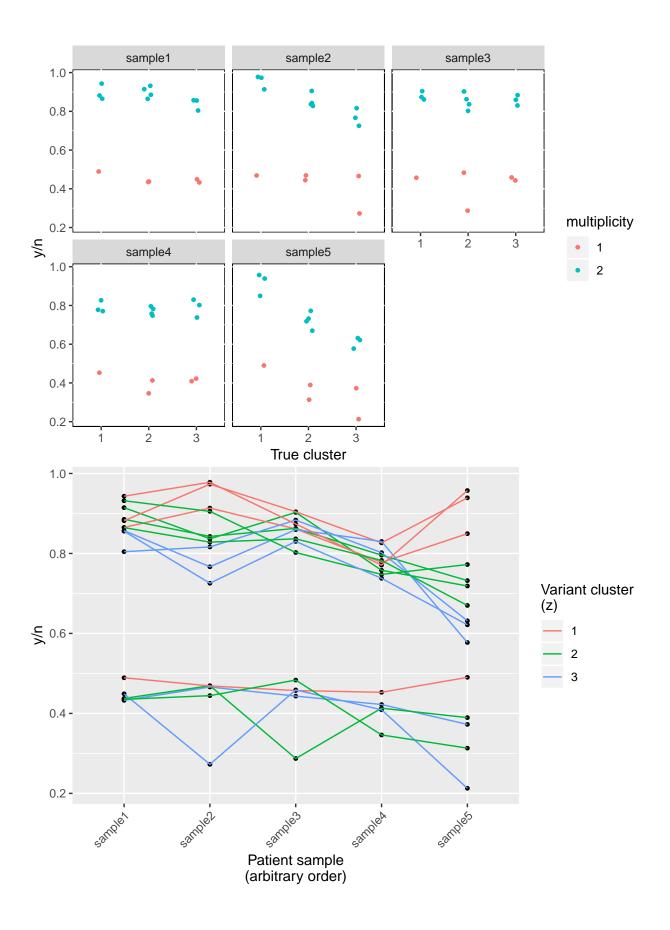
Version 5: 5 samples, 15 variants; truth = 3 clusters, linear

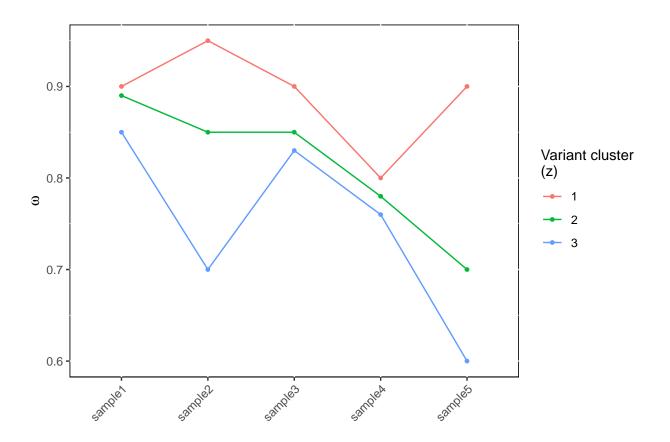
Simulate data

```
I <- 15
K <- 3
S <- 5
set.seed(123)
pi \leftarrow c(0.2, 0.5, 0.3)
z <- sample(1:K, size = I, replace = T, prob = pi)
w \leftarrow matrix(c(0.9, 0.95, 0.9, 0.8, 0.9,
               0.89, 0.85, 0.85, 0.78, 0.7,
               0.85, 0.7, 0.83, 0.76, 0.6),
             byrow=T,
             nrow=K, ncol=S)
\# w \leftarrow matrix(c(0.9, 0.95, 0.9, 0.8, 0.9,
                 0.7, 0.8, 0.85, 0.78, 0.7,
#
                 0.7, 0.2, 0.83, 0.76, 0.5),
#
               byrow=T,
               nrow=K, ncol=S)
colnames(w) <- paste0("sample", 1:S)</pre>
tcn <- matrix(2, nrow=I, ncol=S)</pre>
m <- matrix(rep(sample(1:2, size = I, replace = T), S),</pre>
             nrow=I, ncol=S)
W \leftarrow w[z,]
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])
}
test.data <- list("I" = I, "S" = S, "K" = K,
                    "y" = y, "n" = n,
                   "m" = m, "tcn" = tcn)
```

Visualize densities of simulated data

Clustering is by ω





functions

```
runMCMC <- function(data, K, jags.file, inits, params, n.iter, thin) {</pre>
  data$K <- K
  jags.m <- jags.model(jags.file, data,</pre>
                          n.chains = 1,
                          inits = inits,
                          n.adapt = 1000)
  samps <- coda.samples(jags.m, params, n.iter=n.iter, thin=thin)</pre>
  samps
}
getParamChain <- function(samps, param) {</pre>
  chains <- do.call(rbind, samps)</pre>
  chain <- chains[, grep(param, colnames(chains))]</pre>
}
reshapeW <- function(w, S, K) {</pre>
  w.mat <- matrix(w, nrow = K)</pre>
  colnames(w.mat) <- paste0("sample", 1:S)</pre>
  w.mat
}
calcLogLik <- function(z.iter, w.iter, data) {</pre>
  W <- w.iter[z.iter, ]</pre>
  theta <- calcTheta(data$m, data$tcn, W)</pre>
```

```
sum(dbinom(data$y, data$n, theta, log=T))
}

calcChainLogLik <- function(samps, data, K) {
    z.chain <- getParamChain(samps, "z")
    w.chain <- getParamChain(samps, "w")
    lik <- c()
    for(iter in 1:nrow(z.chain)) {
        z.iter <- z.chain[iter,]
        w.iter <- reshapeW(w.chain[iter,], data$S, K)
        lik <- c(lik, calcLogLik(z.iter, w.iter, data))
    }
    mean(lik)
}

calcBIC <- function(n, k, ll) log(n)*k - 2*11</pre>
```

JAGS

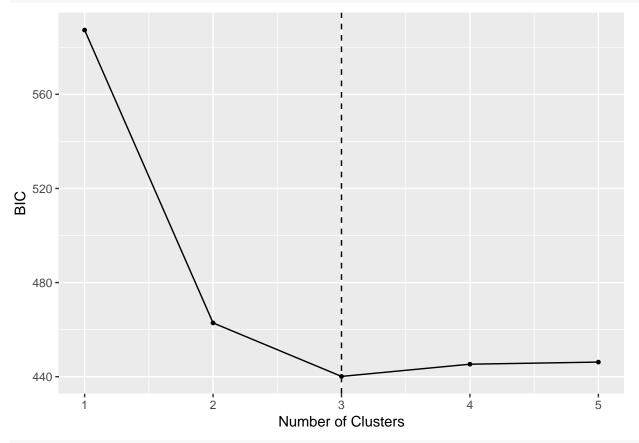
```
jags.files <- c(file.path(models.dir, "w.jags"),</pre>
                 file.path(models.dir, "w_K1.jags"))
inits <- list(".RNG.name" = "base::Wichmann-Hill",</pre>
                ".RNG.seed" = 123)
test.data \leftarrow list("I" = I, "S" = S,
                    "y" = y, "n" = n,
                    "m" = m, "tcn" = tcn)
params <- c("z", "w", "ystar")</pre>
n.iter = 10000
thin = 7
kToTest <- 1:5
BIC \leftarrow c()
samps.list <- list()</pre>
for(ix in 1:length(kToTest)) {
  if(kToTest[ix] == 1) {
    jags.file <- jags.files[2]</pre>
  } else {
    jags.file <- jags.files[1]</pre>
  K <- kToTest[ix]</pre>
  samps <- runMCMC(test.data, K, jags.file, inits, params, n.iter, thin)</pre>
  z.chain <- getParamChain(samps, "z")</pre>
  w.chain <- getParamChain(samps, "w")</pre>
  bic <- calcBIC(length(test.data$y), K, calcChainLogLik(samps, test.data, K))</pre>
  BIC <- c(BIC, bic)
  samps.list[[ix]] <- samps</pre>
```

Warning in jags.model(jags.file, data, n.chains = 1, inits = inits, n.adapt

```
## = 1000): Unused variable "K" in data
## Compiling model graph
##
      Resolving undeclared variables
      Allocating nodes
##
## Graph information:
      Observed stochastic nodes: 75
##
##
      Unobserved stochastic nodes: 80
##
      Total graph size: 460
##
## Initializing model
##
  Compiling model graph
##
      Resolving undeclared variables
##
      Allocating nodes
##
  Graph information:
      Observed stochastic nodes: 75
##
      Unobserved stochastic nodes: 101
##
##
      Total graph size: 1183
##
## Initializing model
##
##
  Compiling model graph
      Resolving undeclared variables
##
##
      Allocating nodes
## Graph information:
##
      Observed stochastic nodes: 75
##
      Unobserved stochastic nodes: 106
      Total graph size: 1188
##
##
## Initializing model
##
##
  Compiling model graph
      Resolving undeclared variables
##
##
      Allocating nodes
## Graph information:
      Observed stochastic nodes: 75
##
##
      Unobserved stochastic nodes: 111
##
      Total graph size: 1193
##
## Initializing model
##
## Compiling model graph
##
      Resolving undeclared variables
##
      Allocating nodes
##
  Graph information:
##
      Observed stochastic nodes: 75
      Unobserved stochastic nodes: 116
##
##
      Total graph size: 1198
##
## Initializing model
names(BIC) <- paste0("K", kToTest)</pre>
BIC
```

```
## 587.3050 462.8531 440.1249 445.3188 446.2112
best <- which.min(BIC)
s1 <- samps.list[[best]]
bestK <- kToTest[best]

ggplot(data.frame(numClust = kToTest, BIC), aes(x=numClust, y=BIC)) +
    geom_point(size=1) +
    geom_line() +
    xlab("Number of Clusters") +
    geom_vline(xintercept=bestK, linetype="dashed")</pre>
```



s1.w <- getParamChain(s1, "w")</pre>

##

K1

K2

КЗ

K4

