BERTIN MATRICES

AN ℝ IMPLEMENTATION

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in preparation

CONTENTS

1. BERTIN PLOTS

Among the rich material on graphical presentation of information, in (Bertin, 1977) (engl. (Bertin, 1999)) J. Bertin discusses the presentation of data matrices, with a particular view to seriation. (de Falguerolles et al., 1997) gives an appraisal of this aspect of Bertin's work. The methods illustrated in (de Falguerolles et al., 1997) have been first implemented in the Voyager system (Sawitzki, 1996). They have been partially re-implemented in R, and this paper gives an introduction to the R-implementation.

The R-implementation can be downloaded as a package `bertin` from `http://bertin.r-forge.r-project.org/`. (de Falguerolles et al., 1997) is included in the documentation section of the package.

Bertin uses a small data set on hotel occupancy data to illustrate his ideas.

	Jan	Fev	Mars	Avril	May	Juin	Juil	Aout	Sept	Oct	Nov	Dec
ClienteleFeminine	26	21	26	28	20	20	20	20	20	40	15	40
Locale	69	70	77	71	37	36	39	39	55	60	68	72
USA	7	6	3	6	23	14	19	14	9	6	8	8
AmerSud	0	0	0	0	8	6	6	4	2	12	0	0
Europe	20	15	14	15	23	27	22	30	27	19	19	17
MOrientAfrique	1	0	0	8	6	4	6	4	2	1	0	1
Asie	3	10	6	0	3	13	8	9	5	2	5	2
Business	78	80	85	86	85	87	70	76	87	85	87	80
Touristes	22	20	15	14	15	13	30	24	13	15	13	20
ResDirecte	70	70	78	74	69	68	74	75	68	68	64	75
ResAgents	20	18	19	17	27	27	19	19	26	27	21	15
EquipageAeriens	10	12	6	9	4	5	7	6	6	5	15	10
MoinsDe20	2	2	4	2	2	1	1	2	2	4	2	5
De20a55	25	27	37	35	25	25	27	28	24	30	24	30
De35a55	48	49	42	48	54	55	53	51	55	46	55	43
PlusDe55	25	22	17	15	19	19	19	19	19	20	19	22
Prix	163	167	166	174	152	155	145	170	157	174	165	156
Duree	1.65	1.71	1.65	1.91	1.9	2	1.54	1.6	1.73	1.82	1.66	1.44
Occupation	67	82	70	83	74	77	56	62	90	92	78	55
Foires	0	0	0	1	1	1	0	0	1	1	1	1

TABLE 1. Bertin's hotel data

2. BERTIN MATRICES

To repeat from (de Falguerolles et al., 1997): In abstract terms, a Bertin matrix is a matrix of displays. Bertin matrices allow rearrangements to transform an initial matrix to a more homogeneous structure. The rearrangements are row or column permutations, and groupings of rows or columns. To fix ideas, think of a data matrix, variable by case, with real valued variables. For each variable, draw a bar chart of variable value by case. Highlight all bars representing a value above some sample threshold for that variable. See Figure 1.

Variables are collected in a matrix to display the complete data set. Figure 2. By convention, Bertin shows variables in rows and cases in columns. To make periodic structures more visible, Bertin repeats the data cyclically.

```
_____ Output _____
parasp:  0.8170732 1.2 0.6808943
changing.
```
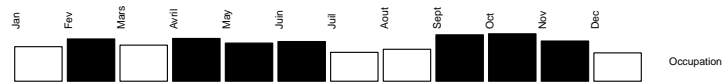
FIGURE 1. Display of one variable. Observations above average are highlighted.
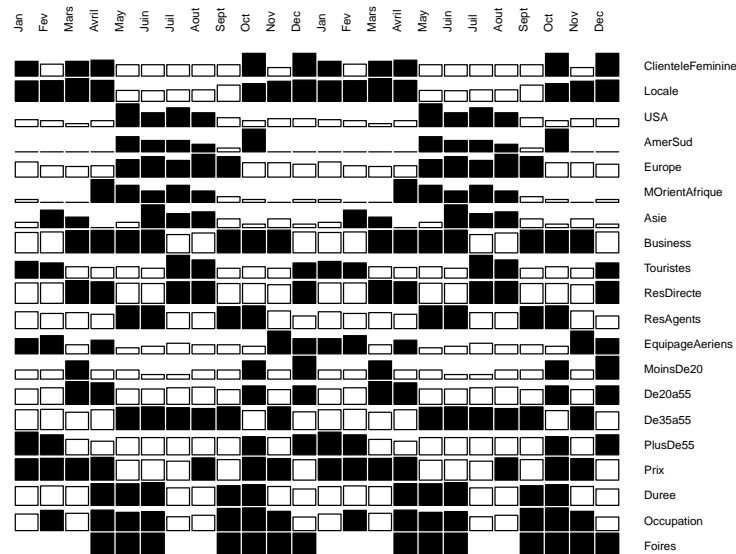

FIGURE 2. Display of a data matrix. Variables are shown as rows. Time axis is duplicated.

scaled

As Bertin points out, the indexing used is arbitrary. You can rearrange rows and/or columns to reveal the information of interest. If you run a hotel, of course the percentage of hotel occupation and the duration of the visits are most interesting for you. Move these variables to the top of the display, and rearrange the other variables by similarity or dissimilarity to these target variables.

```
───────────────────────────────── Output ──────────────────────────────────
parasp:  0.8153846 1.2 0.6794872
changing.
```

As a second example, we use the the *USJudgeRatings* data set.

```
───────────────────────────────── Output ──────────────────────────────────
```
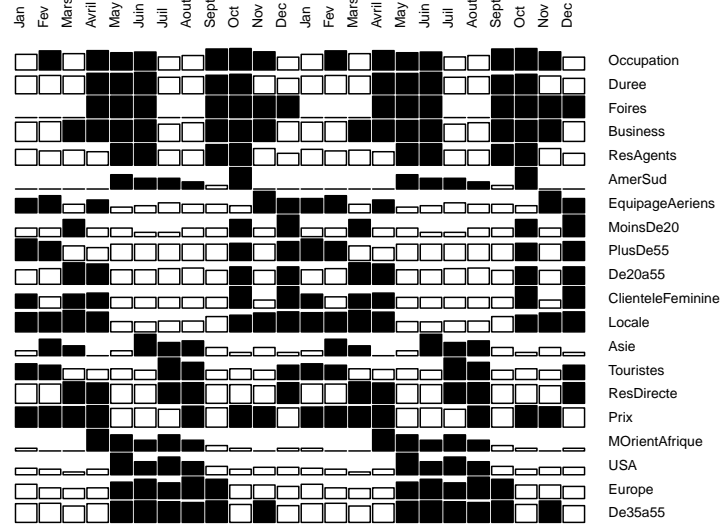
FIGURE 3. Display of a data matrix. Variables rearranged by similarity to occupation and duration.

```
parasp:  1.035971 3.583333 0.2891082
changing.
```

At this early point, let us put Bertin's work in place. Visualizing information is but one aspect. In statistics, as we see it today, visualization may be one part of an analysis. The outcome will be a decision leading to an action, and then there is a loss (or gain) depending on the action taken on the one hand, and the "true" state of the world on the other. Statistics has formulated a few standard problems, and given a suggestion to handle these. In our example, the problem can be seen as a prediction problem: find a prediction model to predict occupation and duration, based on the other variables. This is a control problem, and the statistical contribution is to find a regression model for occupation and duration, based on the other variables. The visualization can be seen as one way to hint at a regression model. There are few classical problems. Regression is one of them, and prediction is closely related. Classification and clustering is another, closely related pair of problems, and their relation to Bertin matrices should be obvious.
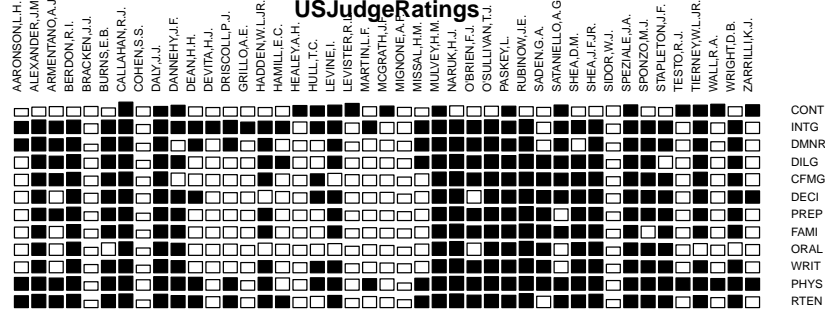
FIGURE 4. Display of a data matrix. Lawyers' ratings of state judges in the US Superior Court.

3. WORK FLOW

We prefer to see Bertin matrices as a part of a work flow.

In a first step, we transfer the input data to allow for common, or comparable scales. In the Hotel example, Bertin rescales by the maximum value of each variable. The dichotomous variable Faires is recoded as 0/1. Our implementation default is to rescale for *(0, max)* for positive variables, *(min, 0)* for negative variables, *(min, max)* for general variables. Out preferred, or recommended rescaling however is to use ranks. We use the term *score* for the rescaled variables. Orientation of the data set is critical in this step. Usually, rescaling should be by variable, not by case. Depending on the orientation, this can lead for example to ranks by row or by column. We allow global scaling as an additional option for those situations where all data are already on a common scale. Following Bertin, our implementation default is to expect variables in rows.

In a second step, the scores are translated to visualization attributes. Colour is handled in two steps. The scores are translated to a colour index, which is used together with a colour palette to determine the display colour for a data element. This allows rapid experiments with various colour palettes, as long as the length of the palette is compatible. We strongly recommend to always look at the inverted table

together with a chosen one to mitigate the effects of colour perception. Simple image displays limit the visualization attributes to colour. `rect` for example allows rectangle geometry, colour, and border width. Shading and shading should be considered as an alternative for print media.

Visualization attributes may reflect different aspect. So for example in the classical Bertin display, height of a rectangle is used to reflect the value of a data element, colour is used to show an indicator whether the value is above or below variable mean.

A third step controls the actual placement of the graphical elements. With a matrix layout, it is specified by possible permutations of rows and of columns. This may be related to information used in the first two steps, but should be considered an independent step. A vector or row orders and of column orders is the critical information from this step. Various seriation methods apply. The typical situation is to select scores and display attributes, and then search for optimal or good seriations. The arrangement often leads to hard optimization problems. Placing this step later allows to use information from score transformation and attributes, which may allow more efficient algorithms.

The final step is to merge these informations and render a display.

4. Permutation, Seriation, Arrangement

As Bertin has pointed out,

> *Ce point est fondamental. C'est la mobilité interne de l'image qui charactérise la graphique moderne.* [Bertin 1977, p. 5]

Once we have solved the problem, the way can often be formulated as an optimization problem. But while we are searching for a solution, experimenting is necessary. In our implementation, we separate to lines of experiment. Finding an adequate display is one branch. This amounts to building up a collection of proven models, and a certain data set can contribute by hinting at specific needs. This is repeated not so often. Stability of implementation has priority over speed. We will provide a small number of basic model implementations.

The classic Bertin display shown above is one of the examples. Following the ideas, but deviating in the details, is to use a simple gray scale image for representation. This may be the most economic variant. But it is most economic in the use of display space. See figure 5.

We will follow the classic Bertin display and an image display as main examples.

Output

```
parasp:  1 1.2 0.8333333
changing.
```
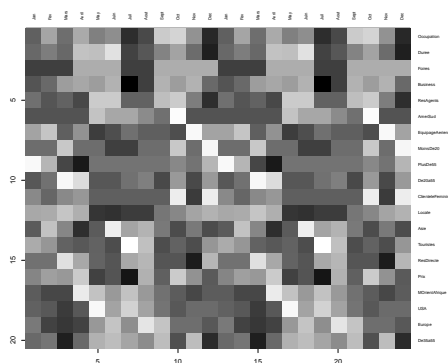

FIGURE 5. Display of a data matrix as gray scale image. Variables rearranged by similarity to occupation and duration. scale=0.5

For a chosen display, we have to compare different arrangements (seriations, for example, If we allow for interactive work, speed of display has priority. We try to cache the information that is invariant of the permutation.

As a final aspect, display space is limited. The number of variables and cases that can be displayed simultaneously is limited by the pixel size of the display. We can increase it by one or two magnitudes by using e series of detail displays. Any display calibration however should be constant for this series We try to allow for this global calibration.

The restriction to a matrix structure is arbitrary and can be omitted. Bertin has been working as a cartographer, and his main work applies to geographical data. What we call the Bertin matrices has been introduced in the very beginning of his book and are but a starting point.

5. COLOUR, PERCEPTION AND PITFALLS

still to fix

Perception is an active process, and any visual presentation may be swayed by the intricacies of perception. Colour perception is particularly complex. When working with colour (and this includes black and

white), we strongly suggest to have a look at the image with inverted
colours as well.

Here is a sample implementation. On the R level, provide a plotting
function

```
—————————————————————— Input ———————————————————————
 sampleimagem <- function(z,
          col = grey((1:256)/256), xlab, ylab, main,
          colinvert=FALSE){
          if (colinvert) col <- col[length(col):1]
          # x1, x2. y1, y2
          oldpar <- par(fig=c(0, 1, 0.2, 1),
                  mar=c(2.5,1.5,0.5,0.5), new=FALSE)
          imagem(z, col=col)
          par(yaxt="n", fig=c(0, 1, 0, 0.2),
                  mar=c(3.5,4.0,0.5,4.0), new=TRUE)
          zrange <- range(z, finite=TRUE)
          image(z=t(matrix(seq(zrange[1],zrange[2],length.out=length(col)),
                  1, length(col))),
                  zlim=zrange,main="", ylab="", xlab="", col=col)
          par(oldpar)
 }
```

and run it with `colinvert=FALSE` and `colinvert=TRUE`. If your are
using *Sweave*, use two separate chunks, and place the figure output
side by side in TEX.

```
—————————————————————— Input ———————————————————————
 hotelrk <- bertinrank(Hotel)
 sampleimagem(hotelrk)
```

```
—————————————————————— Output ——————————————————————
parasp:  1.353535 0.6 2.255892
changing.
```

See Figure 6 left.

```
—————————————————————— Input ———————————————————————
 sampleimagem(hotelrk, colinvert=TRUE)
```

```
—————————————————————— Output ——————————————————————
parasp:  1.353535 0.6 2.255892
changing.
```

See Figure 6 right.

FIGURE 6. Test matrix: same information, but colour table inverted on the right.

6. COORDINATE SYSTEM AND CONVENTIONS
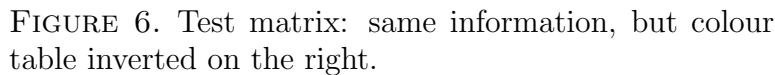
We provide prototypes for the display of Bertin matrices. To simplify the implementation of extensions, we chose a coordinate system that allots unit square to each matrix cell. If we want to add separator lines to structure our data, we have to allocate additional space `sepwd`, measured in user space.

`image` requires increasing scales for both axes. To align with matrix conventions, we have to transform the row numbers.

The total user space has the size `(nrrows * (1+2*sepwd), nrcols * (1+2*sepwd)`. The drwing area for cell `x[i, j]` is a unit square with bottom left coordinates `(nrrows-i+1)* (1+2*sepwd) - sepwd* (1+2*sepwd) - sepwd, j* (1+2*sepwd) - sepwd` .

7. TEST MATRICES

To test the implementation, a series of matrices is provided.
7.1. **Random Matrices.**

```
────────────────────────────── Input ──────────────────────────────
nrow <- 5
ncol <- 3
BMunif <- matrix(runif(nrow*ncol), nrow, ncol)
colnames(BMunif) <- colnames(BMunif, do.NULL=FALSE)
```

```
rownames(BMunif) <- rownames(BMunif, do.NULL=FALSE)
BMnorm <- matrix(rnorm(nrow*ncol), nrow, ncol)
colnames(BMnorm) <- colnames(BMnorm, do.NULL=FALSE)
rownames(BMnorm) <- rownames(BMnorm, do.NULL=FALSE)
```

7.2. **Pure Vanilla.** The most simple case: all variables are on a common scale, and the sequence is given (no seriation possible) or irrelevant (no seriation necessary).

If we want to build test matrices, there are two free parameters to be set, for example

─────────────────────── Input ───────────────────────
```
BMExplRows=8
BMExplCols=6
```

Typical cases are :

─────────────────────── Input ───────────────────────
```
BMExplUnif  <- matrix( runif(BMExplRows*BMExplCols),
      nrow= BMExplRows, ncol= BMExplCols)
BMExplNorm  <- matrix( rnorm(BMExplRows*BMExplCols),
      nrow= BMExplRows, ncol= BMExplCols)
```

─────────────────────── Input ───────────────────────
```
oldpar <- par(mfrow=c(1,2))
imagem(BMExplUnif)
```

─────────────────────── Output ───────────────────────
```
parasp:  0.3470149 0.75 0.4626866
changing.
```

─────────────────────── Input ───────────────────────
```
image.bertin(BMExplUnif)
```

─────────────────────── Output ───────────────────────
```
parasp:  0.3470149 0.75 0.4626866
changing.
```

─────────────────────── Input ───────────────────────
```
par(oldpar)
```
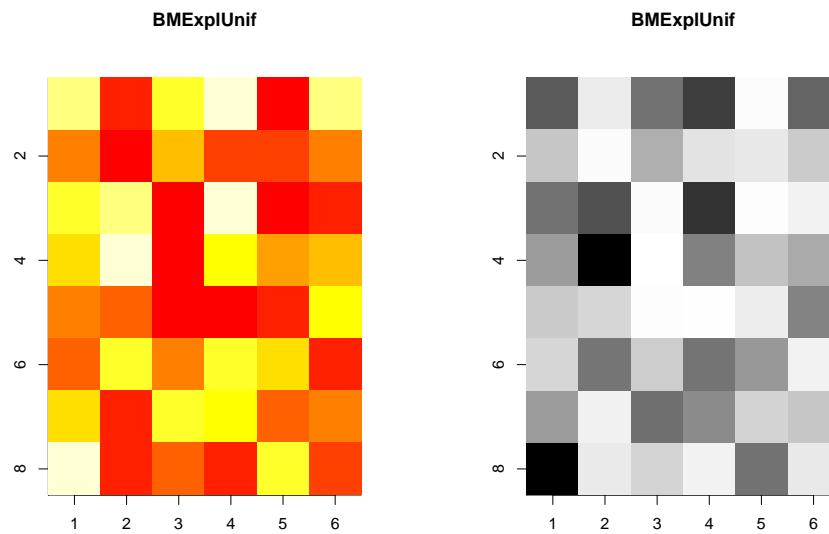
BMExplUnif   **BMExplUnif**

```
oldpar <- par(mfrow=c(1,2))
bertinrect(BMExplUnif)
```

```
parasp:  0.442446 0.75 0.5899281
changing.
```

```
plot.bertin(BMExplUnif)
```

```
parasp:  0.442446 0.75 0.5899281
changing.
```

```
par(oldpar)
```

BMExplUnif

BMExplUnif

───────────────────────────── *Input* ─────────────────────────────
```
oldpar <- par(mfrow=c(1,2))
imagem(BMExplNorm)
```

───────────────────────────── Output ─────────────────────────────
```
parasp:  0.3470149 0.75 0.4626866
changing.
```

───────────────────────────── *Input* ─────────────────────────────
```
image.bertin(BMExplNorm)
```

───────────────────────────── Output ─────────────────────────────
```
parasp:  0.3470149 0.75 0.4626866
changing.
```

───────────────────────────── *Input* ─────────────────────────────
```
par(oldpar)
```

BMExplNorm

BMExplNorm

─────────────────────────── *Input* ───────────────────────────

```
oldpar <- par(mfrow=c(1,2))
bertinrect(BMExplNorm)
```

─────────────────────────── Output ───────────────────────────

```
parasp:  0.442446 0.75 0.5899281
changing.
```

─────────────────────────── *Input* ───────────────────────────

```
plot.bertin(BMExplNorm)
```

─────────────────────────── Output ───────────────────────────

```
parasp:  0.442446 0.75 0.5899281
changing.
```

─────────────────────────── *Input* ───────────────────────────

```
par(oldpar)
```
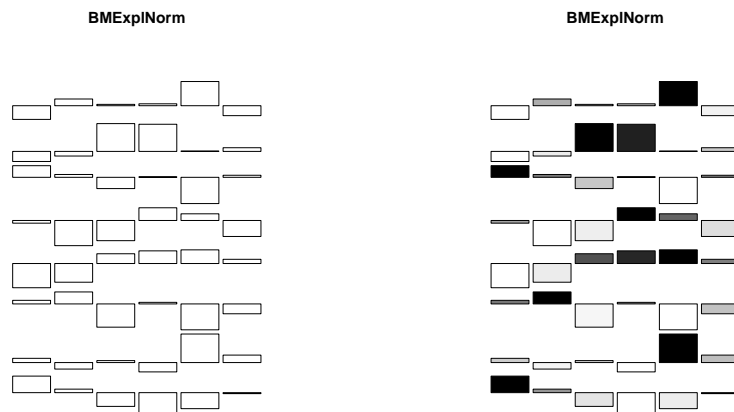

The deficits are obvious. The colours used by default appear qualitatively different, but they do not convey quantitative information. This can be easily overcome by using better colour scales. The minimum to do is to use gray scales, but of course better solutions are readily available.

The other obvious problem is that image does not reflect the orientation of the matrix (image uses image-conventions with origin at the bottom left, wheres matrix numeration uses an origin at to top left). Moreover, the aspect ratio of the image does not correspond to the aspect ratio of the matrix. The only solution is to rewrite image to a variant that is adapted to matrix conventions.

─────────────────────────── *Input* ───────────────────────────

A hidden problem with image representations is that they provide more information than usually can be processed. The colours ore grey tones may reflect to many differences. The second cheap solution to present a matrix is an array of histograms:

7.3. **Vanilla.** The next round of test cases are numeric, but not on a common scale. We provide some test vectors which we can use to construct various test matrices.

─────────────────── *Input* ───────────────────
```
# Test vectors, used to build a matrix
Bzero <- rep(0, BMExplCols)
Bone <- rep(1, BMExplCols)
Bmone <- rep(-1, BMExplCols)
Binc <- (1:BMExplCols)/BMExplCols
Bdec <- (BMExplCols:1)/BMExplCols
Bstep <- c(Bmone[1:floor(BMExplCols/2)],
        Bone[(1+floor(BMExplCols/2)):BMExplCols])
Bhat <- Bone
Bhat[(floor(BMExplCols/3)+1):(BMExplCols-floor(BMExplCols/3)) ] <- 0.5
Bnazero <- rep(c(NA,0),length.out= BMExplCols)
Bnanzero <- rep(c(NaN,0),length.out= BMExplCols)
Binf <- rep(c(Inf,0,-Inf),length.out= BMExplCols)
```

─────────────────── *Input* ───────────────────
```
# Basic test matrices
Brmatrix <- rbind(Bzero, Bone, Bmone, Binc, Bdec, Bstep, Bhat)
colnames(Brmatrix) <- colnames(Brmatrix,FALSE)
```

─────────────────── *Input* ───────────────────
```
## R may use internal housekeeping
## to keep matrix columns homogeneous.
## Check!
## Use row matrix and column matrix for tests.
Bcmatrix <- cbind(Bzero, Bone, Bmone, Binc, Bdec, Bstep, Bhat)
rownames(Bcmatrix) <- rownames(Bcmatrix,FALSE)
# Basic test matrices with random error
BrRndmatrix <- Brmatrix+rnorm(nrow(Brmatrix)*ncol(Brmatrix))
```

─────────────────── *Input* ───────────────────

```
# Test matrices with IEEE specials
Brmatrixx <- rbind(Bzero, Bone, Bmone, Binc, Bdec, Bstep, Bhat,
        Bnazero, Bnanzero, Binf)
Bcmatrixx <- cbind(Bzero, Bone, Bmone, Binc, Bdec, Bstep, Bhat,
        Bnazero, Bnanzero,Binf)
BrRndmatrixx <- Brmatrixx+rnorm(nrow(Brmatrixx)*ncol(Brmatrixx))
```

	1	2	3	4	5	6
Bzero	0.00	0.00	0.00	0.00	0.00	0.00
Bone	1.00	1.00	1.00	1.00	1.00	1.00
Bmone	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00
Binc	0.17	0.33	0.50	0.67	0.83	1.00
Bdec	1.00	0.83	0.67	0.50	0.33	0.17
Bstep	-1.00	-1.00	-1.00	1.00	1.00	1.00
Bhat	1.00	1.00	0.50	0.50	1.00	1.00
Bnazero		0.00		0.00		0.00
Bnanzero		0.00		0.00		0.00
Binf	Inf	0.00	-Inf	Inf	0.00	-Inf

TABLE 2. Brmatrixx: matrix with special values, by row

References

Bertin, J. 1977. *La graphique et le traitement graphique de l'information*, Flammarion, Paris.

———. 1999. *Graphics and graphic information processing*, Readings in information visualization, pp. 62–65.

de Falguerolles, Antoine, Felix Friedrich, and Günther Sawitzki. 1997. *A tribute to J. Bertin's graphical data analysis*, Softstat '97 (advances in statistical software 6), pp. 11–20.

Sawitzki, Günther. 1996. *Extensible statistical software: On a voyage to oberon.*, Journal of Computational and Graphical Statistics **5**, no. 3.

```
$Id: bertinR.Rnw 32 2011-09-10 09:53:39Z gsawitzki $
$Revision: 32 $
$Date: 2011-09-10 11:53:39 +0200 (Sat, 10 Sep 2011) $
$Author: gsawitzki $
```

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