

## A.4 Simulation

Question A.4.3 is the most relevant on this sheet for MSc MCF students. Since it builds on the others, their statements, particularly A.4.1(e), A.4.2(b)-(c) are also relevant.

1. (a) Let  $(X_t)_{t \geq 0}$  be a Gamma process with  $X_t \sim \text{Gamma}(t, 1)$  for all  $t > 0$ . Consider  $A = X_a$  and  $B = X_{a+b} - X_a$  for some  $a > 0$  and  $b > 0$ . Show that  $R = A/(A + B)$  and  $S = A + B$  are independent and that  $R \sim \text{Beta}(a, b)$ , where Gamma and Beta densities are recalled as follows:

$$f_S(s) = \frac{s^{a+b-1}e^{-s}}{\Gamma(a+b)}, \quad s \in (0, \infty), \quad f_R(r) = \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)}r^{a-1}(1-r)^{b-1}, \quad r \in (0, 1).$$

Deduce, vice versa, the distribution of  $(SR, S(1 - R))$  for independent  $S \sim \text{Gamma}(c, 1)$  and  $R \sim \text{Beta}(cp, c(1 - p))$ , for some  $c > 0$  and  $p \in (0, 1)$ .

- (b) Let  $U \sim \text{Unif}(0, 1)$  and  $a > 0$ . Show that  $X = U^{1/a} \sim \text{Beta}(a, 1)$ .
- (c) Let  $U \sim \text{Unif}(0, 1)$  and  $V \sim \text{Unif}(0, 1)$  be independent and  $a \in (0, 1)$ . Calculate for  $Y = U^{1/a}$  and  $Z = V^{1/(1-a)}$

$$\mathbb{P}\left(\frac{Y}{Y+Z} \leq t, Y+Z \leq 1\right)$$

and deduce that the conditional distribution of  $W = Y/(Y+Z)$  given  $Y+Z \leq 1$  is  $\text{Beta}(a, 1 - a)$ . *Hint: Write both inequalities as constraints on  $Z$  to find the bounds when writing the probability as a double integral.*

- (d) In the setting of (c), show that the conditional distribution of  $TW$  given  $Y + Z \leq 1$ , for an independent  $T \sim \text{Exp}(1) = \text{Gamma}(1, 1)$  random variable, is  $\text{Gamma}(a, 1)$ .
- (e) Consider the following procedure due to Johnk. Let  $a \in (0, 1)$ .
  1. Generate two independent random numbers  $U, V \sim \text{Unif}(0, 1)$ .
  2. Set  $Y = U^{1/a}$  and  $Z = V^{1/(1-a)}$ .
  3. If  $Y + Z \leq 1$  go to 4., else go to 1.
  4. Generate an independent  $C \sim \text{Unif}(0, 1)$  and set  $T = -\ln(C)$ .
  5. Return the number  $TY/(Y + Z)$ .

What is this procedure doing? Explain its relevance for simulations.

2. (a) Based on Question A.4.1, explain how to generate a  $\text{Beta}(a, b)$  random variable from a sequence of  $\text{Unif}(0, 1)$  random variables, for any  $a > 0$  and  $b > 0$ . *Hint: Consider  $a \in (0, 1)$  first and use the additivity of Gamma variables to generate  $\text{Gamma}(a, 1)$  variables, from which the Beta variable can be constructed.*
- (b) Set  $X_0 = 0$  and generate  $X_1 \sim \text{Gamma}(1, 1)$ . For  $n \geq 0$ , having generated  $X_{k2^{-n}}$ ,  $k = 0, \dots, 2^n$ , generate  $B_{k,n} \sim \text{Beta}(2^{-n-1}, 2^{-n-1})$  and set  $X_{(2k-1)2^{-n-1}} = X_{(k-1)2^{-n}} + B_{k,n}(X_{k2^{-n}} - X_{(k-1)2^{-n}})$ ,  $1 \leq k \leq 2^n$ . Show that this approximates a Gamma process on the time interval  $[0, 1]$ .
- (c) What are the advantages of this method when compared with the plain version of the time discretisation method (Method 1)?

3. Consider a variant of the Variance Gamma process of the form  $V_t = at + G_t - H_t$  where  $a \in \mathbb{R}$ ,  $G_1 \sim \text{Gamma}(\alpha_+, \beta_+)$  and  $H_1 \sim \text{Gamma}(\alpha_-, \beta_-)$
- For what values of  $a, \alpha_{\pm}, \beta_{\pm}$  is  $V$  a martingale?
  - Write out the steps needed to simulate  $V_t$ 
    - by Method 1 (using a random walk with increment distribution  $\sim V_{\delta}$ )
    - by Method 1 (applied to  $G$  and  $H$  separately)
    - by the refinement of Method 1 given in A.4.3
    - by Method 2 (simulating the Poisson point process of jumps truncated at  $\varepsilon$ )
  - Carry out 9 simulations for a range of parameters  $\alpha_+ \in \{1, 10, 100\}$  and  $\alpha_- \in \{10, 100, 1000\}$ ,  $\beta_{\pm} = \alpha_{\pm}^2/2$  and  $a$  such that  $V$  is a martingale. *This part of this question is optional.*

*Warning: The incomplete Gamma function  $\Gamma_t(a) = \int_0^t x^{a-1}e^{-x}dx$  cannot be simplified into closed form (nor expressed in terms of the Gamma function), except for some special values of  $a$  such as  $a \in \mathbb{N}$ . There are, however, numerical procedures to evaluate  $\Gamma_t(a)$ , which we will not address in this course.*

*If you have not used R, but would like to, you will find the “First steps with R” at*

*[http://www.stats.ox.ac.uk/~myers/stats\\_materials/R\\_intro/WA5\\_R.pdf](http://www.stats.ox.ac.uk/~myers/stats_materials/R_intro/WA5_R.pdf)*

*useful. Following are brief explanations of the commands used in the sample file*

*<http://www.stats.ox.ac.uk/~winkel/gammaavgamma.R>*

*This is a script file, which has to be run in the command window “R Console” to make the new commands available, e.g. select “Run all” in the drop-down menu “Edit”.*

- `runif(n,a,b)` generates an  $n$ -vector of uniform variables on  $[a, b]$ .
- `qgamma(u,a,b)` evaluates  $F^{-1}(u)$  for the  $\text{Gamma}(a, b)$  inverse distribution function  $F^{-1}$  at  $u$ . If  $\mathbf{u}$  is a vector, `qgamma` is applied to each component.
- `1:n` generates the vector  $(1, 2, \dots, n)$ . Multiplication of vectors  $\mathbf{v}$  by scalars  $\mathbf{a}$  can be written as `a*v`, similarly for addition and subtraction of vectors.
- `plot(x,y,pch=".",sub=paste("text"))` produces a scatter plot of pairs  $(x_i, y_i)$  for vectors  $\mathbf{x}$  and  $\mathbf{y}$ , with `.` marking the points, and `text` in the caption.
- `psum <- function(vector){...}` defines a new command `psum` that takes a vector `vector` as an argument. When this line is executed, the command is just made available. To execute the command, type `psum(v)` for a vector  $\mathbf{v}$  to get the partial sums of  $\mathbf{v}$  displayed, or `s=psum(v)` to create a new vector  $\mathbf{s}$  containing the partial sums of  $\mathbf{v}$ .