

Replication Code

Usage of package samMCMC

Suppose we want to sample from a simple normal distribution. This of course can most easily and efficiently be done with the `rnorm` function in the core `stats` package; however, to demonstrate the usage of `samMCMC` we can start with this simple case. To sample from a normal distribution we first define its density

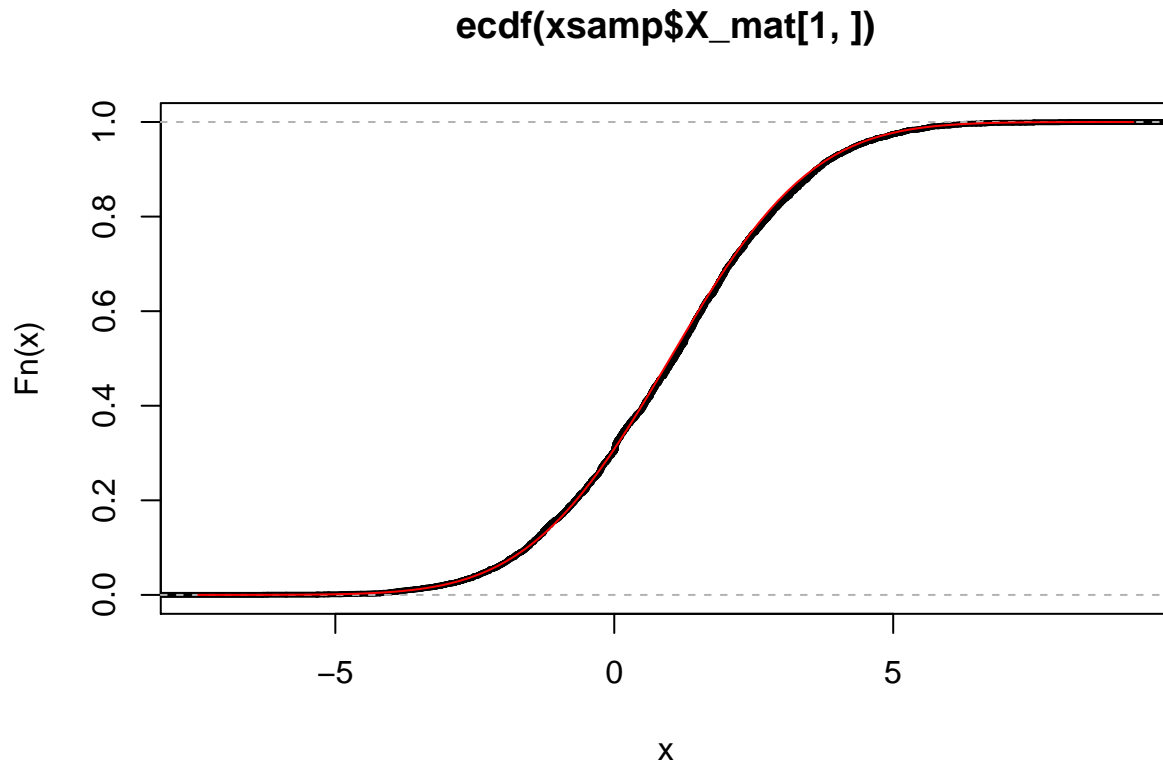
```
# normal density where `x` is the variable, the first element of `pars` is the mean  
# and the second element of `pars` is the S.D.  
egFunc <- function(x, pars) {  
  dnorm(x, pars[1], pars[2])  
}
```

Now we can use of `samMCMC` to sample from, in this example, a normal $N(1,4)$ distribution.

```
# control parameters for `samMCMC`  
numSampBurn <- 100  
t0 <- 100  
numSamp <- 1000  
numSampBurn <- 10000  
thinning <- 100  
  
# sample; we set control parameters `direct` to `TRUE` because we are directly  
# sampling from the density  
xsamp <- samMCMC(egFunc, init = 0, pars = c(1, 2),  
  control = list(numSamp = numSamp, t0 = t0, numSampBurn = numSampBurn,  
    thinning = thinning, direct = TRUE))
```

We can then compare the sample from `samMCMC` to its known generative distribution visually

```
plot(ecdf(xsamp$X_mat[1, ]), lwd = 3)  
curve(pnorm(x, 1, 2), add = TRUE, col = 'red')
```



We can also use `samMCMC` to sample given a cost function as described earlier. To do this we define a cost function (for simplicity we again use a normal distribution) and specify `direct = FALSE` in the control parameters in addition to specifying a temperature.

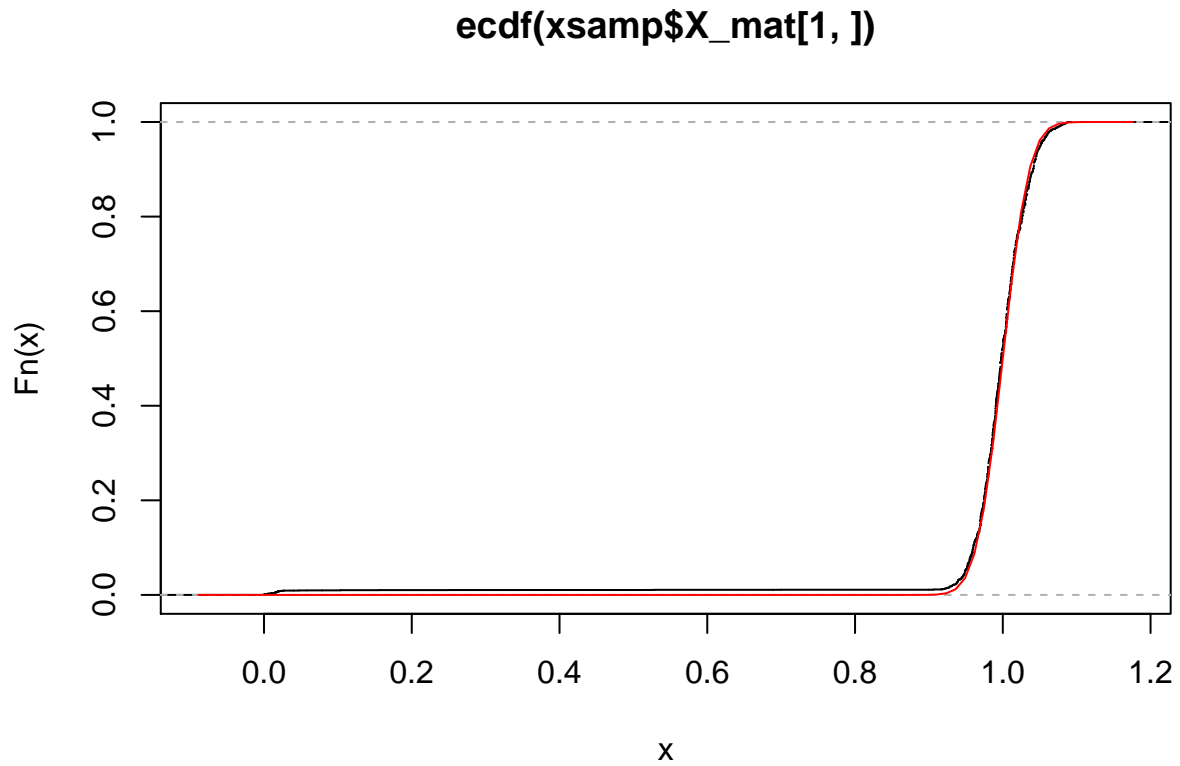
```
# normal likelihood where `m` is the mean (sd is fixed at 2) and `dat` are the data
egCostFunc <- function(m, dat) {
  - sum(dnorm(dat, m, 2, log = TRUE))
}

# example data
egData <- rnorm(5000, 1, 2)

# sample using previously defined control parameters but now with `direct = TRUE` and `temp = 1`
xsamp <- samMCMC(egCostFunc, init = 0, dat = egData,
  control = list(numSamp = numSamp, t0 = t0, numSampBurn = numSampBurn,
    thinning = thinning, direct = FALSE, temp = 1))
```

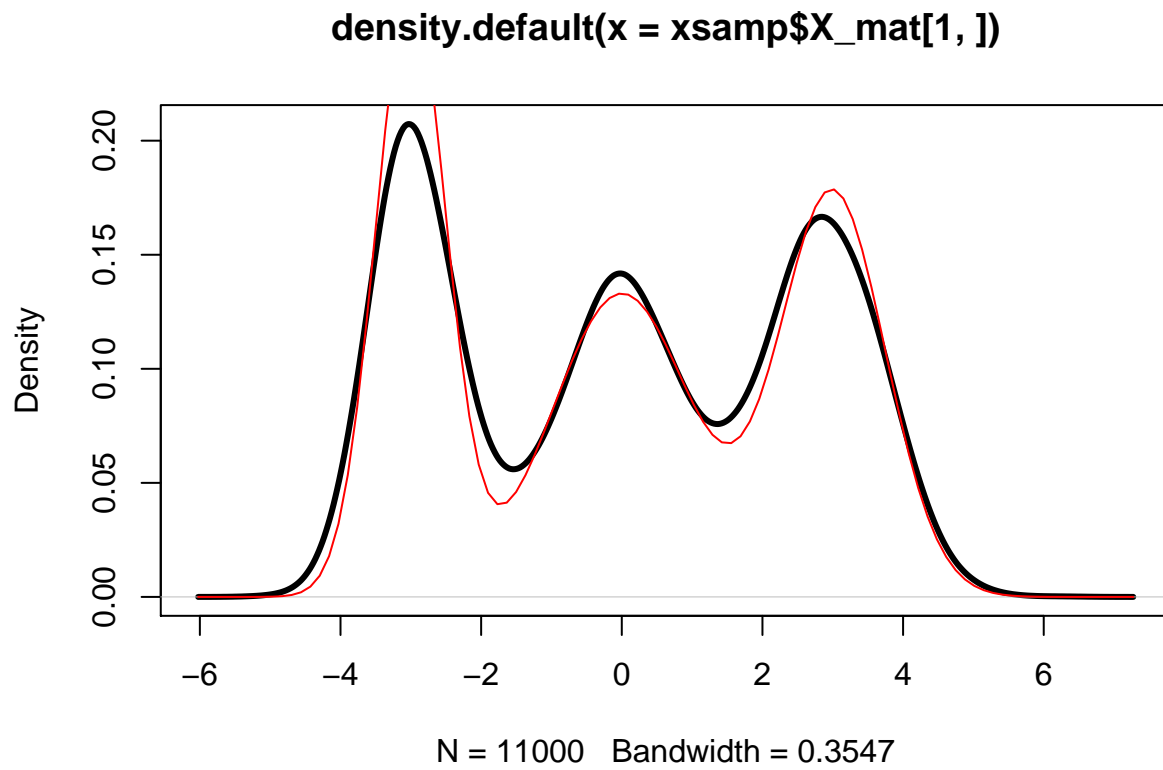
Again, we can compare this sample from `samMCMC` with its known distribution, in this case the sampling distribution of the MLE for the mean of a normal distribution with known variance, which is $N(\hat{\mu}, \sqrt{\sigma^2/N})$.

```
plot(ecdf(xsamp$X_mat[1, ]))
mleSD <- sqrt(4 / length(egData))
curve(pnorm(x, mean = 1, sd = mleSD), add = TRUE, col = 'red')
```



A more realistic use case of `samMCMC` is when the distribution we would like to sample from cannot be easily inverted, or the distribution we would like to approximate with a sample cannot be expressed analytically. We take as a simple example a mixture of normal distributions. In this example we use an equal mixture of three normals: $N(-1, 0.5)$, $N(0, 1.5)$, $N(1, 0.75)$.

```
mixFunc <- function(x) {  
  1/3 * dnorm(x, -3, 0.5) + 1/3 * dnorm(x, 0, 1) + 1/3 * dnorm(x, 3, 0.75)  
}  
  
xsamp <- samMCMC(mixFunc, init = 0,  
  control = list(numSamp = numSamp, t0 = t0, numSampBurn = numSampBurn,  
    thinning = thinning, direct = TRUE))  
  
plot(density(xsamp$X_mat[1, ]), lwd = 3)  
curve(mixFunc(x), add = TRUE, col = 'red')
```



Usage of package parTempeR