

# Package ‘ExtMallows’

January 20, 2025

**Type** Package

**Title** An Extended Mallows Model and Its Hierarchical Version for  
Ranked Data Aggregation

**Version** 0.1.0

**Date** 2018-06-28

**Description** For multiple full/partial ranking lists, R package 'ExtMallows' can (1) de-  
tect whether the input ranking lists are over-correlated, and (2) use the Mallows model or ex-  
tended Mallows model to integrate the ranking lists, and (3) use hierarchical extended Mal-  
lows model for rank integration if there are groups of over-correlated ranking lists.

**Author** Han Li, Minxuan Xu, Jun S. Liu and Xiaodan Fan

**Maintainer** Han Li <hli@szu.edu.cn>

**Depends** R (>= 3.1.0)

**License** GPL (>= 2)

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2018-07-05 15:30:10 UTC

## Contents

corrRankings . . . . .	2
EMM . . . . .	3
HEMM . . . . .	4
MM . . . . .	5
NBArankings . . . . .	6
simu1 . . . . .	8
simu2 . . . . .	9
simu3 . . . . .	9

## Index

11

**corrRankings***p value for measuring the correlation of pairwise rankings***Description**

It calculates the p values that measure the correlation of pairwise rankings.

**Usage**

```
corrRankings(rankings)
```

**Arguments**

rankings	A n by m data frame, with each column representing a ranking list, which ranks the items from the most preferred to the least preferred. For missing items, use 0 to denote them.
----------	---

**Value**

pair.pvalue	a symmetric matrix of p values, with the (i,j)-th element denoting the p value of the i,j-th rankings.
-------------	--

**Note**

Note that the input rankings should have at least 8 rankings. When constructing the samples of rescaled V distance for a given rank position, the number of samples should at least be 28 and the number of rankings that have items up to this position should account for at least 2/3 of the total number of rankings, otherwise the p value calculation stops at this position.

**Author(s)**

Han Li, Minxuan Xu, Jun S. Liu and Xiaodan Fan

**References**

An extended Mallows model for ranked data aggregation

**Examples**

```
data(simu3)
pvalue=corrRankings(rankings = simu3)

#threshold the p values

threshold=0.05
pvalue.trunc=ifelse(pvalue<=0.05, pvalue, 1)

#plot the p values
```

```

x=y=1:ncol(pvalue)
par(mfrow=c(1,2))
image(x, y, pvalue, xlab = NA, ylab = NA, sub = "rank coefficient")
image(x, y, pvalue.trunc, xlab = NA, ylab = NA, sub = "rank coefficient < 0.05")

```

## Description

It uses the extended Mallows model to aggregate multiple full/partial ranking lists.

## Usage

```
EMM(rankings, initial.method, it.max)
```

## Arguments

<code>rankings</code>	A n by m matrix, with each column representing a ranking list, which ranks the items from the most preferred to the least preferred. For missing items, use 0 to denote them.
<code>initial.method</code>	the method for initializing the value of pi0, with four options: mean, median, geometric and random (the mean of three randomly sampled ranking lists). By default, <code>initial.method="mean"</code> .
<code>it.max</code>	the maximum number of iterations. By default, <code>it.max=20</code> .

## Value

<code>op.phi</code>	optimal value of phi
<code>op.omega</code>	optimal value of omega
<code>op.alpha</code>	optimal value of alpha
<code>op.pi0</code>	optimal value of pi0, ranking the items from the most preferred to the least preferred
<code>max.logL</code>	maximum value of log-likelihood

## Author(s)

Han Li, Minxuan Xu, Jun S. Liu and Xiaodan Fan

## References

An extended Mallows model for ranked data aggregation

## Examples

```
data(simu1)
res=EMM(rankings = simu1, initial.method = "mean", it.max = 20)
res$op.phi
res$op.omega
res$op.pi0
```

HEMM

*A hierarchical extended Mallows model for aggregating multiple ranking lists*

## Description

It uses the hierarchical extended Mallows model to aggregate multiple full/partial ranking lists.

## Usage

```
HEMM(rankings, num.kappa, is.kappa.ranker, initial.method, it.max)
```

## Arguments

rankings	A n by m matrix, with each column representing a ranking list, which ranks the items from the most preferred to the least preferred. For missing items, use 0 to denote them.
num.kappa	the number of over-correlated ranking groups
is.kappa.ranker	a list of over-correlated ranking groups, with the k-th element denoting the column numbers of the rankings that belong to the k-th group
initial.method	the method for initializing the value of pi0, with four options: mean, median, geometric and random (the mean of three randomly sampled ranking lists). By default, initial.method="mean".
it.max	the maximum number of iterations. By default, it.max=20.

## Value

op.phi	optimal value of phi
op.phi1	optimal value of phi1, the phi value in over-correlated ranking groups
op.omega	optimal value of omega
op.alpha	optimal value of alpha
op.pi0	optimal value of pi0, ranking the items from the most preferred to the least preferred
op.kappa	optimal value of kappa, denoting the items from the most preferred to the least preferred
max.logL	maximum value of log-likelihood

**Author(s)**

Han Li, Minxuan Xu, Jun S. Liu and Xiaodan Fan

**References**

An extended Mallows model for ranked data aggregation

**Examples**

```
data(simu3)
res=HEMM(rankings = simu3, num.kappa = 2, is.kappa.ranker = list(1:5, 6:10),
          initial.method = "mean", it.max = 20)
res$op.phi
res$op.phi1
res$op.omega
res$op.pi0

data(NBArankings)
res=HEMM(rankings = NBArankings, num.kappa = 1, is.kappa.ranker = list(1:6),
          initial.method = "mean", it.max = 20)
res$op.omega
res$op.pi0
res$op.kappa
```

**Description**

It uses the Mallows model to aggregate multiple full/partial ranking lists.

**Usage**

```
MM(rankings, initial.method, it.max)
```

**Arguments**

- |                             |  |
|-----------------------------|--|
| <code>rankings</code>       | A n by m matrix, with each column representing a ranking list, which ranks the items from the most preferred to the least preferred. For missing items, use 0 to denote them.                            |
| <code>initial.method</code> | the method for initializing the value of pi0, with four options: mean, median, geometric and random (the mean of three randomly sampled ranking lists). By default, <code>initial.method="mean"</code> . |
| <code>it.max</code>         | the maximum number of iterations. By default, <code>it.max=20</code> .   |

**Value**

op.phi	optimal value of phi
op.pi0	optimal value of pi0, ranking the items from the most preferred to the least preferred
max.logL	maximum value of log-likelihood

**Author(s)**

Han Li, Minxuan Xu, Jun S. Liu and Xiaodan Fan

**References**

Mallows, C. L. (1957). Non-null ranking models, Biometrika 44(1/2): 114-130.

**Examples**

```
data(simu1)
res=MM(rankings = simu1, initial.method = "mean", it.max = 20)
res$op.phi
res$op.pi0
```

NBArankings

*A real example of rankings of NBA teams*

**Description**

This example is about aggregating the multiple rankings of NBA teams and was studied by Deng et al. (2014). They collected 34 rankings, including 6 professional rankings and 28 amateur rankings, for the 30 NBA teams in the 2011-2012 season. For the missing items in the partial rankings, we use number 0 to denote them.

**Usage**

```
data("NBArankings")
```

**Format**

A data frame with 30 observations on the following 34 variables.

- V1 a factor with levels 76ers Bobcats Bucks Bulls Cavaliers Celtics Clippers Grizzlies Hawks Heat Hornets Jazz Kings Knicks Lakers Magic Mavericks Nets Nuggets Pacers Pistons Raptors Rockets Spurs Suns Thunder Timberwolves TrailBlazers Warriors Wizards
- V2 a factor with levels 76ers Bobcats Bucks Bulls Cavaliers Celtics Clippers Grizzlies Hawks Heat Hornets Jazz Kings Knicks Lakers Magic Mavericks Nets Nuggets Pacers Pistons Raptors Rockets Spurs Suns Thunder Timberwolves TrailBlazers Warriors Wizards

- V3 a factor with levels 76ers Bobcats Bucks Bulls Cavaliers Celtics Clippers Grizzlies Hawks Heat Hornets Jazz Kings Knicks Lakers Magic Mavericks Nets Nuggets Pacers Pistons Raptors Rockets Spurs Suns Thunder Timberwolves TrailBlazers Warriors Wizards
- V4 a factor with levels 76ers Bobcats Bucks Bulls Cavaliers Celtics Clippers Grizzlies Hawks Heat Hornets Jazz Kings Knicks Lakers Magic Mavericks Nets Nuggets Pacers Pistons Raptors Rockets Spurs Suns Thunder Timberwolves TrailBlazers Warriors Wizards
- V5 a factor with levels 76ers Bobcats Bucks Bulls Cavaliers Celtics Clippers Grizzlies Hawks Heat Hornets Jazz Kings Knicks Lakers Magic Mavericks Nets Nuggets Pacers Pistons Raptors Rockets Spurs Suns Thunder Timberwolves TrailBlazers Warriors Wizards
- V6 a factor with levels 76ers Bobcats Bucks Bulls Cavaliers Celtics Clippers Grizzlies Hawks Heat Hornets Jazz Kings Knicks Lakers Magic Mavericks Nets Nuggets Pacers Pistons Raptors Rockets Spurs Suns Thunder Timberwolves TrailBlazers Warriors Wizards
- V7 a factor with levels 0 Bulls Celtics Hawks Heat Lakers Pacers Spurs Thunder
- V8 a factor with levels 0 Bulls Celtics Clippers Heat Knicks Lakers Spurs Thunder
- V9 a factor with levels 0 Bulls Celtics Heat Knicks Lakers Mavericks Spurs Thunder
- V10 a factor with levels 0 Bulls Celtics Clippers Heat Lakers Mavericks Spurs Thunder
- V11 a factor with levels 0 Bulls Celtics Heat Knicks Lakers Nuggets Warriors Wizards
- V12 a factor with levels 0 Bulls Celtics Clippers Heat Lakers Mavericks Spurs Thunder
- V13 a factor with levels 0 Bulls Celtics Hornets Jazz Kings Lakers Magic Rockets
- V14 a factor with levels 0 76ers Celtics Heat Kings Lakers Rockets Spurs Suns
- V15 a factor with levels 0 Bulls Celtics Heat Lakers Mavericks Rockets Spurs Thunder
- V16 a factor with levels 0 Celtics Hawks Heat Lakers Mavericks Raptors Spurs Thunder
- V17 a factor with levels 0 76ers Celtics Heat Knicks Lakers Mavericks Nets Thunder
- V18 a factor with levels 0 76ers Bulls Cavaliers Celtics Heat Lakers Mavericks Thunder
- V19 a factor with levels 0 Bulls Heat Kings Lakers Rockets Spurs Suns Warriors
- V20 a factor with levels 0 Bucks Celtics Heat Lakers Magic Mavericks Rockets Suns
- V21 a factor with levels 0 Celtics Heat Kings Lakers Mavericks Spurs Suns Timberwolves
- V22 a factor with levels 0 Celtics Heat Kings Lakers Spurs Suns Thunder Timberwolves
- V23 a factor with levels 0 Bobcats Celtics Heat Lakers Mavericks Nuggets Spurs Suns
- V24 a factor with levels 0 76ers Heat Knicks Lakers Pistons Rockets Spurs Wizards
- V25 a factor with levels 0 76ers Celtics Hawks Heat Knicks Lakers Magic Thunder
- V26 a factor with levels 0 Bulls Cavaliers Celtics Hawks Heat Knicks Lakers Rockets
- V27 a factor with levels 0 76ers Clippers Lakers Magic Mavericks Pacers Raptors Warriors
- V28 a factor with levels 0 76ers Bulls Celtics Heat Lakers Pistons Rockets Wizards
- V29 a factor with levels 0 76ers Bulls Grizzlies Hawks Kings Knicks Nets Timberwolves
- V30 a factor with levels 0 76ers Bucks Bulls Knicks Raptors Rockets Thunder Timberwolves

V31 a factor with levels 0 76ers Heat Lakers Magic Mavericks Pacers Pistons Suns  
 V32 a factor with levels 0 76ers Bulls Celtics Heat Knicks Lakers Magic Pacers  
 V33 a factor with levels 0 Clippers Heat Knicks Lakers Mavericks Nets Nuggets Wizards  
 V34 a factor with levels 0 Bulls Hawks Heat Jazz Knicks Nets Rockets Timberwolves

## References

Deng, K., Han, S., Li, K. J. and Liu, J. S. (2014). Bayesian aggregation of order-based rank data, Journal of the American Statistical Association 109(507): 1023-1039.

## Examples

```
data(NBArankings)
dim(NBArankings)
```

*simu1*

*Simulation data 1*

## Description

This data set is simulated as described in the Simulation Study 1 of the reference. It is a 30 by 6 data frame, representing 6 independent top-30 partial rankings.

## Usage

```
data("simu1")
```

## Format

A data frame with 30 observations on the following 6 variables.

V1 a numeric vector  
 V2 a numeric vector  
 V3 a numeric vector  
 V4 a numeric vector  
 V5 a numeric vector  
 V6 a numeric vector

## References

An extended Mallows model for ranked data aggregation

## Examples

```
data(simu1)
dim(simu1)
```

---

**simu2***Simulation data 2*

---

**Description**

This data set is simulated as described in the Simulation Study 2 of the reference. It is a 40 by 6 data frame, representing 6 independent top-40 partial rankings.

**Usage**

```
data("simu2")
```

**Format**

A data frame with 40 observations on the following 6 variables.

V1 a numeric vector  
V2 a numeric vector  
V3 a numeric vector  
V4 a numeric vector  
V5 a numeric vector  
V6 a numeric vector

**References**

An extended Mallows model for ranked data aggregation

**Examples**

```
data(simu2)  
dim(simu2)
```

---

**simu3***Simulation data 3*

---

**Description**

This data set is simulated as described in the Simulation Study 3 of the reference. It is a 100 by 20 data frame, representing 20 full rankings. The columns 1-5 and the columns 6-10 represent two highly correlated ranking groups, respectively.

**Usage**

```
data("simu3")
```

**Format**

A data frame with 100 observations on the following 20 variables.

V1 a numeric vector  
V2 a numeric vector  
V3 a numeric vector  
V4 a numeric vector  
V5 a numeric vector  
V6 a numeric vector  
V7 a numeric vector  
V8 a numeric vector  
V9 a numeric vector  
V10 a numeric vector  
V11 a numeric vector  
V12 a numeric vector  
V13 a numeric vector  
V14 a numeric vector  
V15 a numeric vector  
V16 a numeric vector  
V17 a numeric vector  
V18 a numeric vector  
V19 a numeric vector  
V20 a numeric vector

**References**

An extended Mallows model for ranked data aggregation

**Examples**

```
data(simu3)
dim(simu3)
```

# Index

## \* datasets

NBArankings, 6

simu1, 8

simu2, 9

simu3, 9

corrRankings, 2

EMM, 3

HEMM, 4

MM, 5

NBArankings, 6

simu1, 8

simu2, 9

simu3, 9