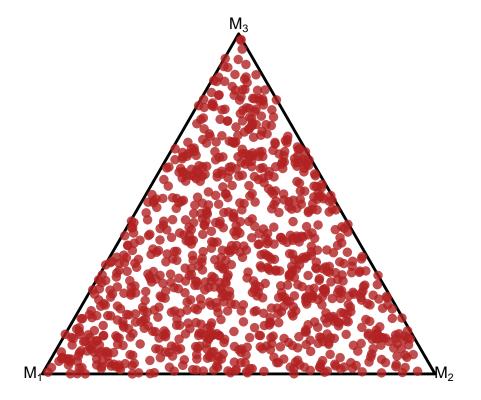
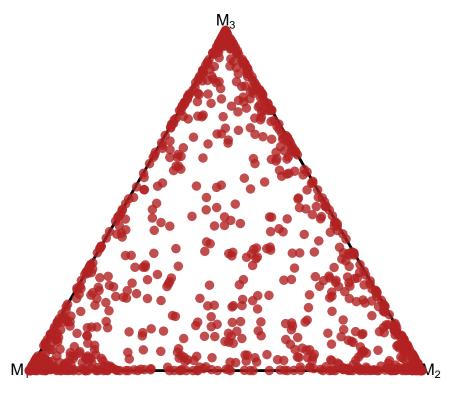
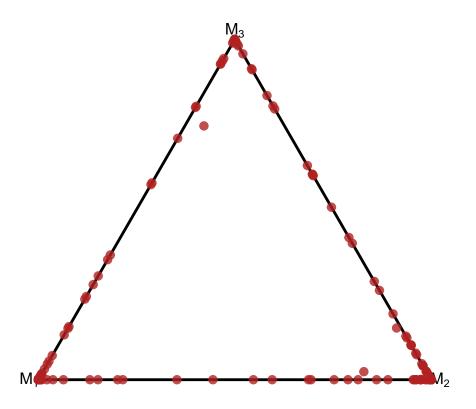
# Finite mixtures, Dirichlet processes and (dependent) Dirichlet process mixtures

2023-12-12

## Reproduce plots from the Dirichlet distribution







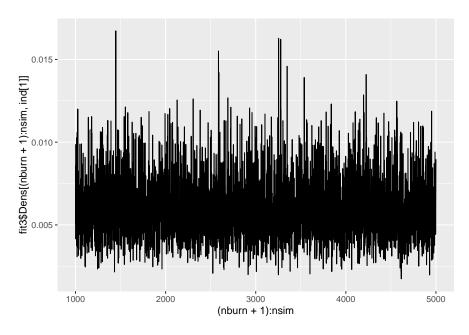
#### Finite mixture model

#### Simulated data example

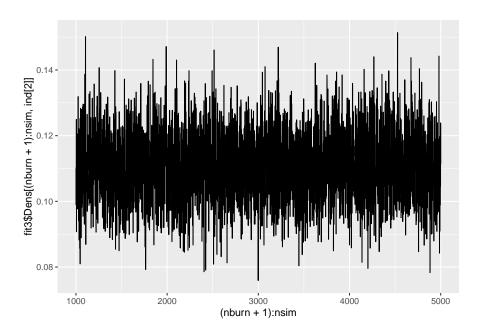
```
require(mixtools) #to simulate data from mixture of normals
require(MCMCpack) #to simulate random numbers from a Dirichlet distribution
require(coda)
\# function implementing a location scale mixture of K
# (K>=2) normal dists
fmm <- function(y, grid, K, amu, b2mu, asigma2, bsigma2, alpha,</pre>
    nsim) {
    n <- length(y)
    ngrid <- length(grid)</pre>
    prop <- prob <- matrix(0, nrow = n, ncol = K)</pre>
    P <- Mu <- Sigma2 <- matrix(0, nrow = nsim, ncol = K)
    Dens <- array(0, c(nsim, ngrid, K))</pre>
    Densm <- matrix(0, nrow = nsim, ncol = ngrid)</pre>
    z \leftarrow rep(1, n)
    ns \leftarrow rep(0, K)
    P[1, ] <- rdirichlet(1, alpha)</pre>
    Mu[1, ] <- rep(mean(y), K)
    Sigma2[1, ] <- rep(var(y), K)</pre>
```

```
for (i in 2:nsim) {
                     for (k in 1:K) {
                                prop[, k] \leftarrow P[i - 1, k] * dnorm(y, mean = Mu[i -
                                           1, k], sd = sqrt(Sigma2[i - 1, k]))
                     prob <- prop/apply(prop, 1, sum)</pre>
                     for (1 in 1:n) {
                                z[1] = sample(1:K, size = 1, prob = prob[1, ])
                     P[i, ] <- rdirichlet(1, alpha + tabulate(z, nbins = K))</pre>
                     for (k in 1:K) {
                                ns[k] <- length(which(z == k))</pre>
                     for (k in 1:K) {
                                 varmu \leftarrow 1/((1/b2mu) + (ns[k]/Sigma2[i - 1, k]))
                                meanmu \leftarrow ((sum(y[z == k])/Sigma2[i - 1, k]) + (amu/b2mu))/((1/b2mu) + (amu/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/((1/b2mu))/
                                            (ns[k]/Sigma2[i - 1, k]))
                                Mu[i, k] <- rnorm(1, mean = meanmu, sd = sqrt(varmu))</pre>
                                 Sigma2[i, k] \leftarrow 1/rgamma(1, asigma2 + (ns[k]/2),
                                           bsigma2 + 0.5 * sum((y[z == k] - Mu[i, k])^2))
                                Dens[i, , k] <- P[i, k] * dnorm(grid, Mu[i, k], sqrt(Sigma2[i,</pre>
                                           k]))
                     }
                     for (j in 1:ngrid) {
                                Densm[i, j] <- sum(Dens[i, j, ])</pre>
                     }
          return(list(P = P, Mu = Mu, Sigma2 = Sigma2, Dens = Densm))
}
# simulating data and defining grid where to evaluate the
# density
n <- 500
set.seed(123)
y \leftarrow rnormmix(n, c(0.3, 0.3, 0.4), c(-6, 0, 6), c(1, 1, 1))
grid \leftarrow seq(min(y) - 1, max(y) + 1, len = 200)
ngrid <- length(grid)</pre>
nsim <- 5000
# fitting the model for K=2, 3, 4, 5, 20, and 50
set.seed(123)
fit2 \leftarrow fmm(y = y, grid = grid, K = 2, amu = 0, b2mu = 100, asigma2 = 0.1,
          bsigma2 = 0.1, alpha = rep(1, 2), nsim = nsim)
```

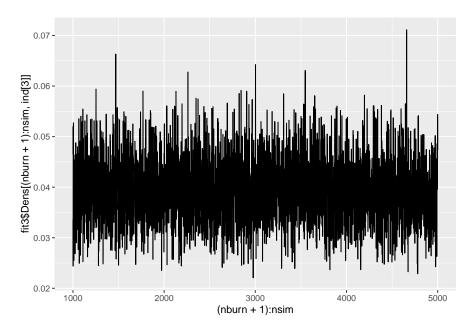
```
set.seed(123)
fit3 \leftarrow fmm(y = y, grid = grid, K = 3, amu = 0, b2mu = 100, asigma2 = 0.1,
    bsigma2 = 0.1, alpha = rep(1, 3), nsim = nsim)
set.seed(123)
fit4 \leftarrow fmm(y = y, grid = grid, K = 4, amu = 0, b2mu = 100, asigma2 = 0.1,
    bsigma2 = 0.1, alpha = rep(1, 4), nsim = nsim)
set.seed(123)
fit5 < -fmm(y = y, grid = grid, K = 5, amu = 0, b2mu = 100, asigma2 = 0.1,
    bsigma2 = 0.1, alpha = rep(1, 5), nsim = nsim)
set.seed(123)
fit20 \leftarrow fmm(y = y, grid = grid, K = 20, amu = 0, b2mu = 100,
    asigma2 = 0.1, bsigma2 = 0.1, alpha = rep(1, 20), nsim = nsim)
set.seed(123)
fit50 <- fmm(y = y, grid = grid, K = 50, amu = 0, b2mu = 100,
    asigma2 = 0.1, bsigma2 = 0.1, alpha = rep(1, 50), nsim = nsim)
set.seed(123)
fit_overfitted <- fmm(y = y, grid = grid, K = 50, amu = 0, b2mu = 100,
    asigma2 = 0.1, bsigma2 = 0.1, alpha = rep(0.1, 50), nsim = nsim)
# looking at the traceplots of the estimated densities (the
# quantities that we care about) for sake of space only for
# K=3
nburn <- 1000
ind <- sample(1:ngrid, 5, replace = TRUE)</pre>
qplot((nburn + 1):nsim, fit3$Dens[(nburn + 1):nsim, ind[1]],
geom = "line")
## Warning: 'qplot()' was deprecated in ggplot2 3.4.0.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```



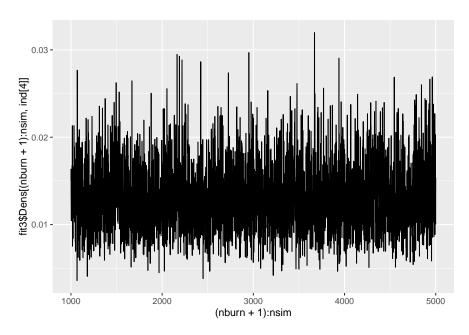
qplot((nburn + 1):nsim, fit3\$Dens[(nburn + 1):nsim, ind[2]],
 geom = "line")



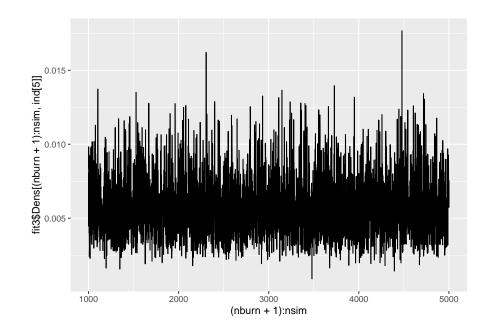
```
qplot((nburn + 1):nsim, fit3$Dens[(nburn + 1):nsim, ind[3]],
    geom = "line")
```



qplot((nburn + 1):nsim, fit3\$Dens[(nburn + 1):nsim, ind[4]],
 geom = "line")



```
qplot((nburn + 1):nsim, fit3$Dens[(nburn + 1):nsim, ind[5]],
    geom = "line")
```



```
effectiveSize(fit3$Dens[(nburn + 1):nsim, ind[1]])
##
       var1
## 3576.282
geweke.diag(fit3$Dens[(nburn + 1):nsim, ind[1]])
##
## Fraction in 1st window = 0.1
## Fraction in 2nd window = 0.5
##
##
     var1
## 0.2552
effectiveSize(fit3$Dens[(nburn + 1):nsim, ind[2]])
##
       var1
## 3126.878
geweke.diag(fit3$Dens[(nburn + 1):nsim, ind[2]])
##
## Fraction in 1st window = 0.1
## Fraction in 2nd window = 0.5
##
##
      var1
## -0.4439
effectiveSize(fit3$Dens[(nburn + 1):nsim, ind[3]])
```

##

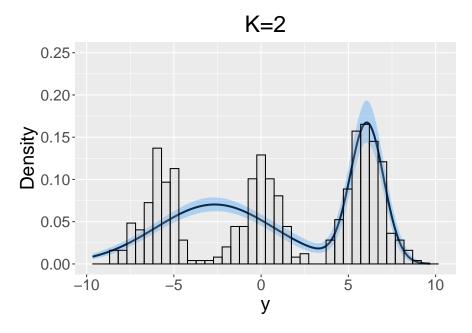
var1

## 2970.933

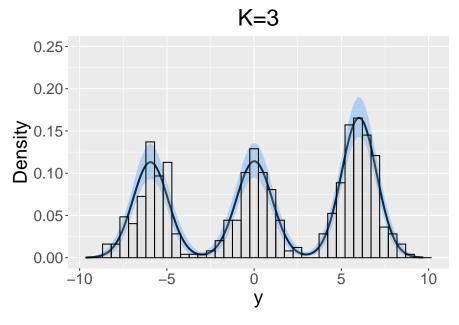
```
geweke.diag(fit3$Dens[(nburn + 1):nsim, ind[3]])
##
## Fraction in 1st window = 0.1
## Fraction in 2nd window = 0.5
##
##
      var1
## -0.0566
effectiveSize(fit3$Dens[(nburn + 1):nsim, ind[4]])
##
       var1
## 2847.535
geweke.diag(fit3$Dens[(nburn + 1):nsim, ind[4]])
##
## Fraction in 1st window = 0.1
## Fraction in 2nd window = 0.5
##
##
    var1
## 0.7302
effectiveSize(fit3$Dens[(nburn + 1):nsim, ind[5]])
       var1
## 3712.453
geweke.diag(fit3$Dens[(nburn + 1):nsim, ind[5]])
##
## Fraction in 1st window = 0.1
## Fraction in 2nd window = 0.5
##
## var1
## -1.95
# estimated densities
dens2m <- apply(fit2$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens21 <- apply(fit2$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.025)
dens2h <- apply(fit2$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.975)</pre>
dens3m <- apply(fit3$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens31 <- apply(fit3$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.025)</pre>
dens3h <- apply(fit3$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.975)</pre>
dens4m <- apply(fit4$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens41 <- apply(fit4$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.025)</pre>
dens4h <- apply(fit4$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.975)
```

```
dens5m <- apply(fit5$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens51 <- apply(fit5$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.025)</pre>
dens5h <- apply(fit5$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.975)</pre>
dens20m <- apply(fit20$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens201 <- apply(fit20$Dens[(nburn + 1):nsim, ], 2, quantile,</pre>
    prob = 0.025)
dens20h <- apply(fit20$Dens[(nburn + 1):nsim, ], 2, quantile,</pre>
    prob = 0.975)
dens50m <- apply(fit50$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens501 <- apply(fit50$Dens[(nburn + 1):nsim, ], 2, quantile,</pre>
    prob = 0.025)
dens50h <- apply(fit50$Dens[(nburn + 1):nsim, ], 2, quantile,</pre>
    prob = 0.975)
densoverfittedm <- apply(fit_overfitted$Dens[(nburn + 1):nsim,</pre>
    ], 2, mean)
densoverfittedl <- apply(fit_overfitted$Dens[(nburn + 1):nsim,</pre>
    ], 2, quantile, prob = 0.025)
densoverfittedh <- apply(fit_overfitted$Dens[(nburn + 1):nsim,</pre>
    ], 2, quantile, prob = 0.975)
dfhist <- data.frame(y = y)</pre>
dfdens2 <- data.frame(dm = dens2m, dl = dens2l, dh = dens2h,
    seggrid = grid)
dfdens3 <- data.frame(dm = dens3m, dl = dens3l, dh = dens3h,
    seqgrid = grid)
dfdens4 <- data.frame(dm = dens4m, dl = dens4l, dh = dens4h,
    seqgrid = grid)
dfdens5 <- data.frame(dm = dens5m, dl = dens5l, dh = dens5h,
    seggrid = grid)
dfdens20 <- data.frame(dm = dens20m, dl = dens20l, dh = dens20h,
    seggrid = grid)
dfdens50 <- data.frame(dm = dens50m, dl = dens50l, dh = dens50h,
    seggrid = grid)
dfdensoverfitted <- data.frame(dm = densoverfittedm, dl = densoverfittedl,
    dh = densoverfittedh, seqgrid = grid)
ggplot(dfdens2, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens2, aes(x = seggrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + xlab("y") + ylab("Density") +
    ylim(0, 0.25) + geom_histogram(data = dfhist, aes(x = y,
    y = after_stat(density)), alpha = 0.2, bins = 40, inherit.aes = FALSE,
    fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
    ggtitle("K=2") + theme(plot.title = element_text(hjust = 0.5))
## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
```

## generated.

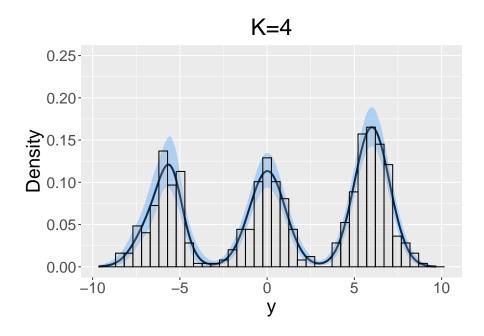


```
ggplot(dfdens3, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
   geom_ribbon(data = dfdens3, aes(x = seqgrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + xlab("y") + ylab("Density") +
   ylim(0, 0.25) + geom_histogram(data = dfhist, aes(x = y,
   y = after_stat(density)), alpha = 0.2, bins = 40, inherit.aes = FALSE,
   fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
   ggtitle("K=3") + theme(plot.title = element_text(hjust = 0.5))
```

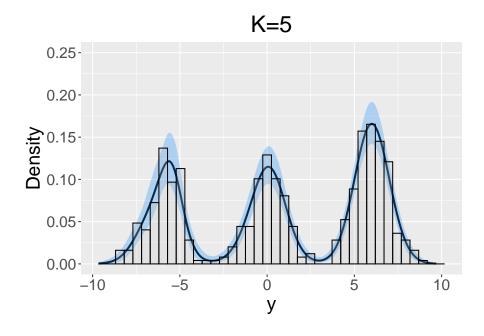


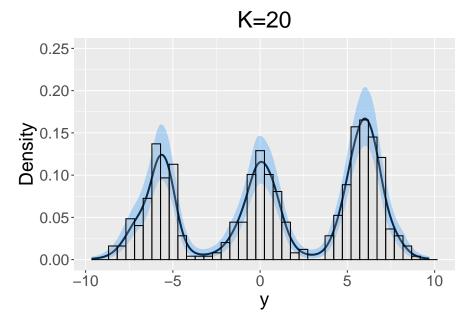
```
ggplot(dfdens4, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
   geom_ribbon(data = dfdens4, aes(x = seqgrid, ymin = dl, ymax = dh),
   alpha = 0.3, fill = "dodgerblue1") + xlab("y") + ylab("Density") +
```

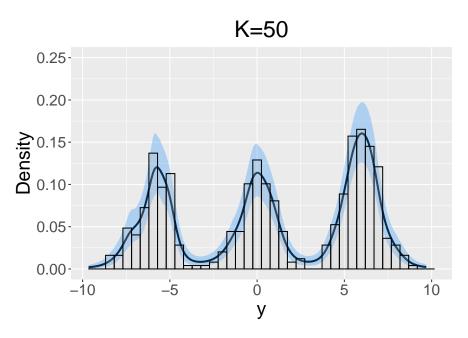
```
ylim(0, 0.25) + geom_histogram(data = dfhist, aes(x = y,
y = after_stat(density)), alpha = 0.2, bins = 40, inherit.aes = FALSE,
fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
ggtitle("K=4") + theme(plot.title = element_text(hjust = 0.5))
```

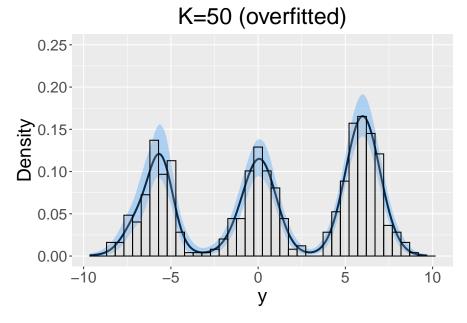


```
ggplot(dfdens5, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens5, aes(x = seqgrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + xlab("y") + ylab("Density") +
    ylim(0, 0.25) + geom_histogram(data = dfhist, aes(x = y,
    y = after_stat(density)), alpha = 0.2, bins = 40, inherit.aes = FALSE,
    fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
    ggtitle("K=5") + theme(plot.title = element_text(hjust = 0.5))
```









```
# function to calculate the CPO/LPML and WAIC
good_fit_criteria <- function(y, P, Mu, Sigma2, nburn) {
    nsim <- nrow(P)
    K <- ncol(Mu)
    n <- length(y)</pre>
```

```
Densy <- array(0, c((nsim - nburn), n, K))</pre>
    Densym <- matrix(0, nrow = (nsim - nburn), ncol = n)</pre>
    for (i in (nburn + 1):nsim) {
        for (k in 1:K) {
            Densy[i - nburn, , k] <- P[i, k] * dnorm(y, Mu[i,</pre>
                 k], sqrt(Sigma2[i, k]))
        }
        for (j in 1:n) {
            Densym[i - nburn, j] <- sum(Densy[i - nburn, j, ])</pre>
        }
    }
    cpoinv <- apply(1/Densym, 2, mean)</pre>
    cpo <- 1/cpoinv
    lpml <- sum(log(cpo))</pre>
    lpd <- sum(log(apply(exp(log(Densym)), 2, mean)))</pre>
    p2 <- sum(apply(log(Densym), 2, var))</pre>
    waic <- -2 * (1pd - p2)
    return(list(CPO = cpo, LPML = lpml, WAIC = waic))
}
ghc2 <- good_fit_criteria(y = y, P = fit2$P, Mu = fit2$Mu, Sigma2 = fit2$Sigma2,
    nburn = nburn)
ghc2$LPML
## [1] -1378.841
ghc2$WAIC
## [1] 2757.683
ghc3 <- good_fit_criteria(y = y, P = fit3$P, Mu = fit3$Mu, Sigma2 = fit3$Sigma2,</pre>
    nburn = nburn)
ghc3$LPML
## [1] -1260.408
ghc3$WAIC
## [1] 2520.809
ghc4 <- good_fit_criteria(y = y, P = fit4$P, Mu = fit4$Mu, Sigma2 = fit4$Sigma2,
    nburn = nburn)
ghc4$LPML
## [1] -1259.599
```

```
ghc4$WAIC
## [1] 2519.197
ghc5 <- good_fit_criteria(y = y, P = fit5$P, Mu = fit5$Mu, Sigma2 = fit5$Sigma2,
   nburn = nburn)
ghc5$LPML
## [1] -1259.899
ghc5$WAIC
## [1] 2519.768
ghc20 <- good_fit_criteria(y = y, P = fit20$P, Mu = fit20$Mu,</pre>
    Sigma2 = fit20$Sigma2, nburn = nburn)
ghc20$LPML
## [1] -1266.232
ghc20$WAIC
## [1] 2532.418
ghc50 <- good_fit_criteria(y = y, P = fit50$P, Mu = fit50$Mu,</pre>
    Sigma2 = fit50$Sigma2, nburn = nburn)
ghc50$LPML
## [1] -1278.796
ghc50$WAIC
## [1] 2557.563
ghcoverfitted <- good_fit_criteria(y = y, P = fit_overfitted$P,</pre>
    Mu = fit_overfitted$Mu, Sigma2 = fit_overfitted$Sigma2, nburn = nburn)
ghcoverfitted$LPML
## [1] -1263.77
ghcoverfitted$WAIC
## [1] 2527.51
```

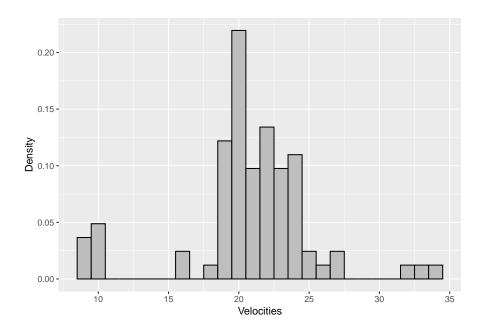
#### Galaxy data example

```
require(MASS)

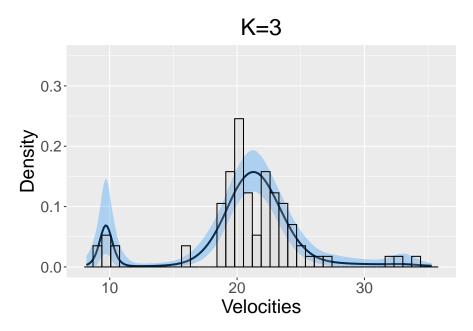
y <- galaxies/1000

ggplot(data.frame(y = y), aes(x = y)) + geom_histogram(aes(y = ..density..),
    binwidth = 1, fill = "gray", colour = "black") + xlab("Velocities") +
    ylab("Density")</pre>
```

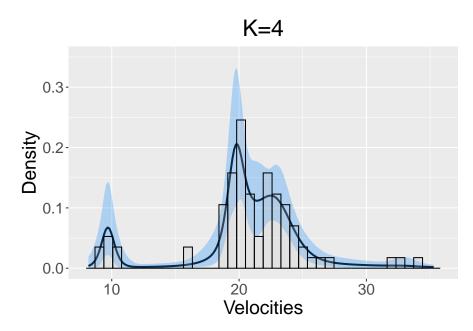
```
## Warning: The dot-dot notation ('..density..') was deprecated in ggplot2 3.4.0.
## i Please use 'after_stat(density)' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```



```
fit6 \leftarrow fmm(y = y, grid = grid, K = 6, amu = 0, b2mu = 100, asigma2 = 0.1,
    bsigma2 = 0.1, alpha = rep(1, 6), nsim = nsim)
set.seed(123)
fit_overfitted <- fmm(y = y, grid = grid, K = 50, amu = 0, b2mu = 100,
    asigma2 = 0.1, bsigma2 = 0.1, alpha = rep(0.1, 50), nsim = nsim)
dens3m <- apply(fit3$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens31 <- apply(fit3$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.025)</pre>
dens3h <- apply(fit3$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.975)</pre>
dens4m <- apply(fit4$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens41 <- apply(fit4$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.025)</pre>
dens4h <- apply(fit4$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.975)</pre>
dens5m <- apply(fit5$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens51 <- apply(fit5$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.025)</pre>
dens5h <- apply(fit5$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.975)
dens6m <- apply(fit6$Dens[(nburn + 1):nsim, ], 2, mean)</pre>
dens61 <- apply(fit6$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.025)
dens6h <- apply(fit6$Dens[(nburn + 1):nsim, ], 2, quantile, prob = 0.975)</pre>
densoverfittedm <- apply(fit_overfitted$Dens[(nburn + 1):nsim,</pre>
    ], 2, mean)
densoverfittedl <- apply(fit_overfitted$Dens[(nburn + 1):nsim,</pre>
    ], 2, quantile, prob = 0.025)
densoverfittedh <- apply(fit_overfitted$Dens[(nburn + 1):nsim,</pre>
    ], 2, quantile, prob = 0.975)
dfhist \leftarrow data.frame(y = y)
dfdens3 <- data.frame(dm = dens3m, dl = dens3l, dh = dens3h,
    seggrid = grid)
dfdens4 <- data.frame(dm = dens4m, dl = dens4l, dh = dens4h,
    seqgrid = grid)
dfdens5 <- data.frame(dm = dens5m, dl = dens5l, dh = dens5h,
    seggrid = grid)
dfdens6 <- data.frame(dm = dens6m, dl = dens6l, dh = dens6h,
    seggrid = grid)
dfdensoverfitted <- data.frame(dm = densoverfittedm, dl = densoverfittedl,
    dh = densoverfittedh, seqgrid = grid)
ggplot(dfdens3, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens3, aes(x = seqgrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + xlab("Velocities") +
    ylab("Density") + ylim(0, 0.35) + geom_histogram(data = dfhist,
    aes(x = y, y = after_stat(density)), alpha = 0.2, bins = 40,
    inherit.aes = FALSE, fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
    ggtitle("K=3") + theme(plot.title = element_text(hjust = 0.5))
```

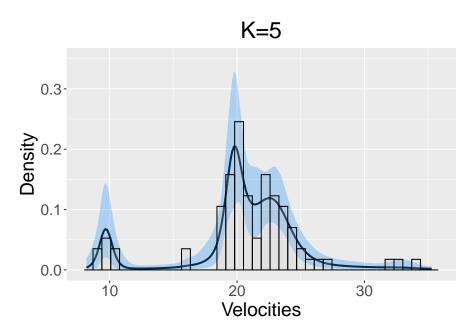


```
ggplot(dfdens4, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens4, aes(x = seqgrid, ymin = dl, ymax = dh),
    alpha = 0.3, fill = "dodgerblue1") + xlab("Velocities") +
    ylab("Density") + ylim(0, 0.35) + geom_histogram(data = dfhist,
    aes(x = y, y = after_stat(density)), alpha = 0.2, bins = 40,
    inherit.aes = FALSE, fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
    ggtitle("K=4") + theme(plot.title = element_text(hjust = 0.5))
```

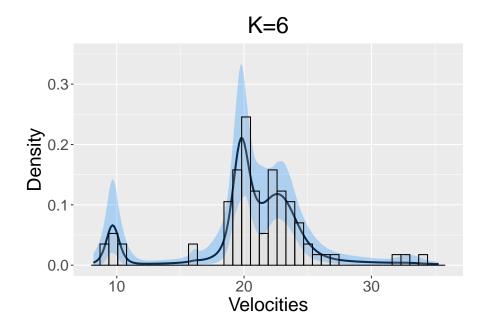


```
ggplot(dfdens5, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens5, aes(x = seqgrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + xlab("Velocities") +
    ylab("Density") + ylim(0, 0.35) + geom_histogram(data = dfhist,
    aes(x = y, y = after_stat(density)), alpha = 0.2, bins = 40,
```

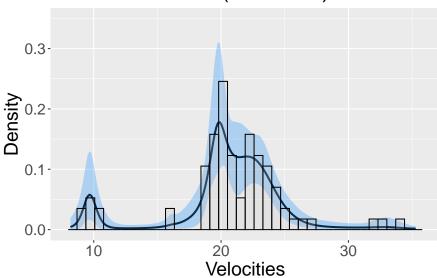
```
inherit.aes = FALSE, fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
ggtitle("K=5") + theme(plot.title = element_text(hjust = 0.5))
```



```
ggplot(dfdens6, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens6, aes(x = seqgrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + xlab("Velocities") +
    ylab("Density") + ylim(0, 0.35) + geom_histogram(data = dfhist,
    aes(x = y, y = after_stat(density)), alpha = 0.2, bins = 40,
    inherit.aes = FALSE, fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
    ggtitle("K=6") + theme(plot.title = element_text(hjust = 0.5))
```

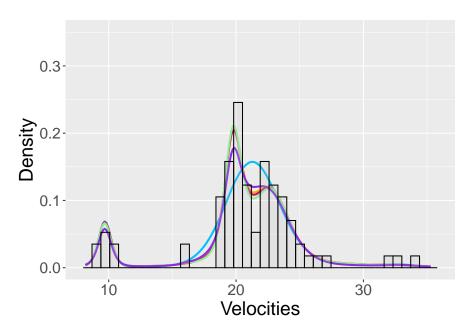


# K=50 (overfitted)



```
dfj <- data.frame(dm3 = dens3m, dm4 = dens4m, dm5 = dens5m, dm6 = dens6m,
    dmover = densoverfittedm, seqgrid = grid)

ggplot(dfj, aes(x = seqgrid, y = dm3)) + geom_line(size = 1,
    color = "deepskyblue1") + geom_line(aes(y = dm4), size = 1,
    color = "orange") + geom_line(aes(y = dm5), size = 1, color = "deeppink3") +
    geom_line(aes(y = dm6), size = 1, color = "lightgreen") +
    geom_line(aes(y = dmover), size = 1, color = "blueviolet") +
    xlab("Velocities") + ylab("Density") + ylim(0, 0.35) + geom_histogram(data = dfhist,
    aes(x = y, y = after_stat(density)), alpha = 0.2, bins = 40,
    inherit.aes = FALSE, fill = "gray", colour = "black") + theme(text = element_text(size = 20))</pre>
```



## [1] -214.3474

ghc3\$WAIC

## [1] 428.1348

## [1] -212.1646

ghc4\$WAIC

## [1] 423.9259

## [1] -212.8018

ghc5\$WAIC

## [1] 425.0935

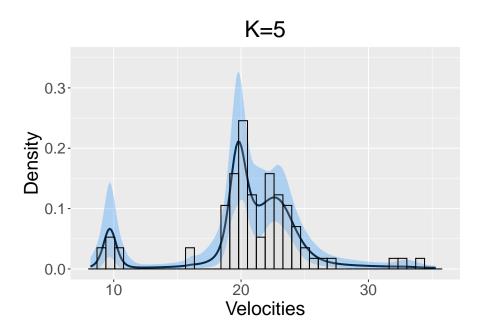
#### Finite mixture model in JAGS

## [1] 436.2959

The below code shows how to implement exactly the same mixture model but using JAGS. it has the advantage that one can be much more flexible in the prior distributions (or the mixture kernel), without having to recode everything. Once we extract the parameters, everything proceeds as before.

```
require(rjags)
model_string <- "model{</pre>
for(i in 1:n){
y[i]~dnorm(mu[z[i]],tau[z[i]])
z[i]~dcat(p[])
p[1:K]~ddirch(alpha)
for(k in 1:K){
mu[k]~dnorm(0,0.01)
tau[k]~dgamma(0.1,0.1)
sigma2[k] <- 1/tau[k]</pre>
}
}"
K <- 5
alpha \leftarrow rep(1, K)
data \leftarrow list(n = length(y), y = y, K = K, alpha = alpha)
model <- jags.model(textConnection(model_string), n.chains = 1,</pre>
 data = data, n.adapt = 2000)
```

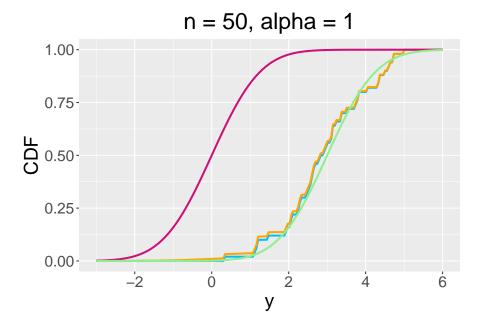
```
## Compiling model graph
##
      Resolving undeclared variables
##
      Allocating nodes
## Graph information:
##
      Observed stochastic nodes: 82
      Unobserved stochastic nodes: 93
##
      Total graph size: 356
##
##
## Initializing model
update(model, nburn)
res <- jags.samples(model, variable.names = c("mu", "sigma2",
    "p"), n.iter = nsim)
# extracting the parameters
P <- Mu <- Sigma2 <- matrix(0, nrow = nsim, ncol = K)
for (k in 1:K) {
    Mu[, k] \leftarrow res mu[k, , 1]
    Sigma2[, k] \leftarrow res$sigma2[k, , 1]
    P[, k] \leftarrow res p[k, , 1]
}
# computing the density (as we did before inside the fmm
# function)
dens <- array(0, c(nsim, ngrid, K))</pre>
dens1 <- matrix(0, nrow = nsim, ncol = ngrid)</pre>
for (i in 1:nsim) {
    for (k in 1:K) {
        dens[i, , k] <- P[i, k] * dnorm(grid, Mu[i, k], sqrt(Sigma2[i,</pre>
            k]))
    }
    for (j in 1:ngrid) {
        dens1[i, j] <- sum(dens[i, j, ])</pre>
    }
}
densm <- apply(dens1, 2, mean)</pre>
dens1 <- apply(dens1, 2, quantile, prob = 0.025)
densh <- apply(dens1, 2, quantile, prob = 0.975)
dfdens <- data.frame(dm = densm, dl = densl, dh = densh, seggrid = grid)
ggplot(dfdens, aes(x = seggrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens, aes(x = seqgrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + xlab("Velocities") +
    ylab("Density") + ylim(0, 0.35) + geom_histogram(data = dfhist,
    aes(x = y, y = after_stat(density)), alpha = 0.2, bins = 40,
    inherit.aes = FALSE, fill = "gray", colour = "black") + theme(text = element_text(size = 20)) +
    ggtitle("K=5") + theme(plot.title = element_text(hjust = 0.5))
```



## Dirichlet process prior

DP conjugacy - posterior mean

```
n <- 50
y <- rnorm(n, 3, 1)
alpha <- 1
Femp <- ecdf(y)</pre>
grid <- seq(-3, 6, len = 200)
ngrid <- length(grid)</pre>
post_mean <- numeric(ngrid)</pre>
for (j in 1:ngrid) {
    post_mean[j] \leftarrow (alpha/(alpha + n)) * pnorm(grid[j], 0, 1) +
        (n/(alpha + n)) * Femp(grid[j])
}
df <- data.frame(demp = Femp(grid), pm = post_mean, seqgrid = grid)</pre>
ggplot(df, aes(x = seqgrid, y = demp)) + geom_line(size = 1,
    color = "deepskyblue1") + geom_line(aes(y = pm), size = 1,
    color = "orange") + stat_function(fun = pnorm, args = list(mean = 0,
    sd = 1), size = 1, color = "deeppink3") + stat_function(fun = pnorm,
    args = list(mean = 3, sd = 1), size = 1, color = "lightgreen") +
    ylab("CDF") + xlab("y") + theme(text = element_text(size = 20)) +
    ggtitle("n = 50, alpha = 1") + theme(plot.title = element_text(hjust = 0.5))
```

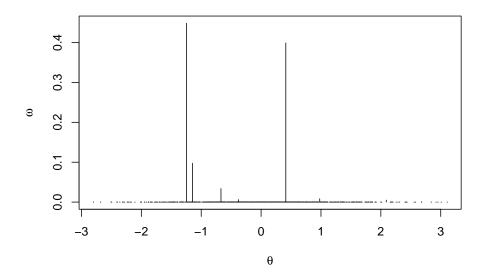


```
## Sethuraman representation
alpha <- 0.5
K <- 1000

# baseline (G_0) is standard normal distribution
theta <- rnorm(K, 0, 1)

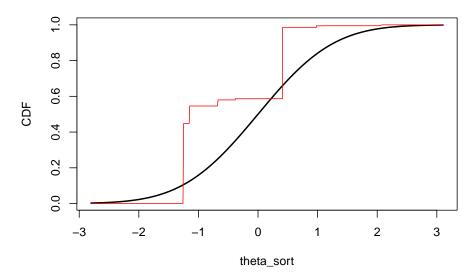
v <- rbeta(K, 1, alpha)
omega <- numeric(K)
omega[1] <- v[1]
cumv = cumprod(1 - v)
for (k in 2:(K - 1)) {
   omega[k] <- v[k] * cumv[k - 1]
}
omega[K] <- 1 - sum(omega[1:(K - 1)])

plot(theta, omega, type = "h", main = "", xlab = expression(theta),
   ylab = expression(omega))</pre>
```



```
theta_sort <- sort(theta)
omega_ordered <- omega[order(theta)]
fdist <- numeric(K)
for (k in 1:K) {
    fdist[k] <- cumsum(omega_ordered)[k]
}

plot(theta_sort, pnorm(theta_sort, 0, 1), type = "l", lwd = 2,
    ylab = "CDF")
lines(theta_sort, fdist, col = "red")</pre>
```

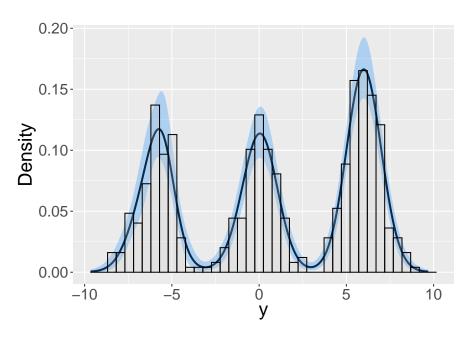


# Dirichlet process mixtures

### R code

```
n <- length(y)
p \leftarrow ns \leftarrow rep(0, L)
v \leftarrow rep(1/L, L)
v[L] <- 1
prop <- prob <- matrix(0, nrow = n, ncol = L)</pre>
z <- matrix(0, nrow = nsim, ncol = n)</pre>
z[1, ] \leftarrow rep(1, n)
P <- Mu <- Sigma2 <- matrix(0, nrow = nsim, ncol = L)
Mu[1, ] <- rep(mean(y), L)
Sigma2[1, ] <- rep(var(y), L)</pre>
for (i in 2:nsim) {
    cumv <- cumprod(1 - v)</pre>
    p[1] \leftarrow v[1]
    for (1 in 2:L) {
        p[1] <- v[1] * cumv[1 - 1]
    }
    for (1 in 1:L) {
         prop[, 1] \leftarrow p[1] * dnorm(y, mean = Mu[i - 1, 1],
             sd = sqrt(Sigma2[i - 1, 1]))
    prob <- prop/apply(prop, 1, sum)</pre>
    for (j in 1:n) {
        z[i, j] \leftarrow sample(1:L, size = 1, prob = prob[j, ])
    P[i, ] <- p
    for (1 in 1:L) {
         ns[1] \leftarrow length(which(z[i, ] == 1))
    for (1 in 1:(L - 1)) {
         v[1] \leftarrow rbeta(1, 1 + ns[1], alpha + sum(ns[(1 + 1):L]))
    }
    for (1 in 1:L) {
         varmu <- 1/((1/b2mu) + (ns[l]/Sigma2[i - 1, l]))</pre>
         meanmu <- ((sum(y[z[i, ] == 1])/Sigma2[i - 1, 1]) +
             (amu/b2mu))/((1/b2mu) + (ns[1]/Sigma2[i - 1,
         Mu[i, 1] <- rnorm(1, mean = meanmu, sd = sqrt(varmu))</pre>
         Sigma2[i, 1] \leftarrow 1/rgamma(1, asigma2 + ns[1]/2, bsigma2 +
             0.5 * sum((y[z[i, ] == 1] - Mu[i, 1])^2))
    }
}
res = list()
res$P = P[(nburn + 1):nsim,]
res$Mu = Mu[(nburn + 1):nsim, ]
```

```
res$Sigma2 = Sigma2[(nburn + 1):nsim, ]
    res$z = z[(nburn + 1):nsim,]
    return(res)
}
n <- 500
set.seed(123)
y \leftarrow rnormmix(n, c(0.3, 0.3, 0.4), c(-6, 0, 6), c(1, 1, 1))
grid \leftarrow seq(min(y) - 1, max(y) + 1, len = 200)
ngrid <- length(grid)</pre>
nsim <- 5000
nburn = 500
set.seed(123)
fitdpm \leftarrow dpm(y = y, amu = 0, b2mu = 100, asigma2 = 1, bsigma2 = 1,
    alpha = 1, L = 50, nsim = nsim, nburn = nburn)
L <- 50
p <- fitdpm$P
mu <- fitdpm$Mu
sigma2 <- fitdpm$Sigma2</pre>
niter <- nrow(p)</pre>
dens <- array(0, c(niter, ngrid, L))</pre>
dens1 <- matrix(0, nrow = niter, ncol = ngrid)</pre>
for (l in 1:niter) {
    for (k in 1:L) {
        dens[l, , k] \leftarrow p[l, k] * dnorm(grid, mu[l, k], sqrt(sigma2[l, k]))
            k]))
    }
    for (j in 1:ngrid) {
        dens1[1, j] <- sum(dens[1, j, ])
    }
}
densm <- apply(dens1, 2, mean)</pre>
densl <- apply(dens1, 2, quantile, prob = 0.025)</pre>
densh <- apply(dens1, 2, quantile, prob = 0.975)
dfhist <- data.frame(y = y)</pre>
dfdens <- data.frame(dm = densm, dl = densl, dh = densh, seggrid = grid)
ggplot(dfdens, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens, aes(x = seqgrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + geom_histogram(data = dfhist,
    aes(x = y, y = after_stat(density)), alpha = 0.2, bins = 40,
    inherit.aes = FALSE, fill = "gray", colour = "black") + xlab(expression(y)) +
    ylab("Density") + theme(text = element_text(size = 20))
```



## [1] -1260.859

ghcdpm\$WAIC

## [1] 2521.71

#### JAGS code

```
model_string = "
model {
 for (i in 1:n) {
   y[i] ~ dnorm(mu[z[i]], tau[z[i]])
    z[i] ~ dcat(pi[])
  }
  for (1 in 1:L) {
    mu[1] ~ dnorm(0, 0.01)
    tau[1] ~ dgamma(1, 1)
    sigma2[1] <- 1/tau[1]
  # Stick breaking
  for (l in 1:(L-1)) {v[l] ~ dbeta(1,alpha)}
  v[L] <- 1
  pi[1] <- v[1]
  for (1 in 2:L) {
    pi[l] <- v[l] * (1-v[l-1]) * pi[l-1] / v[l-1]
```

```
alpha ~ dgamma(2,2)
require(mixtools)
n <- 500
set.seed(123)
y \leftarrow rnormmix(n, c(0.3, 0.3, 0.4), c(-6, 0, 6), c(1, 1, 1))
grid \leftarrow seq(min(y) - 1, max(y) + 1, len = 200)
ngrid <- length(grid)</pre>
L <- 50
data \leftarrow list(n = length(y), y = y, L = L)
model <- jags.model(textConnection(model_string), n.chains = 1,</pre>
data = data)
## Compiling model graph
##
      Resolving undeclared variables
##
      Allocating nodes
## Graph information:
      Observed stochastic nodes: 500
##
##
      Unobserved stochastic nodes: 650
##
      Total graph size: 2354
##
## Initializing model
update(model, nburn)
res <- jags.samples(model, variable.names = c("mu", "sigma2",
    "pi", "z"), n.iter = nsim)
P <- Mu <- Sigma2 <- matrix(0, nrow = nsim, ncol = L)
for (1 in 1:L) {
    Mu[, 1] <- res$mu[1, , 1]
    Sigma2[, 1] <- res$sigma2[1, , 1]
    P[, 1] \leftarrow res pi[1, , 1]
}
grid \leftarrow seq(min(y) - 1, max(y) + 1, len = 200)
ngrid <- length(grid)</pre>
dens <- array(0, c(nsim, ngrid, L))</pre>
dens1 <- matrix(0, nrow = nsim, ncol = ngrid)</pre>
for (i in 1:nsim) {
    for (1 in 1:L) {
        dens[i, , 1] <- P[i, 1] * dnorm(grid, Mu[i, 1], sqrt(Sigma2[i,</pre>
             1]))
    }
    for (j in 1:ngrid) {
        dens1[i, j] <- sum(dens[i, j, ])</pre>
    }
}
densm <- apply(dens1, 2, mean)</pre>
```

```
densl <- apply(dens1, 2, quantile, prob = 0.025)
densh <- apply(dens1, 2, quantile, prob = 0.975)

dfhist <- data.frame(y = y)
dfdens <- data.frame(dm = densm, dl = densl, dh = densh, seqgrid = grid)

ggplot(dfdens, aes(x = seqgrid, y = dm)) + geom_line(size = 1) +
    geom_ribbon(data = dfdens, aes(x = seqgrid, ymin = dl, ymax = dh),
        alpha = 0.3, fill = "dodgerblue1") + geom_histogram(data = dfhist,
    aes(x = y, y = after_stat(density)), alpha = 0.2, bins = 40,
    inherit.aes = FALSE, fill = "gray", colour = "black") + xlab(expression(y)) +
    ylab("Density") + theme(text = element_text(size = 20))</pre>
```

