

# Package: BayesCPclust (via r-universe)

January 29, 2025

**Title** A Bayesian Approach for Clustering Constant-Wise Change-Point Data

**Version** 0.1.0

**Description** A Gibbs sampler algorithm was developed to estimate change points in constant-wise data sequences while performing clustering simultaneously. The algorithm is described in da Cruz, A. C. and de Souza, C. P. E. "A Bayesian Approach for Clustering Constant-wise Change-point Data" <doi:10.48550/arXiv.2305.17631>.

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data

*Error free data for all examples.*

---

### Description

A dataset generated for exemplification of Gibbs sampler using the model proposed in the paper "BayesCPclust: A Bayesian Approach for Clustering Constant-Wise Change-Point Data". The generation process is described in the paper with  $N = 5$ ,  $M = 50$ ,  $w = 10$ ,  $d = 2$ ,  $K = 2$ .

### Usage

data

### Format

A matrix with 50 rows and 5 columns

### References

A.C. da Cruz, C.P.E. de Souza. "BayesCPclust: A Bayesian Approach for Clustering Constant-Wise Change-Point Data" arXiv, arXiv:2305.17631v3 .

---

data_a	<i>Error free data for all examples.</i>
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---

**Description**

A dataset generated for exemplification of Gibbs sampler using the model proposed in the paper "BayesCPclust: A Bayesian Approach for Clustering Constant-Wise Change-Point Data". The generation process is described in the paper with  $N = 5$ ,  $M = 50$ ,  $w = 10$ ,  $d = 2$ ,  $K = 2$ .

**Usage**

data\_a

**Format**

A list with three components: a matrix with 50 rows and 5 columns, a vector with the cluster assignments, a vector with variance components

**References**

A.C. da Cruz, C.P.E. de Souza. "BayesCPclust: A Bayesian Approach for Clustering Constant-Wise Change-Point Data" arXiv, arXiv:2305.17631v3 .

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full_cond	<i>Full conditional for lambda</i>
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---

**Description**

Full conditional for lambda

**Usage**

full\_cond(kstar, lambda, cluster, a1, b1, K, N)

**Arguments**

kstar	A scalar with the number maximum of change points in all clusters
lambda	A scalar defining the parameter for the Truncate Poisson distribution that controls the number of change points (or its initial values)
cluster	A vector containing the cluster assignments for the data sequences (or its initial values)
a1	The hyperparameter value for the shape parameter in the gamma prior for lambda
b1	The hyperparameter value for the scale parameter in the gamma prior for lambda
K	A vector containing the number of change points for each cluster (or its initial values)
N	A scalar representing the number of data sequences

**Value**

'full\_cond' returns a numerical value corresponding to a sample from the full conditional for lambda

**Note**

This function is used within the Gibbs sampler, it is not expected to be used alone.

**Examples**

```
# Using hypothetical values to exemplification purposes
clusters <- c(1,1,2,1,2)
full_cond(kstar = 2, lambda = 3, cluster = clusters, al = 2, bl = 1000, K = c(2, 2), N = 5)
```

---

gibbs\_alg

*Gibbs sampler algorithm for simulated scenarios or real datasets*

---

**Description**

Gibbs sampler algorithm for simulated scenarios or real datasets

**Usage**

```
gibbs_alg(  
  N,  
  w,  
  M,  
  K,  
  Tl,  
  cluster,  
  alpha,  
  sigma2,  
  bs = 1000,  
  as = 2,  
  al = 2,  
  bl = 1000,  
  a = 2,  
  b = 1000,  
  alpha0 = 1/100,  
  kstar,  
  lambda,  
  Y,  
  d,  
  maxIter = 10000  
)
```

**Arguments**

N	A scalar representing the number of observations
w	A scalar representing the minimum number of points in each interval between two change points
M	A scalar representing the number of points available for each observation
K	A vector containing the number of change points for each cluster (or its initial values)
Tl	A list containing a vector for each cluster determining the change-point positions in each cluster (or its initial values)
cluster	A vector containing the cluster assignments for the observations (or its initial values)
alpha	A list containing a vector for each cluster determining the constant level values for each interval between change points in each cluster (or its initial values)
sigma2	A vector with the variances of observations (or its initial values)
bs	The hyperparameter value for the scale parameter in the inverse-gamma prior for the variance component
as	The hyperparameter value for the shape parameter in the inverse-gamma prior for the variance component
a1	The hyperparameter value for the shape parameter in the gamma prior for lambda
b1	The hyperparameter value for the scale parameter in the gamma prior for lambda
a	The hyperparameter value for the shape parameter in the gamma prior for alpha0
b	The hyperparameter value for the scale parameter in the gamma prior for alpha0
alpha0	A scalar defining the parameter for the Dirichlet process prior that controls the number of clusters (or its initial values)
kstar	A scalar with the number maximum of change points in all clusters
lambda	A scalar defining the parameter for the Truncate Poisson distribution that controls the number of change points (or its initial values)
Y	A matrix M x N with the data sequences
d	A scalar representing the number of clusters.
maxIter	A scalar for the number of iteration to run in the Gibbs sampler

**Value**

A list with each component representing the estimates for each iteration of the Gibbs sampler for each parameter

**See Also**

[run\_gibbs()]

**Examples**

```

data(data)
# initial values for each paramter and each cluster
par.values <- list(K = c(0, 0), T1 = list(50, 50), alpha = list(5, 10))
#cluster assignment for each data sequence
cluster <- kmeans(t(data), 2)$cluster
# variance for each data sequence
sigma2 <- apply(data, 2, var)
res <- gibbs_alg(alpha0 = 1/100, N = 5, w = 10, M = 50, K = par.values$K,
T1 = par.values$T1, cluster = cluster, alpha = par.values$alpha, sigma2 = sigma2,
bs = 1000, as = 2, al = 2, bl = 1000, a = 2, b = 1000, kstar = 2, lambda = 2,
Y = data, d = 2, maxIter = 10)

```

---

logsumexp

*Transfor a vector with over- or underflow*


---

**Description**

Transfor a vector with over- or underflow

**Usage**

```
logsumexp(x, min_x = Inf)
```

**Arguments**

x	A vector with numbers
min_x	A numerical value to represent the minimum value to perform comparison with the actual minimum value of 'x'

**Value**

'logsumexp' returns each element of the vector 'x' transformed using the Log-Sum-Exp trick.

**Examples**

```

# Transforming all elements in a vector using the Log-Sum-Exp trick
x <- c(1, 2, 3, 4, 5, 6)
logsumexp(x)

```

---

Mode *Compute the mode of a numerical vector*

---

**Description**

Compute the mode of a numerical vector

**Usage**

```
Mode(x)
```

**Arguments**

x                    A vector with numbers

**Value**

‘Mode‘ returns a value representing the most frequent numerical value in the vector ‘x‘

**Examples**

```
# Finding the mode of a vector of numbers
x <- c(1, 2, 2, 3, 5, 8, 10)
Mode(x)
```

---

pk *Probability mass function for truncated poisson*

---

**Description**

Probability mass function for truncated poisson

**Usage**

```
pk(k, kstar, lambda)
```

**Arguments**

k                    A scalar for the number of changes points in a cluster  
kstar                A scalar with the number maximum of change points in all clusters  
lambda               A scalar defining the parameter for the Truncate Poisson distribution that controls the number of change points (or its initial values)

**Value**

‘pk‘ returns a numerical value representing the marginal probability for a given k

**Note**

This function is used within the Gibbs sampler, it is not expected to be used alone.

**See Also**

[gibbs\_alg()]

**Examples**

```
# Hypothetical values
pk(k = 2, kstar = 3, lambda = 2)
```

---

possigma2n

*Full conditional function for sigma2*

---

**Description**

Full conditional function for sigma2

**Usage**

```
possigma2n(as, bs, M, Yn, k, Tln, alphan)
```

**Arguments**

as	The hyperparameter value for the shape parameter in the inverse-gamma prior for the variance component
bs	The hyperparameter value for the scale parameter in the inverse-gamma prior for the variance component
M	A scalar representing the number of points available for each data sequence
Yn	A vector or matrix with data sequences for a cluster
k	A scalar for the number of changes points in a cluster
Tln	A vector with the change-point positions for a cluster
alphan	A vector with the constant level values for each interval between change points for a cluster

**Value**

A numerical value corresponding to a sampled value from the full conditional of the variance component

**Note**

This function is called within the Gibbs sampler, but it can be used separately as well.



**See Also**

[gibbs\_alg()]

**Examples**

```
data(data)
possigma2n(as = 2, bs = 1000, M = 50, Yn = data[,1], k = 0, Tln = 50, alphan = 15)
```

---

postalpha0

*Posterior for alpha0*

---

**Description**

Posterior for alpha0

**Usage**

```
postalpha0(alpha0, a, b, N, cluster)
```

**Arguments**

alpha0	A scalar defining the parameter for the Dirichlet process prior that controls the number of clusters (or its initial values)
a	The hyperparameter value for the shape parameter in the gamma prior for alpha0
b	The hyperparameter value for the scale parameter in the gamma prior for alpha0
N	A scalar representing the number of data sequences
cluster	A vector containing the cluster assignments for the data sequences (or its initial values)

**Value**

A numerical value corresponding to a sample from the posterior of alpha0

**Note**

This function is called within the Gibbs sampler, but it can be called separately.

**Examples**

```
postalpha0(alpha0 = 1/100, a = 2, b = 1000, N = 5, cluster = c(1,1,2,1,1))
```

---

 postalphak

*Full conditional for alphak*


---

**Description**

Full conditional for alphak

**Usage**

```
postalphak(M, Y, sigma2, K, Tl, cluster, clusteri)
```

**Arguments**

M	A scalar representing the number of points available for each data sequence
Y	A matrix M x N with the data sequences
sigma2	A vector with the variances of the data sequences (or its initial values)
K	A vector containing the number of change points for each cluster (or its initial values)
Tl	A list containing a vector for each cluster determining the change-point positions in each cluster (or its initial values)
cluster	A vector containing the cluster assignments for the data sequences (or its initial values)
clusteri	A scalar with the index of a cluster

**Value**

A numerical vector of size 'K' + 1 with sampled values from the full conditional of alphak for a given cluster 'clusteri'

**Note**

This function is called within the Gibbs sampler, but it can be called separately as well.

**See Also**

[gibbs\_alg()]

**Examples**

```
data(data)
postalphak(M = 50, Y = data, sigma2 = 0.05, K = c(0, 0), Tl = c(50, 50),
  cluster = c(1,1,2,1,2), clusteri = 1)
```

---

postK                      *Marginal probability of K*

---

**Description**

Marginal probability of K

**Usage**

```
postK(kstar, w, M, Y, cluster, sigma2, lambda, clusteri)
```

**Arguments**

kstar	A scalar with the number maximum of change points in all clusters
w	A scalar representing the minimum number of points in each interval between two change points
M	A scalar representing the number of points available for each data sequence
Y	A matrix M x N with the data sequences
cluster	A vector containing the cluster assignments for the data sequences (or its initial values)
sigma2	A vector with the variances of the data sequences (or its initial values)
lambda	A scalar defining the parameter for the Truncate Poisson distribution that controls the number of change points (or its initial values)
clusteri	A scalar with the index of a cluster

**Value**

A numerical value corresponding to the sampled number of change points, k, for a given cluster

**Note**

This function is called within the Gibbs sampler, but it can also be called separately.

**See Also**

[gibbs\_alg()]

**Examples**

```
postK(kstar = 2, w = 10, M = 50, Y = data, cluster = c(1,1,2,1,2),
sigma2 = apply(data, 2, var), lambda = 2, clusteri = 1)
```

---

postK\_mk                      *Marginal probability of K per bin*

---

**Description**

Marginal probability of K per bin

**Usage**

```
postK_mk(k, m0, w, M, Yn, sigma2n, cellsn, mk, Cr)
```

**Arguments**

k	A scalar for the number of changes points in a cluster
m0	A scalar for the number of positions available to define change-points positions
w	A scalar representing the minimum number of points in each interval between two change points
M	A scalar representing the number of points available for each data sequence
Yn	A vector or matrix with data sequences for a cluster
sigma2n	A vector with the variance of the data sequences in a cluster
cellsn	A vector with the indices of the data sequences in a cluster
mk	A matrix with all possible values to distribute between change points
Cr	A scalar with the number of data sequences in a cluster

**Value**

‘postK\_mk’ returns a numerical value representing the non-normalized probability for a given bin, given k, and a given cluster

**Note**

This function is called within [postK()]. It should not be called alone.

**See Also**

[postK()], [gibbs\_alg()]

**Examples**

```
data(data)
M <- 50; k <- 0; w <- 10;
m0 <- M - 1 - (k+1)*w
for(k in 0:2){
mk <- RcppAlgos::permuteGeneral(0:m0, k + 1,
constraintFun = "sum",
comparisonFun = "==", limitConstraints = m0,
```

```

repetition = TRUE)}
out <- postK_mk(k = 0, m0 = m0, w = 10, M = 50, Yn = data[,c(1,2,4)],
  sigma2n = rep(0.05, 3), cellsn = c(1,2,4), mk = mk[1,], Cr = 3)

```

---

postmk

---

*Marginal probability of  $m_1, m_2, m_3, \dots, m_{k+1}$* 


---

### Description

Marginal probability of  $m_1, m_2, m_3, \dots, m_{k+1}$

### Usage

```
postmk(w, M, Y, K, cluster, sigma2, clusteri)
```

### Arguments

w	A scalar representing the minimum number of points in each interval between two change points
M	A scalar representing the number of points available for each data sequence
Y	A matrix $M \times N$ with the data sequences
K	A vector containing the number of change points for each cluster (or its initial values)
cluster	A vector containing the cluster assignments for the data sequences (or its initial values)
sigma2	A vector with the variances of the data sequences (or its initial values)
clusteri	A scalar with the index of a cluster

### Value

A numerical vector of size  $k + 1$  with the sampled number of observations (or bin size,  $m_k$ ) between each change point for a given cluster

### Note

This function is called within the Gibbs sampler, but it can also be called separately.

### Examples

```

data(data)
postmk(w = 10, M = 50, Y = data, K = c(1, 1), cluster = c(2,1,1,1,1), sigma2 = apply(data, 2, var),
  clusteri = 1)

```

---

qn0 *Mixing probability for creating new cluster*

---

**Description**

Mixing probability for creating new cluster

**Usage**

qn0(alpha0, w, N, M, bs, as, kstar, lambda, Yn)

**Arguments**

alpha0	A scalar defining the parameter for the Dirichlet process prior that controls the number of clusters (or its initial values)
w	A scalar representing the minimum number of points in each interval between two change points
N	A scalar representing the number of data sequences
M	A scalar representing the number of points available for each data sequence
bs	The hyperparameter value for the scale parameter in the inverse-gamma prior for the variance component
as	The hyperparameter value for the shape parameter in the inverse-gamma prior for the variance component
kstar	A scalar with the number maximum of change points in all clusters
lambda	A scalar defining the parameter for the Truncate Poisson distribution that controls the number of change points (or its initial values)
Yn	A vector or matrix with data sequences for a cluster

**Value**

A numerical value representing the mixing value term used to compute the probability that the given data sequence should be a singleton cluster

**Note**

This function is called within [gibbs\_alg()]. It should not be called alone.

**See Also**

[gibbs\_alg()]

**Examples**

```
qn0(alpha0 = 1/100, w = 10, N = 5, M = 50, bs = 1000, as = 2, kstar = 2, lambda = 2, Yn = data[,1])
```

---

qn0\_mk

*Mixing probability for creating new cluster per bin*


---

**Description**

Mixing probability for creating new cluster per bin

**Usage**

```
qn0_mk(w, m0, bs, as, M, km, lambda, mk, Yn, kstar)
```

**Arguments**

w	A scalar representing the minimum number of points in each interval between two change points
m0	A scalar for the number of positions available to define change-points positions
bs	The hyperparameter value for the scale parameter in the inverse-gamma prior for the variance component
as	The hyperparameter value for the shape parameter in the inverse-gamma prior for the variance component
M	A scalar representing the number of points available for each data sequence
km	A scalar for the number of changes points in a cluster
lambda	A scalar defining the parameter for the Truncate Poisson distribution that controls the number of change points (or its initial values)
mk	A matrix with all possible values to distribute between change points
Yn	A vector with a data sequence
kstar	A scalar with the number maximum of change points in all clusters

**Value**

A numerical value representing the mixing value term used to compute the probability that the given data sequence should be a singleton cluster for a given bin size.

**Note**

This function is called within [qn0()]. It should not be called alone.

**See Also**

[qn0()], [gibbs\_alg()]

**Examples**

```

data(data)
M <- 50; k <- 0; w <- 10;
m0 <- M - 1 -(k+1)*w
for(k in 0:2){
mk <- RcppAlgos::permuteGeneral(0:m0, k + 1,
constraintFun = "sum",
comparisonFun = "==", limitConstraints = m0,
repetition = TRUE)}
out <- qn0_mk(w = 10, m0 = m0, bs = 1000, as = 2, M = 50, km = 1,
lambda = 2, mk = mk[1,], Yn = data[,1], kstar = 2)

```

qnj

*Mixing probability for getting assigned to an existing cluster***Description**

Mixing probability for getting assigned to an existing cluster

**Usage**

```
qnj(N, M, as, bs, Yn, alpha, cluster, T1, K)
```

**Arguments**

N	A scalar representing the number of data sequences
M	A scalar representing the number of points available for each data sequence
as	The hyperparameter value for the shape parameter in the inverse-gamma prior for the variance component
bs	The hyperparameter value for the scale parameter in the inverse-gamma prior for the variance component
Yn	A vector or matrix with data sequences for a cluster
alpha	A list containing a vector for each cluster determining the constant level values for each interval between change points in each cluster (or its initial values)
cluster	A vector containing the cluster assignments for the data sequences (or its initial values)
T1	A list containing a vector for each cluster determining the change-point positions in each cluster (or its initial values)
K	A vector containing the number of change points for each cluster (or its initial values)

**Value**

A vector of same size as the vector 'cluster' corresponding to the mixing term value used to compute the probability that the given data sequence 'Yn' should be part of each existing cluster



**Note**

This function is called within the Gibbs sampler. It should not be called alone.

**See Also**

[gibbs\_alg()]

**Examples**

```
qnj(N = 5, M = 50, as = 2, bs = 1000, Yn = data[,1], alpha = c(10, 10),
    cluster = c(1,1,2,1,2), Tl = c(50,50), K = c(0,0))
```

---

run\_gibbs

*Runs the Gibbs sampler algorithm using using initial values for the parameters*

---

**Description**

Runs the Gibbs sampler algorithm using using initial values for the parameters

**Usage**

```
run_gibbs(  
  M,  
  N,  
  w,  
  d,  
  as = 2,  
  bs = 100,  
  a1 = 2,  
  b1 = 1000,  
  a = 2,  
  b = 1000,  
  alpha0 = 1/100,  
  lambda = 2,  
  maxIter = 10000,  
  par.values,  
  data,  
  cluster,  
  sigma2  
)
```

**Arguments**

M	A scalar representing the number of points available for each observation
N	A scalar representing the number of observations
w	A scalar representing the minimum number of points in each interval between two change points
d	A scalar representing the number of clusters.
as	The hyperparameter value for the shape parameter in the inverse-gamma prior for the variance component
bs	The hyperparameter value for the scale parameter in the inverse-gamma prior for the variance component
a1	The hyperparameter value for the shape parameter in the gamma prior for lambda
b1	The hyperparameter value for the scale parameter in the gamma prior for lambda
a	The hyperparameter value for the shape parameter in the gamma prior for alpha0
b	The hyperparameter value for the scale parameter in the gamma prior for alpha0
alpha0	A scalar defining the parameter for the Dirichlet process prior that controls the number of clusters (or its initial values)
lambda	A scalar defining the parameter for the Truncate Poisson distribution that controls the number of change points (or its initial values)
maxIter	A scalar for the number of iteration to run in the Gibbs sampler
par.values	A list with lists with parameters for each cluster. The first argument in each list is the number of change points, then the positions for the change points, where $T_1 = 1$ , $T_{last} = M + 1$ , and for each interval between change points you need to specify a value for the constant level. If running the Gibbs sampler for a dataset with unknown number of change points, we suggest setting the number of change points for each cluster to be zero. Check example in README file.
data	a matrix of size M x N with data sequences in the columns
cluster	a vector with cluster assignments for each data sequence
sigma2	a vector with variance components for each data sequence

**Value**

A list with estimates for each iteration of the Gibbs sampler for each parameter

**Examples**

```
d = 2 # two clusters
N = 5 # 5 data sequences
M = 50 # 50 observations for each data sequence
maxIter = 10 # number of Gibbs sampler iterations

data(data)
# initial values for each paramter and each cluster
par.values <- list(K = c(0, 0), T1 = list(50, 50), alpha = list(5, 10))
#cluster assignment for each data sequence
```

```

cluster <- kmeans(t(data), 2)$cluster
# variance for each data sequence
sigma2 <- apply(data, 2, var)
res <- run_gibbs(M, N, w = 10, d, as = 2, bs = 100, al = 2, bl = 1000, a = 2,
  b = 1000, alpha0 = 1/100, lambda = 2, maxIter = 10, par.values, data,
  cluster, sigma2)

```

---

update\_lambda

*Update equation for lambda*


---

### Description

Update equation for lambda

### Usage

```
update_lambda(a = 4, b = 2, kstar, lambda, cluster, al, bl, K, N)
```

### Arguments

a	The hyperparameter value for the shape parameter in the gamma prior for alpha0
b	The hyperparameter value for the scale parameter in the gamma prior for alpha0
kstar	A scalar with the number maximum of change points in all clusters
lambda	A scalar defining the parameter for the Truncate Poisson distribution that controls the number of change points (or its initial values)
cluster	A vector containing the cluster assignments for the data sequences (or its initial values)
al	The hyperparameter value for the shape parameter in the gamma prior for lambda
bl	The hyperparameter value for the scale parameter in the gamma prior for lambda
K	A vector containing the number of change points for each cluster (or its initial values)
N	A scalar representing the number of data sequences

### Value

A numerical value corresponding to a sample from the posterior of the parameter lambda

### Note

This function is called within the Gibbs sampler, but it can also be called separately.

### See Also

[gibbs\_alg()]

**Examples**

```
update_lambda(a = 4, b = 2, kstar = 2, lambda = 2, cluster = c(1,1,2,1,2),  
al = 2, bl = 1000, K = c(2,2), N = 5)
```

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