## Package 'kstMatrix'

July 22, 2025

**Type** Package **Version** 1.2-0

Date 2025-04-02

**Title** Basic Functions in Knowledge Space Theory Using Matrix Representation

Description Knowledge space theory by Doignon and Falmagne (1999) <a href="doi:10.1007/978-3-642-58625-5">doi:10.1007/978-3-642-58625-5</a>> is a set- and order-theoretical framework, which proposes mathematical formalisms to operationalize knowledge structures in a particular domain. The 'kstMatrix' package provides basic functionalities to generate, handle, and manipulate knowledge structures and knowledge spaces. Opposed to the 'kst' package, 'kstMatrix' uses matrix representations for knowledge structures. Furthermore, 'kstMatrix' contains several knowledge spaces developed by the research group around Cornelia Dowling through querying experts.

**Depends** R (>= 4.4.0)

Suggests knitr, rmarkdown,

Imports stats, igraph, grDevices, sets, pks

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NeedsCompilation yes

Repository CRAN

**Encoding** UTF-8

LazyData true

RoxygenNote 7.3.2

VignetteBuilder knitr

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**Date/Publication** 2025-04-02 12:40:02 UTC

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cad	Knowledge spaces on AutoCAD knowledge	

## Description

Bases of knowledge spaces on AutoCAD knowledge obtained from querying experts.

## Usage

cad

## **Format**

A list containing seven bases (cad1 to cad6, and cadmaj) in binary matrix form. Each matrix has 28 columns representing the different knowledge items and a varying number of rows containing the basis elements.

fractions 3

#### **Details**

Six experts were queried about prerequisite relationships between 28 AutoCAD knowledge items (Dowling, 1991; 1993). A seventh basis represents those prerequisite relationships on which the majority (4 out of 6) of the experts agree (Dowling & Hockemeyer, 1998).

#### References

Dowling, C. E. (1991). Constructing Knowledge Structures from the Judgements of Experts. Habilitationsschrift, Technische Universität Carolo-Wilhelmina, Braunschweig, Germany.

Dowling, C. E. (1993). Applying the basis of a knowledge space for controlling the questioning of an expert. Journal of Mathematical Psychology, 37, 21–48.

Dowling, C. E. & Hockemeyer, C. (1998). Computing the intersection of knowledge spaces using only their basis. In Cornelia E. Dowling, Fred S. Roberts, & Peter Theuns, editors, Recent Progress in Mathematical Psychology, pp. 133–141. Lawrence Erlbaum Associates Ltd., Mahwah, NJ.

#### See Also

Other Data: fractions, readwrite, xpl

fractions

Knowledge spaces on fractions

#### **Description**

Bases of knowledge spaces on fractions obtained from querying experts.

#### Usage

fractions

## Format

A list containing four bases (frac1 to frac3, and fracmaj) in binary matrix form. Each matrix has 77 columns representing the different knowledge items and a varying number of rows containing the basis elements.

#### **Details**

Three experts were queried about prerequisite relationships between 77 items on fractions (Baumunk & Dowling, 1997). A forth basis represents those prerequisite relationships on which the majority of the experts agree (Dowling & Hockemeyer, 1998).

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#### References

Baumunk, K. & Dowling, C. E. (1997). Validity of spaces for assessing knowledge about fractions. Journal of Mathematical Psychology, 41,99–105.

Dowling, C. E. & Hockemeyer, C. (1998). Computing the intersection of knowledge spaces using only their basis. In Cornelia E. Dowling, Fred S. Roberts, & Peter Theuns, editors, Recent Progress in Mathematical Psychology, pp. 133–141. Lawrence Erlbaum Associates Ltd., Mahwah, NJ.

#### See Also

Other Data: cad, readwrite, xpl

kmbasis

Compute the basis of a knowledge space

## **Description**

kmbasis returns a matrix representing the basis of a knowledge space. If x is a knowledge structure or an arbitrary family of sets kmreduction returns the basis of the smallest knowledge space containing x.

## Usage

kmbasis(x)

## **Arguments**

Х

Binary matrix representing a knowledge space

## Value

Binary matrix representing the basis of the knowledge space.

#### See Also

Other Different representations for knowledge spaces: kmsurmisefunction(), kmsurmiserelation(), kmunionclosure()

## **Examples**

kmbasis(xpl\$space)

kmbasisdiagram 5

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kmhasi	lsdiagram	Plc

Plot the Hasse diagram of a basis stored as a matrix

## Description

kmbasisdiagram takes a matrix representing a basis and a color vector and draws a Hasse diagram. If the color vector is NULL the states are drawn in green.

## Usage

```
kmbasisdiagram(struc, horizontal = TRUE, colors = NULL)
```

## **Arguments**

struc Binary matrix representing a basis

horizontal Boolean defining orientation of the graph, default TRUE

colors Color vector (default NULL)

## See Also

Other Plotting knowledge structures: kmSRdiagram(), kmhasse()

kmcolors

Determine a color vector based on probabilities

## **Description**

kmcolors takes a probabilty vector and a color palette and creates a color vector to be used with kmhasse.

## Usage

```
kmcolors(prob, palette = cm.colors)
```

## **Arguments**

prob Probability vector

palette Color palette (default = cm.colors)

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kmdist

Compute the distance between a data set and a knowledge structure

## **Description**

kmdist returns a named vector with the frequencies of distances between a set of response patterns and a knowledge structure. This vector can be used to compute, e.g., the Discrepancy Index (DI) or the Distance Agreement Coefficient (DA).

## Usage

```
kmdist(data, struct)
```

#### **Arguments**

data Binary matrix representing a set of response patterns struct Binary matrix representing a knowledge structure

#### Value

Distance distribution vector

#### See Also

Other Validating knowledge spaces: kmSRvalidate(), kmvalidate()

## **Examples**

```
kmdist(xpl$data, xpl$space)
```

kmegreduction

Reduce a family of knowledge states with respect to item equivalence

## Description

kmeqreduction takes a family of knowledge states and returns its reduction to non-equivalent items.

#### Usage

```
kmeqreduction(x)
```

#### **Arguments**

Χ

Binary matrix

kmfringe 7

## Value

Binary matrix reduced by equivalences

#### See Also

Other Properties of knowledge structures: kmiswellgraded(), kmnotions()

## **Examples**

kmeqreduction(xpl\$space)

kmfringe

Compute the fringe of a state within a knowledge structure

## Description

kmfringe computes the fringe of a state within a knowledge structure, i.e. the set of items by which the state differs from its neighbours.

## Usage

```
kmfringe(state, struct)
```

## **Arguments**

state Binary vector representing a knowledge state

struct Binary matrix representing a knowledge structure

## Value

Binary vector representing the fringe

#### See Also

Other Neighbourhood & fringe: kmneighbourhood()

## **Examples**

```
kmfringe(c(1,0,0,0), xpl$space)
```

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kmgenerate

Generate a knowledge structure from a set of response patterns

#### **Description**

kmgenerate returns a matrix representing a knowledge structure generated from data. It uses a simplistic approach: patterns with a frequency above a specified threshold are considered as knowledge states. If the specified threshold is 0 (default) a real threshold is computed as N (number of response patterns) divided by 2^IQl. Please note that the number of response patterns should be much higher than the size of the power set of the item set Q. A factor of art least 10 is recommended. Currently, the number of items is limited to the number of bits in a C long minus one (i.e. 31 under Windows and 63 otherwise). But we would probably run into memory problems way earlier anyway.

#### Usage

```
kmgenerate(x, threshold = 0)
```

## **Arguments**

x Binary matrix representing a data set

threshold Threshold for taking response patterns as knowledge states (default 0)

#### Value

Binary matrix representing the generated knowledge structure

#### **Examples**

```
kmgenerate(xpl$sim, 15)
```

kmhasse

Plot the Hasse diagram of a knowledge structure stored as a matrix

## Description

kmhasse takes a matrix representing a knowledge structure and a color vector and draws a Hasse diagram. If the color vector is NULL the states are drawn in green.

## Usage

```
kmhasse(struc, horizontal = TRUE, colors = NULL)
```

kmiswellgraded 9

#### **Arguments**

struc Binary matrix representing a knowledge structure

horizontal Boolean defining orientation of the graph, default TRUE

colors Color vector (default NULL)

#### See Also

Other Plotting knowledge structures: kmSRdiagram(), kmbasisdiagram()

kmiswellgraded

Check for wellgradedness of a knowledge structure

## Description

kmiswellgraded returns whether a knowledge structure is wellgraded.

## Usage

```
kmiswellgraded(x)
```

## **Arguments**

Х

Binary matrix representing a knowledge space

## Value

Logical value specifying whether 'x' is wellgraded

## References

Doignon, J.-P. & Falmagne, J.-C. (1999). Knowledge Spaces. Springer-Verlag, Berlin.

## See Also

Other Properties of knowledge structures: kmeqreduction(), kmnotions()

## **Examples**

```
kmiswellgraded(xpl$space)
```

10 kmnneighbourhood

kmneighbourhood

Compute the neighbourhod of a state within a knowledge structure

#### **Description**

kmneighbourhood computes the neighbourhood of a state within a knowledge structure, i.e. the family of all other states with a symmetric set difference of 1.

## Usage

```
kmneighbourhood(state, struct)
```

#### **Arguments**

state Binary vector representing a knowledge state struct Binary matrix representing a knowledge structure

#### Value

Matrix containing the neighbouring states, one per row

## See Also

```
Other Neighbourhood & fringe: kmfringe()
```

## **Examples**

```
kmneighbourhood(c(1,1,0,0), xpl\$space)
```

kmnneighbourhood

Compute the n-neighbourhod of a state within a knowledge structure

#### **Description**

kmnneighbourhood computes the n-neighbourhood of a state within a knowledge structure, i.e. the family of all other states with a symmetric set difference maximal n.

## Usage

```
kmnneighbourhood(state, struct, distance)
```

#### **Arguments**

state Binary vector representing a knowledge state
struct Binary matrix representing a knowledge structure

distance Size of the n-neighbourhood

kmnotions 11

## Value

Matrix containing the neighbouring states, one per row

## **Examples**

```
kmnneighbourhood(c(1,1,0,0), xpl\$space, 2)
```

kmnotions

Determine the notions of a knowledge structure

## Description

kmnotions returns a matrix representing the notions of a knowledge structure.

#### Usage

```
kmnotions(x)
```

## **Arguments**

Х

Binary matrix representing a knowledge structure

## Value

Binary matrix representing notions in the knowledge structure

The matrix has a '1' in row 'i' and column 'j' if 'i' and 'j' belong to the same notion (i.e. are equivalent). It is a symmetric matrix with '1's in the main diagonal.

## See Also

Other Properties of knowledge structures: kmeqreduction(), kmiswellgraded()

## **Examples**

```
kmnotions(xpl$space)
```

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kmsf2basis

Derive a basis from a surmise function

#### **Description**

kmsf2basis expects a surmise function data frame and returns the corresponding basis.

#### Usage

```
kmsf2basis(sf)
```

#### **Arguments**

sf

Surmise function

#### Value

Matrix representing the basis.

kmsimulate

Simulate a set of response patterns according to the BLIM

## **Description**

kmsimulate returns a data set of n simulated response patterns based on the knowledge structure x given as a binary matrix. The simulation follows the BLIM (Basic Local Independence Model; see Doigon & Falmagne, 1999).

## Usage

```
kmsimulate(x, n, beta, eta)
```

#### **Arguments**

x Binary matrix representing a knowledge space

Number of simulated response patterns
 Careless error probability value or vector
 Lucky guess probability value or vector

## **Details**

The beta and eta parameters must be either single numericals or vectors with a length identical to the number of rows in the x matrix. A mixture is possible.

The 'sample' function used by 'kmsimulate' might work inaccurately for knowledge structures 'x' with 2^31 or more states.

kmSRdiagram 13

## Value

Binary matrix representing the simulated data set

#### References

Doignon, J.-P. & Falmagne, J.-C. (1999). Knowledge Spaces. Springer-Verlag, Berlin.

## **Examples**

```
kmsimulate(xpl$space, 50, 0.2, 0.1) kmsimulate(xpl$space, 50, c(0.2, 0.25, 0.15, 0.2), c(0.1, 0.15, 0.05, 0.1)) kmsimulate(xpl$space, 50, c(0.2, 0.25, 0.15, 0.2), 0)
```

kmSRdiagram

Plot the Hasse diagram of a basis stored as a matrix

## **Description**

kmSRdiagram takes a matrix representing a surmise relation and a color vector and draws a Hasse diagram. If the color vector is NULL the states are drawn in green.

## Usage

```
kmSRdiagram(structure, horizontal = TRUE, colors = NULL)
```

## Arguments

structure Binary matrix representing a surmise relation

horizontal Boolean defining orientation of the graph, default TRUE

colors Color vector (default NULL)

## See Also

Other Plotting knowledge structures: kmbasisdiagram(), kmhasse()

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kmSRvalidate

Validate a surmise relation against a data set

## Description

kmSRvalidate returns a list with two elements, Goodman & Kruskal's gamma value and the violational coefficient (VC).

#### Usage

```
kmSRvalidate(data, sr)
```

## **Arguments**

data Binary matrix representing a set of response patterns

sr Binary matrix representing a surmise relation

#### Value

A list with two elements:

gamma Goodman & Kruskal's gamma index

VC Viola<tional Coefficient

#### See Also

Other Validating knowledge spaces: kmdist(), kmvalidate()

## **Examples**

```
kmSRvalidate(xpl$data, xpl$sr)
```

kmsurmisefunction

Compute the surmise function for a knowledge space or basis

## Description

kmsurmisefunction returns a data frame representing the surmise function for a knowledge space or basis. The rows of the data frame are ordered by item name.

## Usage

kmsurmisefunction(x)

kmsurmiserelation 15

## Arguments

x Binary matrix representing a knowledge space or basis

#### Value

Data frame representing the surmise unction of x.

#### See Also

Other Different representations for knowledge spaces: kmbasis(), kmsurmiserelation(), kmunionclosure()

#### **Examples**

kmsurmisefunction(xpl\$space)

kmsurmiserelation

Compute the surmise relation of a quasi-ordinal knowledge space

## Description

kmsurmiserelation returns a matrix representing the surmise relation of a quasi-ordinal knowledge space. If x is a general knowledge space, a knowledge structure or an arbitrary family of sets, kmsurmiserelation returns the surmise relation of the smallest quasi-ordinal knowledge space containing x.

## Usage

kmsurmiserelation(x)

#### **Arguments**

x Binary matrix representing a quasi-ordinal knowledge space

#### Value

Binary matrix representing the surmise relation of the corresponding quasi-ordinal knowledge space Note: The columns of the surmise relation matrix describe the minimal state for the respective item in the quasi-ordinal knowledge space.

#### See Also

Other Different representations for knowledge spaces: kmbasis(), kmsurmisefunction(), kmunionclosure()

## **Examples**

kmsurmiserelation(xpl\$space)

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kmsymmsetdiff

Compute the symmetric set difference between two sets

## Description

Compute the symmetric set difference between two sets

## Usage

```
kmsymmsetdiff(x, y)
kmsetdistance(x, y)
```

## Arguments

- x Binary vector representing a set
- y Binary vector representing a set

## Value

kmsymmsetdiff: Symmetric set difference between 'x' and 'y'

kmsetdistance: Distance between the sets 'x' and 'y', i.e. the cardinality of the symmetric set difference

## **Examples**

```
kmsymmsetdiff(c(1,0,0), c(1,1,0))
kmsetdistance(c(1,0,0), c(1,1,0))
```

kmtrivial

Create trivial knowledge spaces

## **Description**

These functions create trivial knowledge spaces of a given item number. The minimal space contains just the empty set and the full item set while the maximal space is equal to the power set.

## Usage

```
kmminimalspace(noi)
kmmaximalspace(noi)
```

kmunionclosure 17

## **Arguments**

noi

Number of items

#### **Details**

Please note that the computation time for creating large power sets can grow quite large easily.

#### Value

A binary matrix representing the respective knowledge space

## **Examples**

```
kmminimalspace(5)
kmmaximalspace(5)
```

kmunionclosure

Close a family of sets under union

## **Description**

kmunionclosure returns a matrix representing a knowledge space. Please note that it may take quite some time for computing larger knowledge spaces.

## Usage

kmunionclosure(x)

#### **Arguments**

Х

Binary matrix representing a family of sets

#### Value

Binary matrix representing the corresponding knowledge space, i.e. the closure of the family under union including the empty set and the full set.

kmunionclosure implements the irredundant algorithm developed by Dowling (1993).

## References

Dowling, C. E. (1993). On the irredundant construction of knowledge spaces. Journal of Mathematical Psychology, 37, 49–62.

#### See Also

Other Different representations for knowledge spaces: kmbasis(), kmsurmisefunction(), kmsurmiserelation()

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#### **Examples**

kmunionclosure(xpl\$basis)

kmvalidate

Validate a knowledge structure against a data set

## **Description**

kmvalidate returns a list with three elements, a named vector (dist) with the frequencies of distances between a set of response patterns and a knowledge structure, the Discrepancy Index (DI), and the Distance Agreement Coefficient (DA).

#### Usage

```
kmvalidate(data, struct)
```

#### **Arguments**

data Binary matrix representing a set of response patterns struct Binary matrix representing a knowledge structure

## Value

A list with three elements:

dist Distance distribution vector

**DI** Discrepancy Index

**DA** Distance Agreement Coefficient

## Warning

The DA computation can take quite some time for larger item sets as the power set has to be computed. For item sets with around 30 items or more, it may even crash the system due to huge memory requests.

#### See Also

```
Other Validating knowledge spaces: kmSRvalidate(), kmdist()
```

## **Examples**

```
kmvalidate(xpl$data, xpl$space)
```

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readwrite

Knowledge spaces on reading and writing abilities

## Description

Bases of knowledge spaces on reading/writing abilities obtained from querying experts.

## Usage

readwrite

#### **Format**

A list containing four bases (rw1 to rw3, and rwmaj) in binary matrix form. Each matrix has 48 columns representing the different knowledge items and a varying number of rows containing the basis elements.

#### **Details**

Three experts were queried about prerequisite relationships between 48 items on reading and writing abilities (Dowling, 1991; 1993). A forth basis represents those prerequisite relationships on which the majority of the experts agree (Dowling & Hockemeyer, 1998).

#### References

Dowling, C. E. (1991). Constructing Knowledge Structures from the Judgements of Experts. Habilitationsschrift, Technische Universität Carolo-Wilhelmina, Braunschweig, Germany.

Dowling, C. E. (1993). Applying the basis of a knowledge space for controlling the questioning of an expert. Journal of Mathematical Psychology, 37, 21–48.

Dowling, C. E. & Hockemeyer, C. (1998). Computing the intersection of knowledge spaces using only their basis. In Cornelia E. Dowling, Fred S. Roberts, & Peter Theuns, editors, Recent Progress in Mathematical Psychology, pp. 133–141. Lawrence Erlbaum Associates Ltd., Mahwah, NJ.

## See Also

Other Data: cad, fractions, xpl

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xpl

Small example knowledge space

## Description

Basis and space matrix, surmise relation and surmise function of a small fictional knowledge space, and two data sets (data (7 patterns) and sim (500 patterns) to be used in examples. The latter was produced from the space with kmsimulate() with beta and eta values of 0.1.

## Usage

xpl

## **Format**

A list containing the basis, the space, the surmise relation, the surmise function, and the two data matrices data and sim.

## See Also

Other Data: cad, fractions, readwrite

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