Version 2: 2 clusters; Fix cluster label switching u,v; unif; fix w [0,1]; only force ordering of means in 1 sample; 5 chains

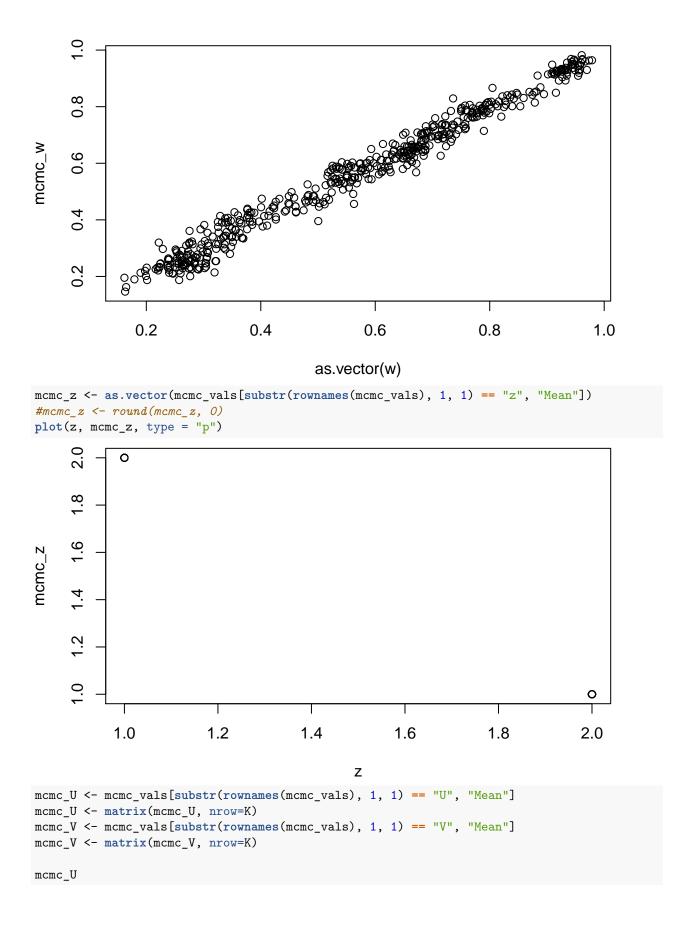
Simulate data

```
I <- 50
K <- 2
S <- 10
# choose diffuse priors for gamma
a gamma <- 2
b_gamma <- 100
avrg <- a_gamma * b_gamma
std.dv <- sqrt(a_gamma*b_gamma^2)</pre>
g_range = seq(0, avrg + 5*std.dv, 0.01)
g_y = dgamma(g_range, a_gamma, rate = 1/b_gamma)
\#plot(g\_range, g\_y, type = "l", ylim=c(0, max(g\_y) + 0.01))
set.seed(123)
a <- matrix(NA, nrow=K, ncol=S)
b <- matrix(NA, nrow=K, ncol=S)
for (s in 1:S) {
  a[, s] <- rgamma(K, a_gamma, rate = 1/b_gamma)
  b[, s] <- rgamma(K, a_gamma, rate = 1/b_gamma)
pi <- as.vector(rdirichlet(1, rep(1, K)))</pre>
z <- sample(1:K, size = I, replace = T, prob = pi)
w <- matrix(NA, nrow=I, ncol=S)
for (s in 1:S) {
  w[, s] \leftarrow rbeta(I, a[,s][z], b[,s][z])
tcn <- matrix(2, nrow=I, ncol=S)</pre>
m <- matrix(rep(sample(1:2, size = I, replace = T), S), nrow=I, ncol=S)
calcTheta <- function(m, tcn, w) {</pre>
  (m * w) / (tcn * w + 2*(1-w))
theta <- calcTheta(m, tcn, w)
n <- replicate(S, rpois(I, 100))</pre>
y <- matrix(NA, nrow=I, ncol=S)
for (i in 1:I) {
for (s in 1:S) {
```

```
y[i, s] <- rbinom(1, n[i, s], theta[i,s])
}</pre>
```

JAGS

```
jags.file <- file.path(models.dir, "v2_uv_unif_fix2.jags")</pre>
test.data <- list("I" = I, "S" = S, "K" = K,
                   "y" = y, "n" = n,
                   "m" = m, "tcn" = tcn)
jags.m <- jags.model(jags.file, test.data,</pre>
                       n.chains = numChains)
## Compiling model graph
##
      Resolving undeclared variables
##
      Allocating nodes
## Graph information:
##
      Observed stochastic nodes: 500
##
      Unobserved stochastic nodes: 591
##
      Total graph size: 8400
##
## Initializing model
params <- c("z", "w", "a", "b", "U", "V")
update(jags.m, n.iter=1000)
samps <- coda.samples(jags.m, params, n.iter=1000)</pre>
s <- summary(samps)</pre>
#effectiveSize(window(samps, start=5001))
pdf(file.path(trace.dir, paste0(runName, "_trace.pdf")))
plot(samps)
dev.off()
## pdf
##
mcmc_vals <- s$statistics</pre>
mcmc_w <- mcmc_vals[substr(rownames(mcmc_vals), 1, 1) == "w", "Mean"]</pre>
plot(as.vector(w), mcmc_w, type = "p")
```



```
[,1]
                       [,2]
                                  [,3]
                                            [,4]
                                                      [,5]
                                                                 [,6]
                                                                           [,7]
## [1,] 0.6430842 0.9421721 0.5635048 0.7008812 0.4683626 0.2466503 0.3576672
## [2,] 0.8390148 0.9185064 0.6097244 0.2675663 0.6738264 0.2383869 0.2775914
##
             [,8]
                       [,9]
                                 [,10]
## [1,] 0.7963742 0.7422414 0.4127624
## [2,] 0.5440963 0.2675650 0.5836539
p \leftarrow seq(0, 1, length = 100)
for (s in 1:S) {
  for (k in 1:K) {
    if (k == 1) {
      # plot mcmc mean U, V
      plot(p, dbeta(p, mcmc_U[k,s] * mcmc_V[k,s], (1-mcmc_U[k,s])*mcmc_V[k,s]),
           main = paste0("S", s),
           ylab = "density", xlab = "w", type = "l", col = k,
           ylim = c(0, 12)
      # plot truth
      lines(p, dbeta(p, a[k,s], b[k,s]), type = "l", col = k, lty=2)
      # add legend
      legend(x = "topleft",
             legend = pasteO(c("mcmc k", "true k"), rep(1:K, each=2)),
             col = rep(1:K, each=2),
             lty = rep(1:2, K),
             cex=0.8)
    } else {
      # plot mcmc mean U, V
      lines(p, dbeta(p, mcmc_U[k,s] * mcmc_V[k,s], (1-mcmc_U[k,s])*mcmc_V[k,s]),
            type = "1", col = k)
      # plot truth
      lines(p, dbeta(p, a[k,s], b[k,s]), type = "1", col = k, lty=2)
    }
  }
}
```

