

# Package ‘Rfmtool’

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**Depends** R (>= 2.9.2)

**Suggests**

**Description** Various tools for handling fuzzy measures, calculating Shapley value and interaction index, Choquet and Sugeno integrals, as well as fitting fuzzy measures to empirical data are provided. Construction of fuzzy measures from empirical data is done by solving a linear programming problem by using 'lpsolve' package, whose source in C adapted to the R environment is included. The description of the basic theory of fuzzy measures is in the manual in the Doc folder in this package.

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**NeedsCompilation** yes

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fm	<i>Rfntool package</i>
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## Description

This function shows a list of function included in this toolbox

## Usage

fm()

## Details

The following functions involve the parameters  $v$  (the array containing the fuzzy measure in standard representation) or  $Mob$  (in Mobius representation),  $n$  - the dimension and  $m = 2^n$ . The values of the fuzzy measure always obey the binary ordering.

## Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

## References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272. [3] G. Beliakov, H. Bustince, T. Calvo, A Practical guide to Averaging, Springer, Berlin, New York, 2016. [4] G. Beliakov, Programming Library Fmtools for Handling and Fiffing Fuzzy Measures, User manual, provided with this package in the Doc folder.

## Examples

```
fm()
```

---

fm.Banzhaf	<i>Banzhaf value computation function</i>
------------	---

---

## Description

Calculates the Banzhaf indices of input criteria from general fuzzy measure.

## Usage

```
fm.Banzhaf(v, env=NULL)
```

## Arguments

v	is fuzzy measure in general representation.
env	Environment variable obtained from fm.Init(n).

## Value

output	The output is an array of size n, which contain Banzhaf indices of input criteria.
--------	--

## Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

## References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

## Examples

```
env<-fm.Init(3)
fm.Banzhaf(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1), env)
```

---

fm.BanzhafMob	<i>Banzhaf value computation function</i>
---------------	---

---

**Description**

Calculates the Banzhaf indices of input criteria from general fuzzy measure in Mobius representation.

**Usage**

```
fm.BanzhafMob(Mob, env=NULL)
```

**Arguments**

Mob	is fuzzy measure in Mobius representation.
env	Environment variable obtained from fm.Init(n).

**Value**

output	The output is an array of size n, which contain Banzhaf indices of input criteria.
--------	--

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.BanzhafMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.Bipartition	<i>Bipartition interaction index computation function</i>
----------------	---

---

**Description**

Calculates the Bipartition interaction indices of input criteria from general fuzzy measure.

**Usage**

```
fm.Bipartition(v,env=NULL)
```

**Arguments**

v	is fuzzy measure in general representation.
env	Environment variable obtained from fm.Init(n).

**Value**

output	The output is an array of size $2^n$ , which contain bipartition interaction indices of input criteria coalitions.
--------	--

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.Bipartition(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.BipartitionBanzhaf *Bipartition Banzhaf interaction index computation function*

---

### Description

Calculates the Banzhaf Bipartition interaction indices of input criteria from general fuzzy measure.

### Usage

```
fm.BipartitionBanzhaf(v,env=NULL)
```

### Arguments

v	is fuzzy measure in general representation.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is an array of size $2^n$ , which contain Banzhaf bipartition interaction indices of input criteria coalitions.
--------	--

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.BipartitionBanzhaf(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

 fm.Choquet

*Choquet function*


---

### Description

Calculates the value of a discrete Choquet integral of input  $x$ , with fuzzy measure in general representation

### Usage

```
fm.Choquet(x, v, env=NULL)
```

### Arguments

$x$	input vector of size $n$ , containing utility value of input criteria. $x$ is in $[0,1]$ .
$v$	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
$env$	Environment variable obtained from <code>fm.Init(n)</code> .

### Value

output	The ouput is a single value of the computed Choquet integral.
--------	---

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.Choquet(c(0.6, 0.3, 0.8), c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```



---

fm.ChoquetMob	<i>ChoquetMob function</i>
---------------	----------------------------

---

### Description

This is an alternative calculation of the Choquet integral from the fuzzy measure in Mobius representation.

### Usage

```
fm.ChoquetMob(x, Mob, env=NULL)
```

### Arguments

x	input vector of size n, containing utility value of input criteria. x is in [0,1].
Mob	the Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	the output is a single value of the computed Choquet integral.
--------	--

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.ChoquetMob(c(0.6, 0.3, 0.8), c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

fm.ConstructLambdaMeasure

*ConstructLambdaMeasure function*

---

### Description

Finds the value of lambda and calculates the rest of the values of the fuzzy measure, given its values at singletons; singletons is an array of size n. The outputs are lambda and v, v is in standard representation and binary ordering.

### Usage

```
fm.ConstructLambdaMeasure(singletons, env)
```

### Arguments

singletons	singletons is an array of n.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is the list (lambda, measure), where measure is a fuzzy measure in standard representation.
--------	--

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
w <- fm.ConstructLambdaMeasure(c(0, 0.3, 0.5), env)
```

---

`fm.ConstructLambdaMeasureMob`*ConstructLambdaMeasureMob function*

---

**Description**

Finds the value of lambda and calculates the rest of the values of the fuzzy measure, given its values at singletons; singletons is an array of size n. The outputs are lambda and measure, measure is in Mobius representation.

**Usage**

```
fm.ConstructLambdaMeasureMob(singletons, env)
```

**Arguments**

singletons	singletons is an array of n.
env	Environment variable obtained from fm.Init(n).

**Value**

output	The output is the list (lambda, measure), where measure is a fuzzy measure in Mobius representation.
--------	--

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
w <- fm.ConstructLambdaMeasureMob(c(0, 0.3, 0.5), env)
w$measure
```

---

fm.dualm	<i>dualm function</i>
----------	-----------------------

---

### Description

Calculates the dual of fuzzy measure  $v$ , returns it as value of the function (array of size  $m$ ).

### Usage

```
fm.dualm(v, env=NULL)
```

### Arguments

$v$	the general fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
$env$	Environment variable obtained from <code>fm.Init(n)</code> .

### Value

output	The output is an array of size $m$ with the dual of fuzzy measure $v$ .
--------	---

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
w <- fm.dualm(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.dualmMob	<i>dualm function</i>
-------------	-----------------------

---

### Description

Calculates the dual of fuzzy measure  $v$ , returns it as value of the function (array of size  $m$ ).

### Usage

```
fm.dualmMob(Mob, env=NULL)
```

### Arguments

Mob	the Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from <code>fm.Init(n)</code> .

### Value

output	The output is an array of size $m$ with the dual of fuzzy measure $v$ .
--------	---

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
w <- fm.dualm(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.EntropyChoquet      *EntropyChoquet function*

---

**Description**

Calculates entropy value of the Choquet integral for the fuzzy measure  $\nu$  in general representation

**Usage**

```
fm.EntropyChoquet( $\nu$ , env)
```

**Arguments**

$\nu$                       the general fuzzy measure of size  $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.

env                      Environment variable obtained from fm.Init(n).

**Value**

output                      The output is entropy value of the Choquet integral for the fuzzy measure.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A EntropyChoquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.EntropyChoquet(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1))
```

---

fm.EntropyChoquetMob *EntropyChoquet function*

---

### Description

Calculates entropy value of the Choquet integral for the fuzzy measure  $v$  in Mobius representation

### Usage

```
fm.EntropyChoquetMob(Mob, env)
```

### Arguments

Mob	the Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is entropy value of the Choquet integral for the fuzzy measure.
--------	--

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A EntropyChoquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.EntropyChoquetMob(c(0.0,0.3,0.5,-0.2,0.4,0.1,-0.2,0.1),env)
```

---

fm.errorcheck	<i>Basic error check</i>
---------------	--------------------------

---

**Description**

This function checks that the environment variable is internally consistent.

**Usage**

```
fm.errorcheck(env)
```

**Arguments**

env	Environment variable obtained from fm.Init(n).
-----	--

**Value**

output	The output is TRUE or FALSE.
--------	------------------------------

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.errorcheck(env)
```

---

fm.fitting	<i>Fuzzy Measure Fitting function</i>
------------	---------------------------------------

---

**Description**

Estimate values of the fuzzy measures from empirical data.

**Usage**

```
fm.fitting(data, env=NULL, kadd="NA")
```



**Arguments**

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0, 1]^n$ is a vector containing utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0, 1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ stores the observed aggregated value $y$ .
env	Environment variable obtained from <code>fm.Init(n)</code> .
kadd	is the value of $k$ -additivity, which is used for reducing the complexity of fuzzy measures. <code>kadd</code> is defined as an optional argument, its default value is <code>kadd = n</code> .

**Value**

output	The output is an array of size $2^n$ containing estimated standard fuzzy measure in binary ordering.
--------	--

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
            nrow=20,
```

```

        ncol=4);
fm.fitting(d,env)

```

---

```
fm.fittingKinteractive
```

*Fuzzy Measure Fitting function*

---

### Description

Estimate values of the k-interactive fuzzy measures from empirical data.

### Usage

```
fm.fittingKinteractive(data, env=NULL, kadd="NA", K="NA")
```

### Arguments

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0, 1]^n$ is a vector containing utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0, 1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ stores the observed aggregated value $y$ .
env	Environment variable obtained from <code>fm.Init(n)</code> .
kadd	is the value of k-interactivity, which is used for reducing the complexity of fuzzy measures. kadd is defined as an optional argument, its default value is <code>kadd = 2</code> .
K	is the value of FM value for sets of cardinality <code>kadd+1</code> , its default value is <code>K = 0.5</code> .

### Value

output	The output is an array of size $2^n$ containing estimated standard fuzzy measure in binary ordering.
--------	--

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```

env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
            nrow=20,
            ncol=4);
fm.fittingKinteractive(d,env,2,0.8)

```

---

fm.fittingKinteractiveAuto

*Fuzzy Measure Fitting function*

---

**Description**

Estimate values of the k-interactive fuzzy measures from empirical data.

**Usage**

```
fm.fittingKinteractiveAuto(data, env=NULL, kadd="NA")
```

**Arguments**

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0,1]^n$ is a vector containing utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0,1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ stores the observed aggregated value $y$ .
env	Environment variable obtained from fm.Init( $n$ ).

**kadd** is the value of k-interactivity, which is used for reducing the complexity of fuzzy measures. **kadd** is defined as an optional argument, its default value is **kadd = 2**. The constant **K** the value of FM value for sets of cardinality **kadd+1** is computed from data.

### Value

**output** The output is an array of size  $2^n$  containing estimated standard fuzzy measure in binary ordering.

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
             nrow=20,
             ncol=4);
fm.fittingKinteractiveAuto(d,env,2)
```

---

 fm.fittingKinteractiveMarginal

*Fuzzy Measure Fitting function*


---

### Description

Estimate values of the k-interactive fuzzy measures from empirical data using marginal representation.

### Usage

```
fm.fittingKinteractiveMarginal(data, env=NULL, kadd="NA", K="NA", submod ="NA")
```

### Arguments

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0, 1]^n$ is a vector containing utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0, 1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ stores the observed aggregated value $y$ .
env	Environment variable obtained from <code>fm.Init(n)</code> .
kadd	is the value of k-interactivity, which is used for reducing the complexity of fuzzy measures. <code>kadd</code> is defined as an optional argument, its default value is <code>kadd = 2</code> .
K	The constant $K$ the value of FM value for sets of cardinality <code>kadd+1</code> is computed from data, default 0.5.
submod	-1 indicates supermodular FM is needed, +1 indicates submodular, 0 otherwise. Should be consistent with $K$ and $n$ , see manual

### Value

output	The output is an array of size $2^n$ containing estimated standard fuzzy measure in binary ordering.
--------	--

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```

env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
             nrow=20,
             ncol=4);
fm.fittingKinteractiveMarginal(d,env,2,0.6, 0)

```

---

fm.fittingKinteractiveMarginalMC

*Fuzzy Measure Fitting function*

---

**Description**

Estimate values of the k-interactive fuzzy measures from empirical data using marginal representation and maximal chains method.

**Usage**

```
fm.fittingKinteractiveMarginalMC(data, env=NULL, kadd="NA", K="NA", submod ="NA")
```

**Arguments**

**data** is the empirical data set in pairs  $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$  where  $x_i$  in  $[0, 1]^n$  is a vector containing utility values of  $n$  input criteria  $x_{i1}, x_{i2}, \dots, x_{in}$ ,  $y_i$  in  $[0, 1]$  is a single aggregated value given by decision makers. The data is stored as a matrix of  $M$  by  $n+1$  elements, where  $M$  is the number of data instances, and  $n$  is the number of input criteria, the column  $n + 1$  stores the observed aggregated value  $y$ .

env	Environment variable obtained from fm.Init(n).
kadd	is the value of k-interactivity, which is used for reducing the complexity of fuzzy measures. kadd is defined as an optional argument, its default value is kadd = 2.
K	The constant K the value of FM value for sets of cardinality kadd+1 is computed from data, default 0.5.
submod	-1 indicates supermodular FM is needed, +1 indicates submodular, 0 otherwise. Should be consistent with K and n, see manual

**Value**

output	The output is an array of size $2^n$ containing estimated standard fuzzy measure in binary ordering.
--------	--

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
             nrow=20,
             ncol=4);
fm.fittingKinteractiveMarginalMC(d,env,2,0.6,0)
```

---

 fm.fittingKinteractiveMC

*Fuzzy Measure Fitting function*


---

### Description

Estimate values of the k-interactive fuzzy measures from empirical data using maximal chains method.

### Usage

```
fm.fittingKinteractiveMC(data, env=NULL, kadd="NA", K="NA")
```

### Arguments

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0, 1]^n$ is a vector containing utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0, 1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ stores the observed aggregated value $y$ .
env	Environment variable obtained from <code>fm.Init(n)</code> .
kadd	is the value of k-interactivity, which is used for reducing the complexity of fuzzy measures. kadd is defined as an optional argument, its default value is <code>kadd = 2</code> .
K	The constant $K$ the value of FM value for sets of cardinality <code>kadd+1</code> is computed from data, default 0.5.

### Value

output	The output is an array of size $2^n$ containing estimated standard fuzzy measure in binary ordering.
--------	--

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.



**Examples**

```

env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
             nrow=20,
             ncol=4);
fm.fittingKinteractiveMC(d,env,2,0.6)

```

---

fm.fittingKmaxitive     *Fuzzy Measure Fitting function*

---

**Description**

Estimate values of the k-maxitive fuzzy measures from empirical data.

**Usage**

```
fm.fittingKmaxitive(data, env=NULL, kadd="NA")
```

**Arguments**

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0, 1]^n$ is a vector containing utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0, 1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ stores the observed aggregated value $y$ .
env	Environment variable obtained from <code>fm.Init(n)</code> .
kadd	is the value of k-maxitivity, which is used for reducing the complexity of fuzzy measures. kadd is defined as an optional argument, its default value is <code>kadd = n</code> .

**Value**

output            The output is an array of size  $2^n$  containing estimated standard fuzzy measure in binary ordering.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
             nrow=20,
             ncol=4);
fm.fittingKmaxitive(d,env,2)
```

---

fm.fittingKtolerant    *Fuzzy Measure Fitting function*

---

**Description**

Estimate values of the k-tolerant fuzzy measures from empirical data.

**Usage**

```
fm.fittingKtolerant(data, env=NULL, kadd="NA")
```

**Arguments**

**data** is the empirical data set in pairs  $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$  where  $x_i$  in  $[0, 1]^n$  is a vector containing utility values of  $n$  input criteria  $x_{i1}, x_{i2}, \dots, x_{in}$ ,  $y_i$  in  $[0, 1]$  is a single aggregated value given by decision makers. The data is stored as a matrix of  $M$  by  $n+1$  elements, where  $M$  is the number of data instances, and  $n$  is the number of input criteria, the column  $n + 1$  stores the observed aggregated value  $y$ .

**env** Environment variable obtained from `fm.Init(n)`.

**kadd** is the value of  $k$ -tolerance, which is used for reducing the complexity of fuzzy measures. `kadd` is defined as an optional argument, its default value is `kadd = n`.

**Value**

**output** The output is an array of size  $2^n$  containing estimated standard fuzzy measure in binary ordering.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
```

```

0.843628, 0.996765, 0.999664, 0.930197,
0.611481, 0.92426, 0.266205, 0.334666,
0.297272, 0.840118, 0.0237427, 0.168081),
nrow=20,
ncol=4);
fm.fittingKtolerant(d,env,2)

```

---

fm.fittingMob

*Mobius Fuzzy Measure Fitting function*


---

### Description

Estimate values of the Mobius fuzzy measures from empirical data.

### Usage

```
fm.fittingMob(data, env=NULL ,kadd="NA")
```

### Arguments

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0,1]^n$ is a vector containing utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0,1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ store the observed aggregating value $y$ .
env	Environment variable obtained from <code>fm.Init(n)</code> .
kadd	is the value of $k$ -additivity, which is used for reducing the complexity of fuzzy measures. <code>kadd</code> is defined as an optional argument, its default value is <code>kadd = n</code> .

### Value

output	The output is an array of size $2^n$ containing estimated Mobius fuzzy measure in binary ordering.
--------	--

### Note

The fit might not be perfect, and not all the constraints can be fully met.

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
            nrow=20,
            ncol=4);
env<-fm.Init(3)
fm.fittingMob(d,env)
```

---

fm.fittingOWA

*Symmetric Fuzzy Measure Fitting function*


---

**Description**

Estimate values of the symmetric fuzzy measures from empirical data. The resulting Choquet integral is the OWA function.

**Usage**

```
fm.fittingOWA(data, env=NULL)
```

**Arguments**

**data** is the empirical data set in pairs  $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$  where  $x_i$  in  $[0, 1]^n$  is a vector containing utility values of  $n$  input criteria  $x_{i1}, x_{i2}, \dots, x_{in}$ ,  $y_i$  in  $[0, 1]$  is a single aggregated value given by decision makers. The data is stored as a matrix of  $M$  by  $n+1$  elements, where  $M$  is the number of data instances, and  $n$  is the number of input criteria, the column  $n + 1$  stores the observed aggregated value  $y$ .

**env** Environment variable obtained from `fm.Init(n)`.

**Value**

output            The output is an array of size n containing estimated OWA coefficients.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
             nrow=20,
             ncol=4);
fm.fittingOWA(d,env)
```

---

fm.fittingWAM

*Additive Fuzzy Measure Fitting function*

---

**Description**

Estimate values of an additive fuzzy measure from empirical data. In this case the Choquet integral is the weighted arithmetic mean WAM.

**Usage**

```
fm.fittingWAM(data, env=NULL)
```

**Arguments**

**data** is the empirical data set in pairs  $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$  where  $x_i$  in  $[0,1]^n$  is a vector containing utility values of  $n$  input criteria  $x_{i1}, x_{i2}, \dots, x_{in}$ ,  $y_i$  in  $[0,1]$  is a single aggregated value given by decision makers. The data is stored as a matrix of  $M$  by  $n+1$  elements, where  $M$  is the number of data instances, and  $n$  is the number of input criteria, the column  $n + 1$  stores the observed aggregated value  $y$ .

**env** Environment variable obtained from `fm.Init(n)`.

**Value**

**output** The output is an array of size  $n$  containing estimated weighting vector of WAM.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
0.808716, 0.584991, 0.479858, 0.544309,
0.350281, 0.895935, 0.822815, 0.625868,
0.746582, 0.174103, 0.858917, 0.480347,
0.71048, 0.513519, 0.303986, 0.387631,
0.0149841, 0.0914001, 0.364441, 0.134229,
0.147308, 0.165894, 0.988495, 0.388044,
0.445679, 0.11908, 0.00466919, 0.0897714,
0.00891113, 0.377869, 0.531647, 0.258585,
0.571167, 0.601746, 0.607147, 0.589803,
0.166229, 0.663025, 0.450775, 0.357412,
0.352112, 0.0570374, 0.607666, 0.270228,
0.783295, 0.802582, 0.519867, 0.583348,
0.301941, 0.875946, 0.726654, 0.562174,
0.955872, 0.92569, 0.539337, 0.633631,
0.142334, 0.462067, 0.235321, 0.228419,
0.862213, 0.209595, 0.779633, 0.498077,
0.843628, 0.996765, 0.999664, 0.930197,
0.611481, 0.92426, 0.266205, 0.334666,
0.297272, 0.840118, 0.0237427, 0.168081),
nrow=20,
```

```

        ncol=4);
fm.fittingWAM(d,env)

```

---

fm.FuzzyMeasureFitLP *Fuzzy Measure Fitting function.*

---

### Description

Estimates values of the fuzzy measures from empirical data. The result is an array containing the values of a standard fuzzy measure in binary ordering. kadd defines the complexity of fuzzy measure. If kadd is not provided, its default value is equal to the number of inputs.

### Usage

```

fm.FuzzyMeasureFitLP(data, env=NULL, kadd="NA",
    options=0, indexlow=(NULL), indexhigh=(NULL) , option1=0, orness=(NULL)
)

```

### Arguments

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0, 1]^n$ is a vector contains utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0, 1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ store the observed aggregating value $y$ .
env	Environment variable obtained from fm.Init(n).
kadd	is the value of $k$ -additivity, which is used for reducing the complexity of fuzzy measures. kadd is defined as an optional argument, its default value is $kadd = n$ . kadd is $k$ in $k$ -additive $f$ -measure, $1 < kadd < n+1$ ; if $kdd=n$ - f.m. is unrestricted
options	options (default value is 0) 1 - lower bounds on Shapley values supplied in indexlow, 2 - upper bounds on Shapley values supplied in indexhigh, 3 - lower and upper bounds on Shapley values supplied in indexlow and indexhigh, 4 - lower bounds on all interaction indices supplied in indexlow, 5 - upper bounds on all interaction indices supplied in indexhigh, 6 - lower and upper bounds on all interaction indices supplied in indexlow and indexhigh. All these value will be treated as additional constraints in the LP.
indexlow	array of size $n$ (options =1,2,3) or $m$ (options=4,5,6) containing the lower bounds on the Shapley values or interaction indices
indexhigh	array of size $n$ (options =1,2,3) or $m$ (options=4,5,6) containing the upper bounds on the Shapley values or interaction indices
option1	if the value is 1, the interval of orness values will be fitted (and the desired low and high orness values should be provided). If 0, no additional orness constraints.
orness	array of size 2, for example $c(0.1, 1)$



**Value**

output            The output is an array of size  $2^n$  containing estimated standard fuzzy measure in binary ordering.

**Note**

The fit might not be perfect, and not all the constraints can be fully met.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
             nrow=20,
             ncol=4);
env<-fm.Init(3)
fm.FuzzyMeasureFitLP(d,env)
indexlow=c(0.1,0.1,0.2);
indexhigh=c(0.9,0.9,0.5);
fm.FuzzyMeasureFitLP(d,env, kadd=2, indexlow, indexhigh,
options=3, option1=1, orness=c(0.1,0.7))
```

---

 fm.FuzzyMeasureFitLPMob

*Mobius Fuzzy Measure Fitting function, R wrapper for FuzzyMeasureFitLP() in fuzzymeasurefit.cpp*

---

### Description

Estimate values of the Mobius fuzzy measures from empirical data. The result is an array containing the values of the fuzzy measure in Mobius, ordered according to set cardinalities. kadd defines the complexity of fuzzy measure. if kadd is not provided, its default value is equal to the number of inputs.

### Usage

```
fm.FuzzyMeasureFitLPMob(data, env=NULL, kadd="NA",
  options=0, indexlow=NULL, indexhigh=NULL, option1=0, orness=NULL)
)
```

### Arguments

data	is the empirical data set in pairs $(x_1, y_1), (x_2, y_2), \dots, (x_d, y_d)$ where $x_i$ in $[0, 1]^n$ is a vector contains utility values of $n$ input criteria $x_{i1}, x_{i2}, \dots, x_{in}$ , $y_i$ in $[0, 1]$ is a single aggregated value given by decision makers. The data is stored as a matrix of $M$ by $n+1$ elements, where $M$ is the number of data instances, and $n$ is the number of input criteria, the column $n + 1$ store the observed aggregating value $y$ .
env	Environment variable obtained from <code>fm.Init(n)</code> .
kadd	is the value of $k$ -additivity, which is used for reducing the complexity of fuzzy measures. kadd is defined as an optional argument, its default value is $kadd = n$ . kadd is $k$ in $k$ -additive $f$ -measure, $1 < kadd < n+1$ ; if $kdd=n$ - $f.m.$ is unrestricted
options	options (default value is 0) 1 - lower bounds on Shapley values supplied in indexlow, 2 - upper bounds on Shapley values supplied in indexhigh, 3 - lower and upper bounds on Shapley values supplied in indexlow and indexhigh, 4 - lower bounds on all interaction indices supplied in indexlow, 5 - upper bounds on all interaction indices supplied in indexhigh, 6 - lower and upper bounds on all interaction indices supplied in indexlow and indexhigh. All these value will be treated as additional constraints in the LP.
indexlow	array of size $n$ (options = 1,2,3) or $m$ (options=4,5,6) containing the lower bounds on the Shapley values or interaction indices
indexhigh	array of size $n$ (options = 1,2,3) or $m$ (options=4,5,6) containing the upper bounds on the Shapley values or interaction indices
option1	if the value is 1, the interval of orness values will be fitted (and the desired low and high orness values should be provided). If 0, no additional orness constraints.
orness	array of size 2, for example $c(0.1, 1)$

**Value**

output            The output is an array of size  $2^n$  containing estimated Mobius fuzzy measure in binary ordering.

**Note**

The fit might not be perfect, and not all the constraints can be fully met.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
d <- matrix( c( 0.00125122, 0.563568, 0.193298, 0.164338,
               0.808716, 0.584991, 0.479858, 0.544309,
               0.350281, 0.895935, 0.822815, 0.625868,
               0.746582, 0.174103, 0.858917, 0.480347,
               0.71048, 0.513519, 0.303986, 0.387631,
               0.0149841, 0.0914001, 0.364441, 0.134229,
               0.147308, 0.165894, 0.988495, 0.388044,
               0.445679, 0.11908, 0.00466919, 0.0897714,
               0.00891113, 0.377869, 0.531647, 0.258585,
               0.571167, 0.601746, 0.607147, 0.589803,
               0.166229, 0.663025, 0.450775, 0.357412,
               0.352112, 0.0570374, 0.607666, 0.270228,
               0.783295, 0.802582, 0.519867, 0.583348,
               0.301941, 0.875946, 0.726654, 0.562174,
               0.955872, 0.92569, 0.539337, 0.633631,
               0.142334, 0.462067, 0.235321, 0.228419,
               0.862213, 0.209595, 0.779633, 0.498077,
               0.843628, 0.996765, 0.999664, 0.930197,
               0.611481, 0.92426, 0.266205, 0.334666,
               0.297272, 0.840118, 0.0237427, 0.168081),
            nrow=20,
            ncol=4);
env<-fm.Init(3)
fm.FuzzyMeasureFitLPMob(d,env)
indexlow=c(0.1,0.1,0.2);
indexhigh=c(0.9,0.9,0.5);
fm.FuzzyMeasureFitLPMob(d,env, kadd=2, indexlow, indexhigh,
options=3, option1=1, orness=c(0.1,0.7))
```

---

fm.Init	<i>Initialisation function</i>
---------	--------------------------------

---

**Description**

This function initialises the internal structures which makes computations faster. The structures are saved in the output environment variable, which should be subsequently passed to other functions. Several environment variables (for different dimensions) can be initialised at the same time.

**Usage**

```
fm.Init(n1)
```

**Arguments**

n1                    the number of variables.

**Value**

output                The output is the environment variable containing the internal structures.

**Author(s)**

Gleb Beliakov

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
```

---

fm.Interaction	<i>Interaction Index computation function</i>
----------------	---

---

**Description**

Calculates all the interaction indices of input criteria for standard fuzzy measure.

**Usage**

```
fm.Interaction(v,env)
```

**Arguments**

v	fuzzy measure value in standard representation
env	Environment variable obtained from fm.Init(n).

**Value**

output	the output is a matrix, whose first column stores the interaction index values, and the second column stores the indices of criteria in coalitions.
--------	---

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.Interaction(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.InteractionB	<i>Banzhaf Interaction Index computation function</i>
-----------------	---

---

**Description**

Calculates all the Banzhaf Interaction indices of input criteria for a standard fuzzy measure.

**Usage**

```
fm.InteractionB(v,env)
```

**Arguments**

v	fuzzy measure value in standard representation
env	Environment variable obtained from fm.Init(n).

**Value**

output	the output is a matrix, whose first column stores the Banzhaf Interaction index values, and the second column stores the indices of criteria in coalitions.
--------	---

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.InteractionB(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.InteractionBMob	<i>Banzhaf InteractionB Index computation function</i>
--------------------	--

---

**Description**

Calculates all the Banzhaf InteractionB indices of input criteria for a Mobius fuzzy measure.

**Usage**

```
fm.InteractionBMob(Mob,env)
```

**Arguments**

Mob	fuzzy measure value in Mobius representation
env	Environment variable obtained from fm.Init(n).

**Value**

output	the output is a matrix, whose first column stores the Banzhaf Interaction index values, and the second column stores the indices of criteria in coalitions.
--------	---

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

## Examples

```
env<-fm.Init(3)
fm.InteractionBMob(c( 0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.InteractionMob      *Interaction Index computation function for Mobius fuzzy measure*

---

## Description

Calculates all the interaction indices of input criteria for a Mobius fuzzy measure.

## Usage

```
fm.InteractionMob(Mob,env )
```

## Arguments

Mob	fuzzy measure value in Mobius representation
env	Environment variable obtained from fm.Init(n).

## Value

output      the output is a matrix, whose first column stores the interaction index values, and the second column stores the indices of criteria in coalitions.

## Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

## References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

## Examples

```
env<-fm.Init(3)
fm.InteractionMob(c( 0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.IsMeasureAdditive *IsMeasureAdditive function*

---

### Description

Returns 1 if yes, 0 if no; v is a fuzzy measure in standard representation.

### Usage

```
fm.IsMeasureAdditive(v,env)
```

### Arguments

v	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureAdditive integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureAdditive(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```



---

`fm.IsMeasureAdditiveMob`*IsMeasureAdditiveMob function*

---

**Description**

Returns 1 if yes, 0 if no;  $v$  is a fuzzy measure in Mobius representation.

**Usage**

```
fm.IsMeasureAdditiveMob(Mob, env)
```

**Arguments**

Mob	the Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from <code>fm.Init(n)</code> .

**Value**

output	The output is 1 if yes, 0 if no.
--------	----------------------------------

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureAdditive integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.IsMeasureAdditiveMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.IsMeasureBalanced *IsMeasureBalanced function*

---

### Description

Returns 1 if yes, 0 if no; v is a fuzzy measure in standard representation.

### Usage

```
fm.IsMeasureBalanced(v, env)
```

### Arguments

v	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureBalanced integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureBalanced(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.IsMeasureBalancedMob  
*IsMeasureBalancedMob function*

---

### Description

Returns 1 if yes, 0 if no; Mob is a fuzzy measure in Mobius representation.

### Usage

```
fm.IsMeasureBalancedMob(Mob, env)
```

### Arguments

Mob	a Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is 1 if yes, 0 if no.
--------	----------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureBalanced integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureBalancedMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.IsMeasureKmaxitive *IsMeasureKmaxitive function*

---

### Description

Returns k; v is a fuzzy measure in standard representation.

### Usage

```
fm.IsMeasureKmaxitive(v,env=NULL)
```

### Arguments

v	the general fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is k. If $k=n$ then not k-maxitive
--------	---

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A fuzzy integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureKmaxitive(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

`fm.IsMeasureKmaxitiveMob`*IsMeasureKmaxitive function*

---

**Description**

Returns k; mob is a fuzzy measure in Moebius representation.

**Usage**

```
fm.IsMeasureKmaxitiveMob(Mob,env=NULL)
```

**Arguments**

Mob	the general fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from <code>fm.Init(n)</code> .

**Value**

output	The output is k. If $k=n$ then not k-maxitive
--------	---

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A fuzzy integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.IsMeasureKmaxitiveMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.IsMeasureSelfdual *IsMeasureSelfdual function*

---

**Description**

Returns 1 if yes, 0 if no; v is a fuzzy measure in standard representation.

**Usage**

```
fm.IsMeasureSelfdual(v, env)
```

**Arguments**

v	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

**Value**

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSelfdual integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.IsMeasureSelfdual(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.IsMeasureSelfdualMob  
*IsMeasureSelfdualMob function*

---

### Description

Returns 1 if yes, 0 if no; Mob is a fuzzy measure in Mobius representation.

### Usage

```
fm.IsMeasureSelfdualMob(Mob, env)
```

### Arguments

Mob	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSelfdual integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)  
fm.IsMeasureSelfdualMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.IsMeasureSubadditive

*IsMeasureSubadditive function*

---

### Description

Returns 1 if yes, 0 if no; v is a fuzzy measure in standard representation.

### Usage

```
fm.IsMeasureSubadditive(v, env)
```

### Arguments

v	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSubadditive integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureSubadditive(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```



---

`fm.IsMeasureSubadditiveMob`*IsMeasureSubadditiveMob function*

---

**Description**

Returns 1 if yes, 0 if no; v is a fuzzy measure in Mobius representation.

**Usage**

```
fm.IsMeasureSubadditiveMob(Mob, env)
```

**Arguments**

Mob	a Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from <code>fm.Init(n)</code> .

**Value**

output	The output is 1 if yes, 0 if no.
--------	----------------------------------

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSubadditive integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.IsMeasureSubadditiveMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

`fm.IsMeasureSubmodular`*IsMeasureSubmodular function*

---

**Description**

Returns 1 if yes, 0 if no;  $v$  is a fuzzy measure in standard representation.

**Usage**

```
fm.IsMeasureSubmodular(v, env=NULL)
```

**Arguments**

$v$	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
$env$	Environment variable obtained from <code>fm.Init(n)</code> .

**Value**

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSubmodular integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.IsMeasureSubmodular(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.IsMeasureSubmodularMob  
*IsMeasureSubmodularMob function*

---

### Description

Returns 1 if yes, 0 if no; Mob is a fuzzy measure in Mobius representation.

### Usage

```
fm.IsMeasureSubmodularMob(Mob, env=NULL)
```

### Arguments

Mob	a Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is 1 if yes, 0 if no.
--------	----------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSubmodular integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureSubmodularMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.IsMeasureSuperadditive  
*IsMeasureSuperadditive function*

---

**Description**

Returns 1 if yes, 0 if no; v is a fuzzy measure in standard representation.

**Usage**

```
fm.IsMeasureSuperadditive(v, env=NULL)
```

**Arguments**

v	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

**Value**

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSuperadditive integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)  
fm.IsMeasureSuperadditive(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.IsMeasureSuperadditiveMob  
*IsMeasureSuperadditiveMob function*

---

### Description

Returns 1 if yes, 0 if no; Mob is a fuzzy measure in Mobius representation.

### Usage

```
fm.IsMeasureSuperadditiveMob(Mob, env=NULL)
```

### Arguments

Mob	a Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is 1 if yes, 0 if no.
--------	----------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSuperadditive integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureSuperadditiveMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.IsMeasureSupermodular

*IsMeasureSupermodular function*

---

### Description

Returns 1 if yes, 0 if no; v is a fuzzy measure in standard representation.

### Usage

```
fm.IsMeasureSupermodular(v, env=NULL)
```

### Arguments

v	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSupermodular integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureSupermodular(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.IsMeasureSupermodularMob  
*IsMeasureSupermodularMob function*

---

### Description

Returns 1 if yes, 0 if no; Mob is a fuzzy measure in Mobius representation.

### Usage

```
fm.IsMeasureSupermodularMob(Mob, env=NULL)
```

### Arguments

Mob	a Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is 1 if yes, 0 if no.
--------	----------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSupermodular integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureSupermodularMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.IsMeasureSymmetric *IsMeasureSymmetric function*

---

### Description

Returns 1 if yes, 0 if no; v is a fuzzy measure in standard representation.

### Usage

```
fm.IsMeasureSymmetric(v,env=NULL)
```

### Arguments

v	the genral fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output	The ouput is 1 if yes, 0 if no.
--------	---------------------------------

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A IsMeasureSymmetric integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.IsMeasureSymmetric(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```



---

`fm.IsMeasureSymmetricMob`*IsMeasureSymmetricMob function*

---

**Description**

Returns 1 if yes, 0 if no;  $v$  is a fuzzy measure in Mobius representation.

**Usage**

```
fm.IsMeasureSymmetricMob(Mob, env=NULL)
```

**Arguments**

Mob	a Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from <code>fm.Init(n)</code> .

**Value**

output	The output is 1 if yes, 0 if no.
--------	----------------------------------

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A `IsMeasureSymmetric` integral toolbox and its application in customer preference analysis, in *Data mining applications with R*, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.IsMeasureSymmetricMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.Mobius	<i>Mobius transform function</i>
-----------	----------------------------------

---

**Description**

calculates Mobius representation of general fuzzy measure, the input and output is an array of size  $2^n=m$  in binary ordering.

**Usage**

```
fm.Mobius(v,env=NULL)
```

**Arguments**

v	fuzzy measure value in standard representation.
env	Environment variable obtained from fm.Init(n).

**Value**

output	the output is the fuzzy measure in Mobius representation.
--------	---

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
##---- Should be DIRECTLY executable !! ----  
env<-fm.Init(3)  
fm.Mobius(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.NonadditivityIndex *Nonadditivity index computation function*

---

**Description**

Calculates the nonadditivity indices of input criteria from general fuzzy measure.

**Usage**

```
fm.NonadditivityIndex(v,env=NULL)
```

**Arguments**

v	is fuzzy measure in general representation.
env	Environment variable obtained from fm.Init(n).

**Value**

output	The output is an array of size $2^n$ , which contain nonadditivity indices of input criteria coalitions.
--------	--

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.NonadditivityIndex(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.NonadditivityIndexMob

*nonadditivity index computation function*

---

### Description

Calculates the nonadditivity indices of input criteria from general fuzzy measure in Moebius representation.

### Usage

```
fm.NonadditivityIndexMob(Mob,env=NULL)
```

### Arguments

Mob	is fuzzy measure in Moebius representation.
env	Environment variable obtained from fm.Init(n).

### Value

output	The output is an array of size $2^n$ , which contain nonadditivity indices of input criteria coalitions.
--------	--

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.NonadditivityIndexMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.OrnessChoquet	<i>OrnessChoquet function</i>
------------------	-------------------------------

---

**Description**

Calculates orness value of the Choquet integral of the fuzzy measure, where  $v$  is a standard representation.

**Usage**

```
fm.OrnessChoquet(v, env=NULL)
```

**Arguments**

$v$	a standard fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
$env$	Environment variable obtained from <code>fm.Init(n)</code> .

**Value**

output	the output is the orness the Choquet integral for the fuzzy measure.
--------	--

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.OrnessChoquet(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.OrnessChoquetMob    *OrnessChoquetMob function*

---

### Description

Calculates orness value of the Choquet integral of the fuzzy measure, where Mob is the Mobius representation.

### Usage

```
fm.OrnessChoquetMob(Mob, env=NULL)
```

### Arguments

Mob	the Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init(n).

### Value

output            the output is the orness the Choquet integral for the fuzzy measure.

### Author(s)

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

### References

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

### Examples

```
env<-fm.Init(3)
fm.OrnessChoquetMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.Shapley	<i>Shapley value computation function</i>
------------	---

---

**Description**

Calculates the Shapley values of input criteria from general fuzzy measure,

**Usage**

```
fm.Shapley(v, env=NULL)
```

**Arguments**

v	is fuzzy measure in general representation.
env	Environment variable obtained from fm.Init(n).

**Value**

output	The output is an array of size n, which contain Shapley values of input criteria.
--------	---

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.Shapley(c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.ShapleyMob	<i>Shapley value computation function</i>
---------------	---

---

**Description**

Calculates the Shapley values of input criteria from Mobius fuzzy measure,

**Usage**

```
fm.ShapleyMob(Mob, env=NULL)
```

**Arguments**

Mob	is fuzzy measure in Mobius representation.
env	Environment variable obtained from fm.Init(n).

**Value**

output            The output is an array of size n, which contain Shapley values of input criteria.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.ShapleyMob(c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```



---

fm.Sugeno	<i>Sugeno function</i>
-----------	------------------------

---

**Description**

Calculates the value of a Sugeno integral of input  $x$ , with fuzzy measure in standard representation

**Usage**

```
fm.Sugeno(x, v, env=NULL)
```

**Arguments**

$x$	input vector of size $n$ , containing utility value of input criteria. $x$ is in $[0,1]$ .
$v$	the general fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
$env$	Environment variable obtained from <code>fm.Init(n)</code> .

**Value**

output	The output is a single value of the computed Sugeno integral.
--------	---

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A Sugeno integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.Sugeno(c(0.6, 0.3, 0.8), c(0, 0.3, 0.5, 0.6, 0.4, 0.8, 0.7, 1),env)
```

---

fm.SugenoMob	<i>Sugeno function in Mobius representation</i>
--------------	---

---

**Description**

Calculates the value of a Sugeno integral of input  $x$ , with fuzzy measure in mobius representation

**Usage**

```
fm.SugenoMob(x, Mob, env=NULL)
```

**Arguments**

$x$	input vector of size $n$ , containing utility value of input criteria. $x$ is in $[0,1]$ .
Mob	the Mobius fuzzy measure of size $m=2^n$ . Its values can be provided by users, or by estimating from empirical data.
env	Environment variable obtained from fm.Init( $n$ ).

**Value**

output	The output is a single value of the computed Sugeno integral.
--------	---

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A Sugeno integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.SugenoMob(c(0.6, 0.3, 0.8), c(0.0, 0.3, 0.5, -0.2, 0.4, 0.1, -0.2, 0.1),env)
```

---

fm.test	<i>A Test function</i>
---------	------------------------

---

**Description**

This function provide some examples of how fuzzy measure operation in this toolbox are used. It can be used to test if the toolbox has been installed successfully or not.

**Usage**

```
fm.test()
```

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
fm.test()
```

---

fm.Zeta	<i>Zeta transform function</i>
---------	--------------------------------

---

**Description**

Calculates the general fuzzy measure from Mobius representation. The input and output is an array of size  $2^n = m$  in binary ordering. This is the inverse of the Mobius function.

**Usage**

```
fm.Zeta(Mob, env)
```

**Arguments**

Mob	fuzzy measure value in Mobius representation.
env	Environment variable obtained from fm.Init(n).

**Value**

output                    the output is the fuzzy measure in general representation.

**Author(s)**

Gleb Beliakov, Andrei Kelarev, Quan Vu, Deakin University

**References**

[1] <http://www.deakin.edu.au/~gleb/fmtools.html> [2] H.Q. Vu, G. Beliakov, G. Li, 2014, A choquet integral toolbox and its application in customer preference analysis, in Data mining applications with R, Elsevier Inc., Waltham, MA., pp.247-272.

**Examples**

```
env<-fm.Init(3)
fm.Zeta(c(0.0,0.3,0.5,-0.2,0.4,0.1,-0.2,0.1),env)
```

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