribution of reflected brownian motion with unit negative drift

Obliquely reflected Ornstein-Uhlenbeck process with penalty term

$$dX(t) = 3(2 - X(t))dt + 2dW(t)$$



Two-dimensional SDE

Comparing 2-D SDE processes using Peacok test

R Code

```
rm(list = ls())
knitr::opts_chunk$set(
                   # don't show code
  echo = FALSE,
  comment = NA,
  error=FALSE ,
  warning = FALSE, # don't show warnings
  message = FALSE, # don't show messages (less serious warnings)
  cache = FALSE, # set to TRUE to save results from last compilation
 fig.align = "center",# center figures
 fig.pos='h'
## libraries
library(MASS)
library(dplyr)
library(tidyr)
library(ggplot2)
library(Peacock.test)
SDE_RF \leftarrow function(begin, end, step=0.01, drift=0, sigma=1, a=100, p=0.1, r=1, X0=0)
   t<-seq(begin,end,step) ### create sequence
   n=length(t)
   dt <- step
   dw<-rnorm(n, 0, sqrt(step))</pre>
   X<-NULL
   X[1] < -X0
   ## penalty function
  penalty<- function(a,p,r,x){</pre>
     ifelse(x>=0,0,r*a*((abs(x))^p))
   if(is.function(drift) & is.function(sigma)){
   ## drift
   d<- as.function(alist(x=,drift))</pre>
   drift < -d(1)
   ## sigma
```

```
sig<- as.function(alist(x=,sigma))</pre>
   sigma<-sig(1)</pre>
   for (i in 2:n) {
     X[i] \leftarrow X[i-1] + drift(X[i-1])*dt + penalty(a,p,r,X[i-1])*dt + sigma(X[i-1])*dw[i]
   }
   else if(is.function(drift) & !is.function(sigma)){
       ## drift
       d<- as.function(alist(x=,drift))</pre>
       drift < -d(1)
       for (i in 2:n) {
       X[i] \leftarrow X[i-1] + drift(X[i-1])*dt + penalty(a,p,r,X[i-1])*dt + sigma*dw[i]
     }
   else if(!is.function(drift) & is.function(sigma)){
       ## sigma
       sig<- as.function(alist(x=,sigma))</pre>
       sigma<-sig(1)</pre>
       for (i in 2:n) {
        X[i] \leftarrow X[i-1] + drift*dt + penalty(a,p,r,X[i-1])*dt + sigma(X[i-1])*dw[i]
    } else {
        for (i in 2:n) {
        X[i] <- X[i-1]+ drift*dt + penalty(a,p,r,X[i-1])*dt + sigma*dw[i]</pre>
   return(data.frame(t,X));
 ## Example
 drift<-function(x){-1}</pre>
 sigma1<-function(x){1}</pre>
 X1<-SDE_RF(0,1,step = 0.001,drift = drift, sigma = sigma1,</pre>
             a=100, p=0.1, r=1, X0=1)
X1%>%gather(key, value, X) %>%
    ggplot(aes(x=t, y=value)) +
    geom_line() +theme_bw()+
  ggtitle("") +
  xlab("t") + ylab("Z")
knitr::include_graphics(
  "C:/Users/Charles/Documents/UNR/BOOKS/SEM8/Stochastic Simulation/Rplot-Exponential")
## Example
 drift < -function(x) \{3*(2-x)\}
```

```
sigma1<-function(x){2}</pre>
 X1<-SDE_RF(0,1,step = 0.001,drift = drift, sigma = sigma1,</pre>
             a=100, p=0.1, r=1, X0=1)
X1%>%gather(key, value, X) %>%
    ggplot(aes(x=t, y=value)) +
    geom_line() +theme_bw()+
  ggtitle("") +
  xlab("t") + ylab("Z")
SDE 2D <- function(Begin, End,Step,X0){</pre>
  mu \leftarrow c(0,0) \# Mean
  sigma <- matrix(c(Step, 0, 0, Step),2)</pre>
  t<-seq(Begin, End, Step) ### create sequence
  n=length(t)
  dt <- Step
  X<- matrix(NA, nrow = n, ncol = 2)</pre>
  X[1,] <- X0
  for (i in 2:n) {
    drf1 < -(t(c(1,-1,1)) % * % matrix(c(X[(i-1), ],1))) * dt
    drf2 < ((t(c(-2,-1))%*\matrix(X[(i-1), ])))*dt
    Wt < c(2*X[(i-1),1]-3,4)* mvrnorm(1, mu = mu, Sigma = sigma)
    X[i,] \leftarrow c(X[(i-1),]) + c(drf1,drf2) + Wt
  }
  X<-data.frame(X)</pre>
  colnames(X)<-c("X1t","X2t")</pre>
  return(data.frame(t,X))
X6 < -SDE_2D(0,2,Step = 0.001,X0 = c(1,-3))
X6%>% gather(key, value, X1t, X2t) %>%
    ggplot(aes(x=t, y=value, colour=key)) +
    geom line() +theme bw()+
  ggtitle("SDE") +
  xlab("t") + ylab("Value")
ggplot(data = X6, aes(x=X1t, y=X2t)) +
    geom_point() +theme_bw()+
  ggtitle("SDE in 2-D") +
  xlab("X1(t)") + ylab("X2(t)")
peacock p value<-NULL
K<-100
for (i in 1:K) {
  Yt7 < -SDE_2D(0,2,Step = 0.001,X0=c(1,-3))
  Yt8 < -SDE_2D(0,4,Step = 0.002,X0=c(1,-3))
  peacock p value[i] <-peacock2(Yt8[,2:3],Yt7[,2:3])</pre>
```