

# Package: MGMM (via r-universe)

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**Title** Missingness Aware Gaussian Mixture Models

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**Description** Parameter estimation and classification for Gaussian Mixture Models (GMMs) in the presence of missing data. This package complements existing implementations by allowing for both missing elements in the input vectors and full (as opposed to strictly diagonal) covariance matrices. Estimation is performed using an expectation conditional maximization algorithm that accounts for missingness of both the cluster assignments and the vector components. The output includes the marginal cluster membership probabilities; the mean and covariance of each cluster; the posterior probabilities of cluster membership; and a completed version of the input data, with missing values imputed to their posterior expectations. For additional details, please see McCaw ZR, Julienne H, Aschard H. "Fitting Gaussian mixture models on incomplete data." <doi:10.1186/s12859-022-04740-9>.

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CalHar

*Calinski-Harabaz Index*

---

### Description

Calculates the Calinski-Harabaz index.

### Usage

CalHar(data, assign, means)

**Arguments**

data	Observations.
assign	Assignments.
means	List of cluster means.

**Value**

Scalar metric.

---

ChooseK

*Cluster Number Selection*

---

**Description**

Function to choose the number of clusters  $k$ . Examines cluster numbers between  $k_0$  and  $k_1$ . For each cluster number, generates boot bootstrap data sets, fits the Gaussian Mixture Model ([FitGMM](#)), and calculates quality metrics ([ClustQual](#)). For each metric, determines the optimal cluster number  $k_{opt}$ , and the  $k_{1SE}$ , the smallest cluster number whose quality is within 1 SE of the optimum.

**Usage**

```
ChooseK(
  data,
  k0 = 2,
  k1 = NULL,
  boot = 100,
  init_means = NULL,
  fix_means = FALSE,
  init_covs = NULL,
  init_props = NULL,
  maxit = 10,
  eps = 1e-04,
  report = TRUE
)
```

**Arguments**

data	Numeric data matrix.
k0	Minimum number of clusters.
k1	Maximum number of clusters.
boot	Bootstrap replicates.
init_means	Optional list of initial mean vectors.
fix_means	Fix the means to their starting value? Must provide initial values.
init_covs	Optional list of initial covariance matrices.

<code>init_props</code>	Optional vector of initial cluster proportions.
<code>maxit</code>	Maximum number of EM iterations.
<code>eps</code>	Minimum acceptable increment in the EM objective.
<code>report</code>	Report bootstrap progress?

**Value**

List containing `Choices`, the recommended number of clusters according to each quality metric, and `Results`, the mean and standard error of the quality metrics at each cluster number evaluated.

**See Also**

See [ClustQual](#) for evaluating cluster quality, and [FitGMM](#) for estimating the GMM with a specified cluster number.

**Examples**

```
set.seed(100)
mean_list <- list(c(2, 2), c(2, -2), c(-2, 2), c(-2, -2))
data <- rGMM(n = 500, d = 2, k = 4, means = mean_list)
choose_k <- ChooseK(data, k0 = 2, k1 = 6, boot = 10)
choose_k$Choices
```

---

ClustQual

*Cluster Quality*


---

**Description**

Evaluates cluster quality. Returns the following metrics:

- BIC: Bayesian Information Criterion, lower value indicates better clustering quality.
- CHI: Calinski-Harabaz Index, higher value indicates better clustering quality.
- DBI: Davies-Bouldin, lower value indicates better clustering quality.
- SIL: Silhouette Width, higher value indicates better clustering quality.

**Usage**

```
ClustQual(fit)
```

**Arguments**

`fit` Object of class `mix`.

**Value**

List containing the cluster quality metrics.

**See Also**

See [ChooseK](#) for using quality metrics to choose the cluster number.

**Examples**

```
set.seed(100)

# Data generation
mean_list = list(
  c(2, 2, 2),
  c(-2, 2, 2),
  c(2, -2, 2),
  c(2, 2, -2)
)

data <- rGMM(n = 500, d = 3, k = 4, means = mean_list)
fit <- FitGMM(data, k = 4)

# Clustering quality
cluster_qual <- ClustQual(fit)
```

---

CombineMIs

*Combine Multiple Imputations*

---

**Description**

Combines point estimates and standard errors across multiple imputations.

**Usage**

```
CombineMIs(points, covs)
```

**Arguments**

points	List of point estimates, potentially vector valued.
covs	List of sampling covariances, potentially matrix valued.

**Value**

List containing the final point estimate ('point') and sampling covariance ('cov').

**Examples**

```
set.seed(100)

# Generate data and introduce missingness.
data <- rGMM(n = 25, d = 2, k = 1)
data[1, 1] <- NA
data[2, 2] <- NA
```

```
data[3, ] <- NA

# Fit GMM.
fit <- FitGMM(data)

# Lists to store summary statistics.
points <- list()
covs <- list()

# Perform 50 multiple imputations.
# For each, calculate the marginal mean and its sampling variance.
for (i in seq_len(50)) {
  imputed <- GenImputation(fit)
  points[[i]] <- apply(imputed, 2, mean)
  covs[[i]] <- cov(imputed) / nrow(imputed)
}

# Combine summary statistics across imputations.
results <- CombineMIs(points, covs)
```

---

DavBou

*Davies-Bouldin Index*

---

### **Description**

Calculates the Davies-Bouldin index.

### **Usage**

```
DavBou(data, assign, means)
```

### **Arguments**

data	Observations
assign	Assignments
means	List of cluster means

### **Value**

Scalar index.

FitGMM

*Estimate Multivariate Normal Mixture***Description**

Given an  $n \times d$  matrix of random vectors, estimates the parameters of a Gaussian Mixture Model (GMM). Accommodates arbitrary patterns of missingness at random (MAR) in the input vectors.

**Usage**

```
FitGMM(
  data,
  k = 1,
  init_means = NULL,
  fix_means = FALSE,
  init_covs = NULL,
  init_props = NULL,
  maxit = 100,
  eps = 1e-06,
  report = TRUE
)
```

**Arguments**

<code>data</code>	Numeric data matrix.
<code>k</code>	Number of mixture components. Defaults to 1.
<code>init_means</code>	Optional list of initial mean vectors.
<code>fix_means</code>	Fix the means to their starting value? Must provide initial values.
<code>init_covs</code>	Optional list of initial covariance matrices.
<code>init_props</code>	Optional vector of initial cluster proportions.
<code>maxit</code>	Maximum number of EM iterations.
<code>eps</code>	Minimum acceptable increment in the EM objective.
<code>report</code>	Report fitting progress?

**Details**

Initial values for the cluster means, covariances, and proportions are specified using  $M_0$ ,  $S_0$ , and  $\pi_0$ , respectively. If the data contains complete observations, i.e. observations with no missing elements, then `fit.GMM` will attempt to initialize these parameters internally using K-means. If the data contains no complete observations, then initial values are required for  $M_0$ ,  $S_0$ , and  $\pi_0$ .

**Value**

- For a single component, an object of class `mvn`, containing the estimated mean and covariance, the final objective function, and the imputed data.
- For a multicomponent model  $k > 1$ , an object of class `mix`, containing the estimated means, covariances, cluster proportions, cluster responsibilities, and observation assignments.

**See Also**

See [rGMM](#) for data generation, and [ChooseK](#) for selecting the number of clusters.

**Examples**

```
# Single component without missingness
# Bivariate normal observations
sigma <- matrix(c(1, 0.5, 0.5, 1), nrow = 2)
data <- rGMM(n = 1e3, d = 2, k = 1, means = c(2, 2), covs = sigma)
fit <- FitGMM(data, k = 1)

# Single component with missingness
# Trivariate normal observations
mean_list <- list(c(-2, -2, -2), c(2, 2, 2))
sigma <- matrix(c(1, 0.5, 0.5, 0.5, 1, 0.5, 0.5, 0.5, 1), nrow = 3)
data <- rGMM(n = 1e3, d = 3, k = 2, means = mean_list, covs = sigma)
fit <- FitGMM(data, k = 2)

# Two components without missingness
# Trivariate normal observations
mean_list <- list(c(-2, -2, -2), c(2, 2, 2))
sigma <- matrix(c(1, 0.5, 0.5, 0.5, 1, 0.5, 0.5, 0.5, 1), nrow = 3)
data <- rGMM(n = 1e3, d = 3, k = 2, means = mean_list, covs = sigma)
fit <- FitGMM(data, k = 2)

# Four components with missingness
# Bivariate normal observations
# Note: Fitting is slow.
mean_list <- list(c(2, 2), c(2, -2), c(-2, 2), c(-2, -2))
sigma <- 0.5 * diag(2)
data <- rGMM(
  n = 1000,
  d = 2,
  k = 4,
  pi = c(0.35, 0.15, 0.15, 0.35),
  m = 0.1,
  means = mean_list,
  covs = sigma)
fit <- FitGMM(data, k = 4)
```

**Description**

Given a matrix of random vectors, estimates the parameters for a mixture of multivariate normal distributions. Accommodates arbitrary patterns of missingness, provided the elements are missing at random (MAR).



**Usage**

```
FitMix(  
  data,  
  k = 2,  
  init_means = NULL,  
  fix_means = FALSE,  
  init_covs = NULL,  
  init_props = NULL,  
  maxit = 100,  
  eps = 1e-06,  
  report = FALSE  
)
```

**Arguments**

data	Numeric data matrix.
k	Number of mixture components. Defaults to 2.
init_means	Optional list of initial mean vectors.
fix_means	Fix means to their starting values? Must initialize.
init_covs	Optional list of initial covariance matrices.
init_props	Optional vector of initial cluster proportions.
maxit	Maximum number of EM iterations.
eps	Minimum acceptable increment in the EM objective.
report	Report fitting progress?

**Value**

Object of class `mix`.

---

FitMVN

*Fit Multivariate Normal Distribution*

---

**Description**

Given a matrix of  $n \times d$ -dimensional random vectors, possibly containing missing elements, estimates the mean and covariance of the best fitting multivariate normal distribution.

**Usage**

```
FitMVN(  
  data,  
  init_mean = NULL,  
  fix_mean = FALSE,  
  init_cov = NULL,
```

```
    maxit = 100,  
    eps = 1e-06,  
    report = TRUE  
  )
```

### Arguments

<code>data</code>	Numeric data matrix.
<code>init_mean</code>	Optional initial mean vector.
<code>fix_mean</code>	Fix the mean to its starting value? Must initialize.
<code>init_cov</code>	Optional initial covariance matrix.
<code>maxit</code>	Maximum number of EM iterations.
<code>eps</code>	Minimum acceptable increment in the EM objective.
<code>report</code>	Report fitting progress?

### Value

An object of class `mvn`.

---

<code>GenImputation</code>	<i>Generate Imputation</i>
----------------------------	----------------------------

---

### Description

Generates a stochastic imputation of a data set from a fitted data set.

### Usage

```
GenImputation(fit)
```

### Arguments

<code>fit</code>	Fitted model.
------------------	---------------

### Value

Numeric matrix with missing values imputed.

**Examples**

```

set.seed(100)

# Generate data and introduce missingness.
data <- rGMM(n = 25, d = 2, k = 1)
data[1, 1] <- NA
data[2, 2] <- NA
data[3, ] <- NA

# Fit GMM.
fit <- FitGMM(data)

# Generate imputation.
imputed <- GenImputation(fit)

```

---

logLik.mix	<i>Log likelihood for Fitted GMM</i>
------------	--------------------------------------

---

**Description**

Log likelihood for Fitted GMM

**Usage**

```

## S3 method for class 'mix'
logLik(object, ...)

```

**Arguments**

object	A mix object.
...	Unused.

---

logLik.mvn	<i>Log likelihood for Fitted MVN Model</i>
------------	--

---

**Description**

Log likelihood for Fitted MVN Model

**Usage**

```

## S3 method for class 'mvn'
logLik(object, ...)

```

**Arguments**

object	A mvn object.
...	Unused.

---

mean.mix	<i>Mean for Fitted GMM</i>
----------	----------------------------

---

**Description**

Mean for Fitted GMM

**Usage**

```
## S3 method for class 'mix'  
mean(x, ...)
```

**Arguments**

x	A mix object.
...	Unused.

---

mean.mvn	<i>Mean for Fitted MVN Model</i>
----------	----------------------------------

---

**Description**

Mean for Fitted MVN Model

**Usage**

```
## S3 method for class 'mvn'  
mean(x, ...)
```

**Arguments**

x	A mvn object.
...	Unused.

---

mix-class	<i>Mixture Model Class</i>
-----------	----------------------------

---

**Description**

Defines a class to hold Gaussian Mixture Models.

**Slots**

Assignments Maximum a posteriori assignments.

Completed Completed data, with missing values imputed to their posterior expectations.

Components Number of components.

Covariances List of fitted cluster covariance matrices.

Data Original data, with missing values present.

Density Density of each component at each example.

Means List of fitted cluster means.

Objective Final value of the EM objective.

Proportions Fitted cluster proportions.

Responsibilities Posterior membership probabilities for each example.

---

MixUpdateMeans	<i>Mean Update for Mixture of MVNs with Missingness.</i>
----------------	--

---

**Description**

Mean Update for Mixture of MVNs with Missingness.

**Usage**

```
MixUpdateMeans(split_data, means, covs, gamma)
```

**Arguments**

split\_data Data partitioned by missingness.

means List of component means.

covs List of component covariances.

gamma List of component responsibilities.

**Value**

List containing the updated component means.

---

mvn-class	<i>Multivariate Normal Model Class</i>
-----------	--

---

**Description**

Defines a class to hold multivariate normal models.

**Slots**

Completed Completed data, with missing values imputed to their posterior expectations.

Covariance Fitted covariance matrix.

Data Original data, with missing values present.

Mean Fitted mean vector.

Objective Final value of the EM objective.

---

PartitionData	<i>Partition Data by Missingness Pattern</i>
---------------	--

---

**Description**

Returns a list with the input data split in separate matrices for complete cases, incomplete cases, and empty cases.

**Usage**

```
PartitionData(data)
```

**Arguments**

data            Data.frame.

**Value**

List containing:

- The original row and column names: 'orig\_row\_names', 'orig\_col\_names'.
- The original row and column numbers: 'n\_row' and 'n\_col'.
- The complete cases 'data\_comp'.
- The incomplete cases 'data\_incomp'.
- The empty cases 'data\_empty'.
- Counts of complete 'n0', incomplete 'n1', and empty 'n2' cases.
- Initial order of the observations 'init\_order'.

---

print.mix	<i>Print for Fitted GMM</i>
-----------	-----------------------------

---

**Description**

Print method for objects of class mix.

**Usage**

```
## S3 method for class 'mix'  
print(x, ...)
```

**Arguments**

x	A mix object.
...	Unused.

---

print.mvn	<i>Print for Fitted MVN Model</i>
-----------	-----------------------------------

---

**Description**

Print for Fitted MVN Model

**Usage**

```
## S3 method for class 'mvn'  
print(x, ...)
```

**Arguments**

x	A mvn object.
...	Unused.

---

ReconstituteData	<i>Reconstitute Data</i>
------------------	--------------------------

---

**Description**

Reassembles a data matrix split by missingness pattern.

**Usage**

```
ReconstituteData(split_data)
```

**Arguments**

split\_data      Split data are returned by [PartitionData](#).

**Value**

Numeric matrix.

---

rGMM	<i>Generate Data from Gaussian Mixture Models</i>
------	---

---

**Description**

Generates an  $n \times d$  matrix of multivariate normal random vectors with observations (examples) as rows. If  $k = 1$ , all observations belong to the same cluster. If  $k > 1$  the observations are generated via two-step procedure. First, the cluster membership is drawn from a multinomial distribution, with mixture proportions specified by `pi`. Conditional on cluster membership, the observation is drawn from a multivariate normal distribution, with cluster-specific mean and covariance. The cluster means are provided using `means`, and the cluster covariance matrices are provided using `covs`. If `miss > 0`, missingness is introduced, completely at random, by setting that proportion of elements in the data matrix to NA.

**Usage**

```
rGMM(n, d = 2, k = 1, pi = NULL, miss = 0, means = NULL, covs = NULL)
```

**Arguments**

n	Observations (rows).
d	Observation dimension (columns).
k	Number of mixture components. Defaults to 1.
pi	Mixture proportions. If omitted, components are assumed equiprobable.
miss	Proportion of elements missing, $miss \in [0, 1)$ .



means	Either a prototype mean vector, or a list of mean vectors. Defaults to the zero vector.
covs	Either a prototype covariance matrix, or a list of covariance matrices. Defaults to the identity matrix.

### Value

Numeric matrix with observations as rows. Row numbers specify the true cluster assignments.

### See Also

For estimation, see [FitGMM](#).

### Examples

```
set.seed(100)
# Single component without missingness.
# Bivariate normal observations.
cov <- matrix(c(1, 0.5, 0.5, 1), nrow = 2)
data <- rGMM(n = 1e3, d = 2, k = 1, means = c(2, 2), covs = cov)

# Single component with missingness.
# Trivariate normal observations.
mean_list <- list(c(-2, -2, -2), c(2, 2, 2))
cov <- matrix(c(1, 0.5, 0.5, 0.5, 1, 0.5, 0.5, 0.5, 1), nrow = 3)
data <- rGMM(n = 1e3, d = 3, k = 2, means = mean_list, covs = cov)

# Two components without missingness.
# Trivariate normal observations.
mean_list <- list(c(-2, -2, -2), c(2, 2, 2))
cov <- matrix(c(1, 0.5, 0.5, 0.5, 1, 0.5, 0.5, 0.5, 1), nrow = 3)
data <- rGMM(n = 1e3, d = 3, k = 2, means = mean_list, covs = cov)

# Four components with missingness.
# Bivariate normal observations.
mean_list <- list(c(2, 2), c(2, -2), c(-2, 2), c(-2, -2))
cov <- 0.5 * diag(2)
data <- rGMM(
  n = 1000,
  d = 2,
  k = 4,
  pi = c(0.35, 0.15, 0.15, 0.35),
  miss = 0.1,
  means = mean_list,
  covs = cov)
```

---

show,mix-method	<i>Show for Fitted Mixture Models</i>
-----------------	---------------------------------------

---

**Description**

Show for Fitted Mixture Models

**Usage**

```
## S4 method for signature 'mix'  
show(object)
```

**Arguments**

object            A mix object.

---

show,mvn-method	<i>Show for Multivariate Normal Models</i>
-----------------	--

---

**Description**

Show for Multivariate Normal Models

**Usage**

```
## S4 method for signature 'mvn'  
show(object)
```

**Arguments**

object            A mvn object.

---

vcov.mix	<i>Covariance for Fitted GMM</i>
----------	----------------------------------

---

**Description**

Covariance for Fitted GMM

**Usage**

```
## S3 method for class 'mix'  
vcov(object, ...)
```

**Arguments**

object	A mix object.
...	Unused.

---

vcov.mvn	<i>Covariance for Fitted MVN Model</i>
----------	--

---

**Description**

Covariance for Fitted MVN Model

**Usage**

```
## S3 method for class 'mvn'  
vcov(object, ...)
```

**Arguments**

object	A mvn object.
...	Unused.

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