

1 Brownian Motion Paths Hitting Lower Barrier

Fix $h > 0$. We will evaluate a standard Brownian motion Y starting at zero at times $t_j = jh$, $j = 1, 2, \dots$. This yields the discrete time process $Y_j = Y(t_j)$ which is a simple random walk with IID increments distributed as $\sqrt{h}N(0, 1)$. This is a very simple Markov process.

Given a sequence of lower barrier values d_j we say that Y *dies at time* t_j if $Y(t_j) < d_j$.

Now we compute the probability that $Y(t)$ is still alive at time $t = t_j$. This is the event

$$E_j = [Y(t_k) \geq d_k, \forall k \leq j]$$

Set $p_j := P[E_j]$. Observing that $E_{j+1} \subseteq E_j$ and using the Markov property we obtain the recursion

$$\begin{aligned} p_{j+1} &= P(E_{j+1}) = P(E_{j+1} \cap E_j) = P(E_j)P(E_{j+1}|E_j) \\ &= p_j P[Y_{j+1} \geq d_{j+1} | E_j] \\ &= p_j P[Y_{j+1} \geq d_{j+1} | Y_j \geq d_j] = p_j H(j) \quad \text{with} \\ H(j) &:= P[Y_{j+1} \geq d_{j+1} | Y_j \geq d_j] \end{aligned}$$

Here we have used the Markov property in the following way:

$$P(Y_{j+1} \geq d_{j+1} | Y_k \geq d_k, k = j, j-1, \dots, 1) = P(Y_{j+1} \geq d_{j+1} | Y_j \geq d_j). \quad (1)$$

Sloppy formulations of the Markov property suggest as much but the Markov property *does not imply this* and in fact in our case it is *not true*. This will then yield incorrect survival probabilities p_j and we will verify by simulation that the p_j so computed deviate substantially from realized survival probabilities in the simulation.

To compute $H(j)$ above, set $t = t_j$ and $h = t_{j+1} - t_j$ so that $t_{j+1} = t + h$ and we have

$$\begin{aligned} H(j) &= P[Y(t+h) \geq d_{j+1} | Y(t) \geq d_j] \\ &= P[Y(t+h) - Y(t) \geq d_{j+1} - Y(t) | Y(t) \geq d_j] \\ &= \frac{1}{P(Y(t) \geq d_j)} \int_{d_j}^{\infty} P[Y(t+h) - Y(t) \geq d_{j+1} - y | Y(t) = y] P_{Y(t)}(dy) \\ &= \frac{1}{P(Y(t) \geq d_j)} \int_{d_j}^{\infty} P[\sqrt{h}N(0, 1) \geq d_{j+1} - y] P_{Y(t)}(dy) \\ &= \frac{1}{F(-d_j/\sqrt{t})} \int_{d_j}^{\infty} F\left(\frac{y - d_{j+1}}{\sqrt{h}}\right) P_{Y(t)}(dy) \end{aligned}$$

where $F(x) = P(N(0, 1) \leq x)$ and we have used that the increment $Y(t+h) - Y(t)$ is independent of $Y(t)$ and distributed as $N(0, h) = \sqrt{h}N(0, 1)$. This yields

the recursion

$$p_{j+1} = p_j \times H(j) \quad \text{where} \quad (2)$$

$$H(j) = \frac{1}{F(-d_j/\sqrt{t_j})} \int_{d_j}^{\infty} F\left(\frac{y - d_{j+1}}{\sqrt{h}}\right) P_{Y(t_j)}(dy), \quad (3)$$

with starting condition (note $t_1 = h$)

$$p_1 = P[Y(t_1) \geq d_1] = F(-d_1/\sqrt{h}). \quad (4)$$

$H(j)$ as in (3) is implemented as `fcn_H`, see file `R/BM.R`.

Using the bivariate normal cumulative distribution function $F_C(I)$ where C is the covariance matrix of the distribution and I a two dimensional rectangle, we can also compute the conditional probability $H(j)$ as

$$\begin{aligned} H(j) &= P(Y(t_{j+1}) \geq d_{j+1} \mid Y(t_j) \geq d_j) = \frac{F_{C_j}(I_j)}{P(Y(t_j) \geq d_j)} \\ &= \frac{F_{C_j}(I_j)}{F(-d_j/\sqrt{t_j})}, \quad \text{where} \\ C_j &= \text{Cov}(Y(t_j), Y(t_{j+1})) = \begin{pmatrix} t_j & t_j \\ t_j & t_{j+1} \end{pmatrix} \quad \text{and} \\ I_j &= [d_j, +\infty] \times [d_{j+1}, +\infty]. \end{aligned} \quad (5)$$

This is implemented as `fcn_H1` in file `R/BM.R`. The R-package `mvtnorm` provides the required multinormal distribution function.

We check that the functions `fcn_H` and `fcn_H1` yield the same result, see `test_H_function` in file `R/Tests.R`.

In the code (function `runSimulation` in file `R/BM.R`) we run a simulation of Brownian paths and monitor the probability q_j of death at time t_j (by counting death events which are much rarer).

For $j > 1$ death at time t_j is the event $E_{j-1} \setminus E_j$ and since $E_j \subseteq E_{j-1}$ it follows that

$$q_j = P(E_{j-1} \setminus E_j) = p_{j-1} - p_j \quad (6)$$

with $p_0 = P(E_0) = 1$ (being alive at the start of the simulation). We observe a drastic difference between theoretical and realized death probabilities showing clearly that (1) fails.

We will compute the true survival probabilities p_j later.

1.1 Simulation

We will simulate paths $Y_j = Y(t_j)$ starting from $Y_0 = 0$ and keep count how many of these paths die at each time step t_j . This gives us an empirical probability q_j^e of death at time t_j which we will compare with the theoretical value q_j above.

There will be a deviation due to sample variation and we will report a p-value associated with this deviation. To do this we write the empirical probability q_j^e as the sample mean

$$q_j^e = (X_1 + X_2 + \dots + X_n)/n \quad (7)$$

where X_k is the indicator variable defined as

$$X_k = \begin{cases} 1 & \text{if } Y(t) \text{ dies at time } t = t_j \text{ in path } k \\ 0 & \text{else} \end{cases}$$

Then the $X_k = 0, 1$ are IID with $P(X_k = 1) = q_j$ from which it follows that $E[X_k] = q_j$ and $Var(X_k) = q_j(1 - q_j)$. By the central Limit Theorem the sample mean q_j^e in (7) is approximately normally distributed with mean $E[q_j^e] = q_j$ and variance

$$Var(q_j^e) = q_j(1 - q_j)/n$$

and because the sample size n in our simulation is extremely large (e.g. 4 million) this approximation is highly accurate.

Setting $\sigma_j := \sqrt{Var(q_j^e)}$ the variable $(q_j^e - q_j)/\sigma_j$ is approximately standard normal leading to a very accurate two sided p-value for a deviation δ of q_j^e from the theoretical value q_j as

$$\begin{aligned} pValue(\delta) &:= P(|q_j^e - q_j| \geq \delta) \\ &= P(|q_j^e - q_j|/\sigma_j \geq \delta/\sigma_j) \\ &= 2 * P(N(0, 1) \leq -\delta/\sigma_j) = 2 * F\left(-\frac{\delta\sqrt{n}}{\sqrt{q_j(1 - q_j)}}\right). \end{aligned} \quad (8)$$

In our simulation this is extremely close to zero for $j \geq 4$ from which it follows that the p_j as computed in (2), (3) above are incorrect. This verifies that (1) fails.

1.2 Correct survival probabilities

The probability p_j of survival to time t_j can be obtained directly from the multivariate normal distribution function $F_C(I)$ where C is the covariance matrix of the distribution and I a rectangle in \mathbb{R}^j . Indeed

$$\begin{aligned} p_j &= P(Y_j \geq d_j, Y_{j-1} \geq d_{j-1}, \dots, Y_1 \geq d_1) = F_{C_j}(I_j), \quad \text{where} \\ I_j &= [d_1, +\infty) \times [d_2, +\infty) \times \dots \times [d_j, +\infty) \end{aligned} \quad (9)$$

and C_j is the covariance matrix

$$C_j = Cov(Y_1, Y_2, \dots, Y_j) = (t_i \wedge t_k)_{i,k=1}^j. \quad (10)$$

The multivariate distribution function $F_C(I)$ is implemented in the R-package *mvtnorm* and we will use this to compute the true survival probabilities p_j from (9).