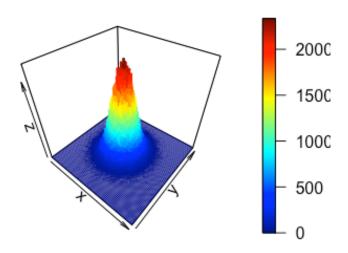
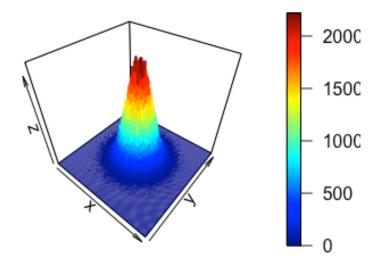
#### iMH generated data



#### By Chenxiao Niu

```
32 # Function of Independent Metropolis-Hastings algorithm
33 - iMH <- function (n, mu1, s1, mu2, s2, rho) {
      # Parameters for bivariate normal distribution
34
35
      mu <- c(mu1,mu2) # Mean</pre>
36
      Sigma <- matrix(c(s1^2, s1*s2*rho, s1*s2*rho, s2^2), 2, 2) # Covariance matrix
37
      I \leftarrow matrix(c(1,0,0,1),2,2)
38
      mat <- matrix(nrow = n, ncol = 2)</pre>
39
      mat[1, ] <- mvrnorm(1, mu, Sigma)</pre>
                                           # initialize the chain from the stationary
40 ▼
      for (t in 2 : n) {
41
        # Sample from proposal
42
        xStar <- mvrnorm(1, mat[t-1, ], I)</pre>
43
44
        # The acceptance ratio
        c = dmvnorm(mat[t-1, ], xStar) / dmvnorm(xStar, mat[t-1, ])
45
        alpha = min(1, dmvnorm(xStar) / dmvnorm(mat[t-1, ]))
46
47
48
        # Decide to accept
49
        mat[t, ] \leftarrow mat[t-1, ] + (xStar - mat[t-1, ]) * (runif(1) < alpha)
      }
50
51
      mat
52
   }
```

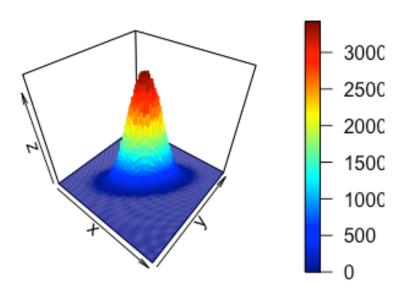
## Random Walk generated data



### By Chenxiao Niu

```
54 # Function of random walk Metropolis-Hastings algorithm
55 rwMH <- function (n, mu1, s1, mu2, s2, rho) {
      # Parameters for bivariate normal distribution
57
      mu <- c(mu1,mu2) # Mean</pre>
58
      Sigma <- matrix(c(s1^2, s1^*s2^*rho, s1^*s2^*rho, s2^2), 2, 2) # Covariance matrix
59
      I \leftarrow matrix(c(1,0,0,1),2,2)
60
      mat <- matrix(nrow = n, ncol = 2)</pre>
      mat[1, ] <- mvrnorm(1, mu, Sigma)</pre>
                                              # initialize the chain from the stationary
61
62 ₹
      for (t in 2:n) {
63
         # Sample from proposal distribution
        xStar <- mvrnorm(1, mat[t-1, ], I)</pre>
64
65
        # Do the random walk
66
         #Y <- dmvnorm(xStar)</pre>
        Y <- mvrnorm(1, xStar, I)
67
68
         # Compute the probability
        prob <- min(1, dmvnorm(Y)/dmvnorm(mat[t-1, ]))</pre>
69
70
        # Decide to accept
        mat[t, ] <- mat[t-1, ] + (Y - mat[t-1, ]) * (runif(1) < prob)</pre>
71
72
73
      mat
74 }
```

# Gibbs generated data



## By Chenxiao Niu

```
77 # Function of Gibbs Sampler to generate
78 gibbs<-function (n, mu1, s1, mu2, s2, rho)
79 ₹ {
80
      mat <- matrix(ncol = 2, nrow = n)</pre>
81
      x <- 0
82
       y <- 0
83
       mat[1, ] \leftarrow c(x, y)
84 -
       for (i in 2:n) {
85
        x \leftarrow rnorm(1, mu1 +
                       (s1/s2) * rho * (y - mu2), sqrt((1 - rho^2)*s1^2))
86
87
        y \leftarrow rnorm(1, mu2 +
                       (s2/s1) * rho * (x - mu1), sqrt((1 - rho^2)*s2^2))
88
89
        mat[i, ] <- c(x, y)
      }
90
91
      mat
92 }
```