

## Q1) Monte Carlo Algorithm for Rainbow Option

a)

For  $d$ -correlated assets the prices at maturity are  $(S_T^{(j)})_{j=1,\dots,d}$ . Under  $\mathbb{Q}$ ,  $T = 1$ , and the Black Scholes model, the solution to the price at maturity for all  $j = 1, 2, \dots, d$  is:

$$S_T^{(j)} = S_0^{(j)} \exp(r - (\sigma^{(j)})^2/2 + \sigma^{(j)} W_T^{(j)})$$

Where for each stock  $j = 1, 2, \dots, d$ :

- $S_0^{(j)}$  is its initial price of stock
- $\sigma^{(j)}$  is its standard deviation
- $W_T^{(j)}$  is the brownian motion at maturity such that for another stock  $i$  we have correlated brownian motion processes  $Cor(W_t^{(j)}, W_t^{(i)}) = \rho$ .

Define the correlation matrix.

$$\Sigma = \begin{bmatrix} 1 & \rho & \dots & \rho \\ \rho & 1 & \dots & \rho \\ \vdots & \vdots & \ddots & \vdots \\ \rho & \rho & \dots & 1 \end{bmatrix} \in \mathbb{R}^{d \times d}$$

We want  $X = (W_T^{(1)}, \dots, W_T^{(d)}) \sim N_d(0, \Sigma)$ , so let  $Z^j \sim N(0, 1)$  for  $j = 1, \dots, d$  such that for a matrix  $A \in \mathbb{R}^{d \times d}$  satisfying  $\Sigma = AA^T$  we have  $X = AZ$  (equal in distribution). Let  $A$  be the **Cholesky factor** of  $\Sigma$ .

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### Algorithm

1. Given  $S_0^{(j)}, \sigma^{(j)}$  for all assets  $j = 1, \dots, d$
2. Initialize  $V_T :: \text{float}[d] \leftarrow 0 \in \mathbb{R}^d$
3. Compute  $A \leftarrow \text{cholesky}(\Sigma)$
4. for  $i = 1, \dots, n$  do:
  - Sample  $Z_1, \dots, Z_d \sim^{iid} N(0, 1)$  and let  $Z = (Z_1, \dots, Z_d)$
  - Let  $X = AZ \in \mathbb{R}^d$  (matrix multiplication)
  - Let  $S_{MAX} = -1$
  - for  $j = 1, \dots, d$ :
    - Let  $S_T^{(j)} = S_0^{(j)} \exp\{r - (\sigma^{(j)})^2/2 + \sigma^{(j)} X[j]\}$
    - If  $(S_T^{(j)} > S_{MAX}) : S_{MAX} = S_T^{(j)}$
  - Set  $V_T[i] = \exp(-r) \times \max(S_{MAX} - K, 0)$
5. return  $\text{mean}(V_T) \pm \frac{1.96 \times \text{sd}(V_T)}{\sqrt{n}}$

b)

```
getAmatrix <-function(rho, d){
  ## Construct correlation matrix Sigma for d-RV's with pairwise correlation rho.
  Sigma <- matrix(rep(rho, d^2), nrow = d, ncol = d) # populate with rho
  diag(Sigma) <- 1 # Set diagonal equal to 1
  ## Get matrix A such that AA^T = Sigma
  t(chol(Sigma))
}

partA <- function(d, S0s, sigmas, rho, K, r, n) {
  A <- getAmatrix(rho, d)
  vals <- rep(0, n) # values at maturity for each simulation
  for (i in 1:n){
    Z <- rnorm(d)
    X <- A%*%Z
    Smax <- -1
    for (j in 1:d){
      STj <- S0s[j] * exp((r-sigmas[j]^2/2) + sigmas[j]*X[j])
      if (STj > Smax) {
        Smax = STj
      }
    }
    vals[i] <- max(exp(-r)*(Smax-K), 0)
  }
  MCest <- mean(vals)
  MCsd <- sd(vals)
  alpha <- qnorm(.975)
  lo <- MCest-alpha*MCsd/sqrt(n)
  hi <- MCest+alpha*MCsd/sqrt(n)
  ans <- c(MCest, lo, hi)
  ## names(ans) <- c("estimate", "low", "high")
  return(ans)
}

partB <- function(rho) {
  set.seed(123)
  ans<-partA(d= 2, S0s = c(100, 100), sigmas = c(.2, .3), rho=rho, K=110, r=0.05, n=100000)
  return(ans[1])
}
```

```
rhos <- seq(-0.9, 0.9, 0.1)
estimates <- sapply(rhos, partB)
df <- data.frame(rhos, estimates)
print(df)
```

```
##      rhos estimates
## 1 -0.9 16.10186
## 2 -0.8 16.04699
## 3 -0.7 15.95444
## 4 -0.6 15.83054
## 5 -0.5 15.68365
## 6 -0.4 15.51868
## 7 -0.3 15.33255
## 8 -0.2 15.12776
## 9 -0.1 14.90380
## 10 0.0 14.66019
## 11 0.1 14.39500
## 12 0.2 14.10670
## 13 0.3 13.79058
## 14 0.4 13.44057
## 15 0.5 13.05431
## 16 0.6 12.62352
## 17 0.7 12.14263
## 18 0.8 11.59120
## 19 0.9 10.93301
```

```
plot(x = rhos, y = estimates, main = "Correlation vs MC price estimates",
     ylab = "estimate", xlab = "pairwise correlation")
```

### Correlation vs MC price estimates

