dm

Henrik Renlund

March 8, 2018

Contents

1 About

`dm` is a package for functions relating to data management.

2 `dm` and related functions

2.1 The problem we want to solve

A statistical report sometimes build an *clinical data base* (CDB) from multiple sources, variables that might need to be renamed and (if categorical) recoded (and possibly transformed), the documentation of which is *significantly boring.*

The `dm` functions is an interactive-ish way of creating an CDB which both inspects the chosen variables and "documents" the process.

2.2 The elevator pitch description

...assuming LaTeX

1. point to variables (from possibly different sources), one at the time, with `dm` (along with possible renaming, recoding and transformation). (Use `dm_find` to look for candidate variables.) This gives a summary of the variable pointed to [1], and the information is stored in a list somewhere.

2. create the CDB by `dm_create`.

3. get easy-to-print documentation of where variables came from (`dm_doc2latex`) and what recodings have been done (`dm_recode2latex`).

So, the point really is to get (3) "for free" in a way that is connected to the creation of the CDB.

2.3 The stuck-in-an-elevator description

If most variables are picked form the same source, this can be set in options.

```
opts_dm$set('default_db' = 'MyDataBase')
```

If that is done, `dm` only needs a `var` argument, the name of the var you want to add. But you can use

- `var`, name of variable in source

- `name`, optional, if you want a new name for the variable (else it is set to `var`)

- `db`, name of data frame (or similar) where `var` exists (else will look at the default location, if set)

- `recode`, a list that specifies the recoding. This is the `L` argument for the `recode` function that this package provides (see the help for that functions)

---

[1]Typically one wants to to this procedure anyway to sanity check all variables that are to be included.

- **transf** a function for transforming (this might be something like a character-to-date function like `ymd` from the lubridate package)

- **comment** if you want to keep some comment about the variable

- **label** if you want to give the variable a "label" (i.e. the value of the label attribute)

- **keep.label** if `var` already has a label in `db`, should this be kept? (only if no `label` is provided)

Then as `dm` is evaluated, information about the variable is printed (to see range, levels and such).

```
dm(var = 'gEndEr', name = 'gender', label = "Perceived Gender")
    ## is followed by information being printed
```

The information about the options is stored in a list (by default "dm_doc" in an environment "dm_envir").[2] The key is the 'name' element, so as long as that is not changed, you can rerun the function with new arguments if something went wrong

```
dm('gEndEr', 'gender', label = "Biological Gender") ## overwrites
    ## the 'gender' entry
```

Else, kill all documentation and start again

```
dm_doc(kill = TRUE, prompt = FALSE) ## or possibly kill only this
  ## entry dm:::dm_doc_set('gender', NULL)
```

The documentation can be accessed

```
myDoc <- dm_doc()
print(myDoc)   ## N.B not all information is printed
```

Once all variables are created you can either store the "documentation" (and point to it later) or go on to create the CDB with `dm_create`. Specify a set if individuals (vector of id's) and, if necessary a vector of how individuals are indentified in different data frames. If the `doc` argument is not provided it will just look in `dm_doc()`.

```
id_key = c('MyDataBase' = 'id', 'Other1' = 'ID', 'Other2' = 'idno')
CDB <- dm_create(set = MyDataBase$id, id.name = id_key)
```

Now you have an CDB and you can print `dm_doc()` to show where all variables come from. You can get all recodings from

---

[2]This is due to it begin poor for functions practice to write to objects in the global environment.

```
lapply(dm_doc(), FUN = function(x) x$recode_table)
```

There is also convience functions `dm_recode2latex` and `dm_doc2latex` which will print all tables and documentation, respectively, in LaTeX format.

2.4 Toy Example

We create some toy data

```
n <- 200
BL <- data.frame(
    id = structure(sprintf("id%d", 1:n),
                   label = "identification"),
    aalder = structure(round(rnorm(n, 50, 10)),
                       label = "Age at some time",
                       foo = 'whatever'),
    vikt = rpois(n, 50),
    gr = sample(c('A', 'A2', 'B', 'C', 'D', 'd1', 'unknown'),
                n, TRUE),
    koon = structure(sample(c('M', 'K'), n, T),
                     label = "The Sex"),
    nar = as.Date("2001-01-01") + runif(n, 0, 3650),
    stringsAsFactors = FALSE
)
BL$vikt[sample(1:n, 15)] <- NA
BL$gr[sample(1:n, 10)] <- NA
m <- .9*n
COMP <- data.frame(
    ID = structure(sample(BL$id, m),
                   label = "identification"),
    foo = rbinom(m, 1, .2),
    bar = structure(rexp(m, 1/150),
                    label = "Time passed")
)
```

There are some functions to help look for relevant variables.

```
db_info(BL) ## prints names and 'label' attributes

##   source variable           label     class
## 1     BL       id  identification character
## 2     BL   aalder Age at some time   numeric
## 3     BL     vikt                    integer
## 4     BL       gr                  character
## 5     BL     koon         The Sex character
## 6     BL      nar                       Date
```

```
dm_find(pattern = 'time') ## looks in names and labels
```

```
## dm_find found:
```

```
##   source variable              label   class
## 2     BL   aalder Age at some time numeric
## 3   COMP      bar     Time passed numeric
```

Most variables of interest are in **BL** so set this as default.

```
opts_dm$set('default_db' = 'BL')
```

Next, we add the first variable (and view the output). We've chosen a variable with a fairly complex recoding to also illustrate the use the **recode** argument (also see the help for the **recode** function that is being utilized). L is a list where each entry has a vector of levels that will aquire the name of that entry, where the order of entries will be the order of the levels.

```
## 'gr' will be recoded in a more complex way
L <- list('A' = 'A2',
          'B' = NULL, ## placeholder to get the order right
          'CD' = c('C', 'D', 'd1'),
          'Unknown' = c('unknown', NA))
## # this would also work:
## L <- list('A' = c('A', 'A2'),
##           'B' = 'B',
##           'CD' = c('C', 'D', 'd1', 'something not in data'),
##           'Unknown' = c('unknown', NA))
dm('gr', recode = L, label = 'Group')
```

```
## ------------------------------------------------------------------
## Adding data base 'BL' entry 'gr' as variable 'gr'
## A variable of class: character
## There are 10 (5 percent) missing
##        and 7 (3.7 percent) unique values
## Since there are less than 20 unique vales we tabulate them:
##
##
##        A       A2       B       C       D      d1 unknown    <NA>
##       22       31      27      28      25      26      31      10
##
## Cross-tabulating the recoding:
##
##          gr
## gr        A  B CD Unknown
##    A      22  0  0       0
```

5

```
## A2        31  0  0        0
## B          0 27  0        0
## C          0  0 28        0
## D          0  0 25        0
## d1         0  0 26        0
## unknown    0  0  0       31
## <NA>       0  0  0       10
```

Next, we add some more variables (but hide the output)

```r
dm('aalder', 'Age')
dm('nar', 'When', comment = "wtf?")
dm('foo', 'event', db = 'COMP',
   recode = list('No' = '0', 'Yes' = 1),
   label = "An event at some time")
dm('bar', 'time', db = 'COMP', transf = log)
dm('koon', 'Gender',
   recode = list('Male' = 'M', 'Female' = 'K'))
```

When we are done, we create the CDB with

```r
CDB <- dm_create(set = BL$id,
                 id.name = c('BL' = 'id', 'COMP' = 'ID'))

## Fixing variable no.1: gr
## Fixing variable no.2: Age
## Fixing variable no.3: When
## Fixing variable no.4: event
## Fixing variable no.5: time
## Fixing variable no.6: Gender

db_info(CDB) ## look at what we've created

##   source variable                 label    class
## 1    CDB       id                         factor
## 2    CDB       gr                  Group  factor
## 3    CDB      Age      Age at some time numeric
## 4    CDB     When                          Date
## 5    CDB    event An event at some time  factor
## 6    CDB     time          Time passed numeric
## 7    CDB   Gender               The Sex  factor
```

We can view, or get the information

```r
## myDoc <- dm_doc()
dm_doc() ## only prints partial information in the doc
```

6

```
##     name    var   db transf                    label comment
## 1     gr     gr   BL                            Group
## 2    Age  aalder   BL            Age at some time
## 3   When    nar   BL                                    wtf?
## 4  event    foo COMP        An event at some time
## 5   time    bar COMP     log           Time passed
## 6 Gender   koon   BL                          The Sex
```

You can store the information from the 'print' of the `dm_doc()` and the recodings with, respectively,

```
pdoc <- print(dm_doc(), print = FALSE)
rtables <- lapply(dm_doc(), FUN = function(x) x$recode_table)
```

and manipulate to output format of your choice.

If we are using LaTeX, we can get the code for this with

```
## dm_doc2latex(doc = myDoc)
dm_doc2latex(caption = "Variables and their origin.")
```

Table 1: Variables and their origin.

name	var	db	label	comment
gr	gr	**BL**	Group	
Age	aalder	**BL**	Age at some time	
When	nar	**BL**		*wtf?*
event	foo	**COMP**	An event at some time	
time	bar	**COMP**	Time passed	
Gender	koon	**BL**	The Sex	

and all recode-information with

```
## dm_recode2latex(doc = myDoc)
dm_recode2latex()
```

Table 2: Recoding of data base entry `gr` into `gr`.

old ↓ new →	A	B	CD	Unknown
A	22	0	0	0
A2	31	0	0	0
B	0	27	0	0
C	0	0	28	0
D	0	0	25	0
d1	0	0	26	0
unknown	0	0	0	31
NA	0	0	0	10

Table 3: Recoding of data base entry `foo` into `event`.

old ↓ new →	No	Yes
0	148	0
1	0	32

Table 4: Recoding of data base entry `koon` into `Gender`.

old ↓ new →	Male	Female
K	0	83
M	117	0

3 `dmf` and related functions

3.1 The problem we want to solve

A clinical data base is sometimes filtered to get an *analytical data base*, the documentation of which is, again, somewhat tedious. As with `dm`, we essentially want to do make the documentation parallell with coding so it is easy to update if the coding is updated.

3.2 Toy example continuation

Now, `dmf` plays the role of `dm`.

```
dmf(f = CDB$gr != 'Unknown', name = 'crit_knowngr',
    comment = "group must be known")

## includes 79.5 perc. (159/200 rows)

dmf(f = CDB$Age >= 20 & CDB$Age <= 80, name = "crit_age",
    comment = "ages between 20 and 80")

## includes 99 perc. (198/200 rows)

dmf(f = CDB$When >= as.Date("2002-01-01") &
        CDB$When <= as.Date("2009-12-31"),
    name = "crit_date",
    comment = "study period 2002-2009")

## includes 79 perc. (158/200 rows)
```

Ok, lets look at the documentation. As with `print.dm_doc` the print method obscures the real structure of the object somewhat.

```
dm_filter()

##        criteria incl excl seq.incl seq.excl
## 1    Population  200    0      200        0
## 2 crit_knowngr  159   41      159       41
## 3     crit_age  198    2      157        2
## 4    crit_date  158   42      119       38
```

We see how many rows each criteria includes/excludes and the inclusion/exclusion of these criteria when applied sequentially. We can change the order of the sequence, by

```
print(dm_filter(), seq = c(2,3,1))
```

```
##      criteria incl excl seq.incl seq.excl
## 1  Population  200    0      200        0
## 2    crit_age  198    2      198        2
## 3   crit_date  158   42      156       42
## 4 crit_knowngr 159   41      119       37
```

3.2.1   Experimental functions

Filter description via a list

```
dm_filter2latexlist()
```

- **crit_knowngr** 'group must be known' includes 79.5 perc. (159 rows)

- **crit_age** 'ages between 20 and 80' includes 99 perc. (198 rows)

- **crit_date** 'study period 2002-2009' includes 79 perc. (158 rows)

119 of 200 are included.
    Filter cluster description

```
dm_filter2dist(plot = TRUE)
```

```
##           crit_knowngr  crit_age
## crit_age     0.5308642
## crit_date    0.9259259 0.5432099
```
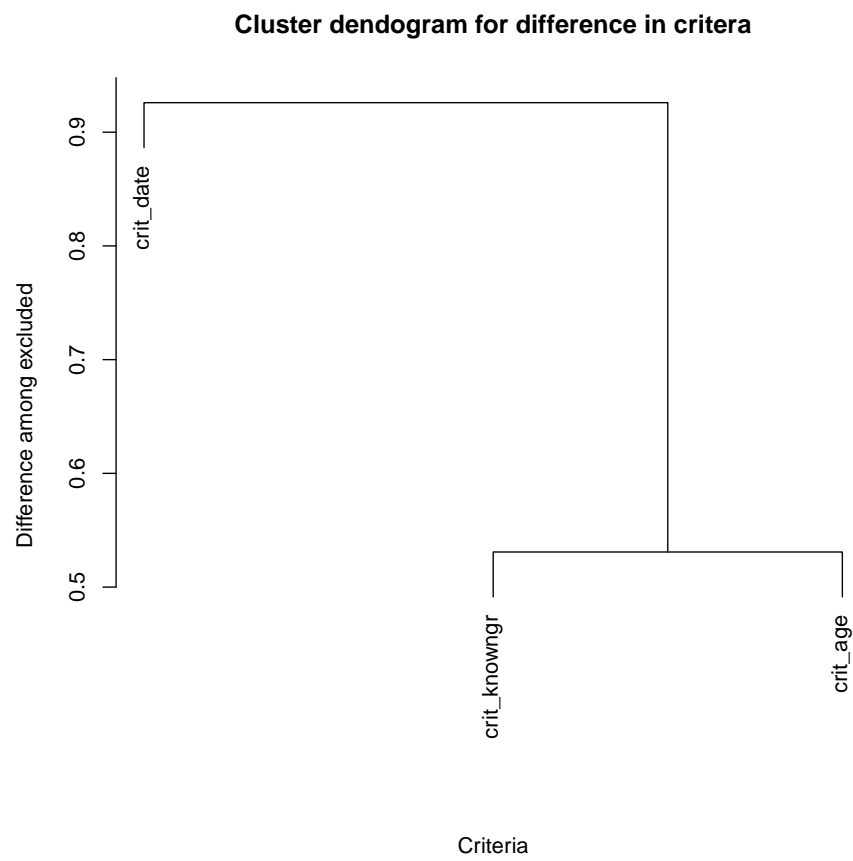
Cluster dendogram for difference in critera

Figure 1: Test of cluster description

4 grepict

(**grep** by **i**ndividual, (possibly) **c**onstrained in **t**ime.)

4.1 The problem we want to solve

Suppose you have a set of units in a data set, and another data set where each unit may occur none or serveral times, each row associated with a date and one or more variables that contains some kind of information you want to search. For each unit we want to find matches in this information, possibly within some specified time range.

The function was specifically written to deal with this situation: the units are selected to be part of some study cohort, possible with different start and end times. Another data set exists which contains the medical records of (some possibly larger) population. At least two tasks are commonly associated with creating an clinical data base:

- Find the medical history of each individual, i.e. look for codes pertaining to different diagnosis that appear before the individual is entered into the cohort.

- Find diagnosis that appear after the individual is entered into the study (and possibly before some end date).

4.2 An example

4.2.1 Generate data

Our cohort will consists of some individuals entering a study during the year 2010, with no longer than one year follow-up.

- Anna enters 2010-01-01, with no medical history or outcomes,

- Barry enter 2010-02-01, with a previous 'foo', but no outcome,

- Christina enters 2010-03-01, with no medical history but a later 'bar',

- David enters 2010-04-01, with a medical history of both 'bar' and 'foo', as well as a later 'quuz',

- Esteban enters 2010-05-01, with no medical history and an outcome 'bar' *after* the end of follow-up.

Moreover, the medical history is to be found in two different variables.

```
POP <- data.frame(
    id = c('Anna', 'Barry', 'Christina',
           'David', 'Esteban'),
    enter = as.Date(c('2010-01-01', '2010-02-01', '2010-03-01',
```

```
                         '2010-04-01', '2010-05-01'))
)
RECORDS <- data.frame(
    identity = c('Barry', 'Christina', 'David',
            'David', 'David', 'Esteban',
            'Other', 'Other'),
    what1 = c('headache', 'bar type I', 'nausea',
            'percutaneous foo', 'quuz', '',
            'other foo', 'other bar'),
    what2 = c('mild foo', 'bar type II', 'severe bar',
            'subcutaneous foo', NA, 'bar-ish',
            'yet other foo', 'yet other bar'),
    what.date = as.Date(c('2010-01-07', '2010-07-23', '1998-06-27',
                        '1996-10-12', '2011-01-18', '2011-05-03',
                        '1999-12-01', '2010-06-01'))
)[sample(1:8),]
options('knitr.kable.NA' = '')
```

The data is tabulated below

```
POP

##            id      enter
## 1       Anna 2010-01-01
## 2      Barry 2010-02-01
## 3 Christina 2010-03-01
## 4      David 2010-04-01
## 5   Esteban 2010-05-01

RECORDS

##    identity            what1            what2  what.date
## 1     Barry         headache         mild foo 2010-01-07
## 4     David percutaneous foo subcutaneous foo 1996-10-12
## 5     David            quuz             <NA> 2011-01-18
## 8     Other        other bar    yet other bar 2010-06-01
## 3     David           nausea       severe bar 1998-06-27
## 6   Esteban                            bar-ish 2011-05-03
## 2 Christina       bar type I      bar type II 2010-07-23
## 7     Other        other foo    yet other foo 1999-12-01
```

4.2.2 Medical history

Now we'll find the medical history of this cohort. We will need to point to the relevant variables in the different data sets, in RECORDS we need to point to

13

identity, `date` and `what.date`. In `POP` we need to point to `id` and specify the search interval 'begin' and 'end'. In this case, we search as far back as we can, which will happen if we set 'begin' to `NULL`. We'll search all the way until the beginning of the study (which is coded in `enter` in the data frame).

There are options for the output format, but typicall we want a *stacked* and *long* format (which will be default).

```r
searchString <- c('Foo' = 'foo', 'Bar' = 'bar', 'Quuz' = 'quuz')

tm <- grepict(
    pattern = searchString, ## what to search for
    x = c('what1', 'what2'), ## search variables in 'data'
    data = RECORDS, ## data set to search in
    id = 'identity', ## name of id variable in 'data'
    date = 'what.date', ## name of date variable in 'data'
    units = POP, ## data set, or vector, containing individuals
    units.id = 'id', ## name of id variable in 'units'
    begin = NULL, ## earliest date to search from
    end = 'enter', ## name of lates date to search,
    ## long = TRUE, ## long output format is default
    ## stack = TRUE, ## stacked results are default
    verbose = FALSE ## give calculation progress info?
)
```

`grepict` will return a data frame with many variables and, with this configuration, at least one row per individual and search string, and possibly as many as one per search string times variable searched in. Output (names are fixed):

- `id` the identifier

- `begin` first date searched from (inclusive)

- `end` last date searched untill (inclusive)

- `date` the date of the match

- `event` indicator for match

- `time` days between `begin` and `date`

- `match` that which matched

- `match.in` name of variable of match

- `pattern` pattern searched for

- `alias` name of pattern searched for

- `first.id` indicator for the first match for each individual and pattern

14

- **first.id_date** indicator for the first match for each individual, date and pattern

A few of these are tabulated below.

```r
tm[, c('id', 'event', 'alias', 'match', 'match.in', 'first.id')]
```

```
##            id event alias            match match.in first.id
## 1       Barry     1   Foo         mild foo    what2        1
## 2       David     1   Foo percutaneous foo    what1        1
## 3       David     1   Foo subcutaneous foo    what2        0
## 4   Christina     0   Foo             <NA>     <NA>        1
## 5     Esteban     0   Foo             <NA>     <NA>        1
## 6        Anna     0   Foo             <NA>     <NA>        1
## 7       David     1   Bar       severe bar    what2        1
## 8       Barry     0   Bar             <NA>     <NA>        1
## 9   Christina     0   Bar             <NA>     <NA>        1
## 10    Esteban     0   Bar             <NA>     <NA>        1
## 11       Anna     0   Bar             <NA>     <NA>        1
## 12      Barry     0  Quuz             <NA>     <NA>        1
## 13  Christina     0  Quuz             <NA>     <NA>        1
## 14      David     0  Quuz             <NA>     <NA>        1
## 15    Esteban     0  Quuz             <NA>     <NA>        1
## 16       Anna     0  Quuz             <NA>     <NA>        1
```

For the history, we typically only care whether at least one instance of each search term is found. Also, we might want to transform this to a wide format.

```r
tmp <- subset(tm, first.id == 1, select = c('id', 'event', 'alias'))
(medhist <- reshape(tmp, idvar = 'id',
                    timevar = c('alias'), direction = 'wide'))
```

```
##            id event.Foo event.Bar event.Quuz
## 1       Barry         1         0          0
## 2       David         1         1          0
## 4   Christina         0         0          0
## 5     Esteban         0         0          0
## 6        Anna         0         0          0
```

```r
names(medhist) <- gsub("event", "prior", names(medhist),
                       fixed = TRUE)
```

Now, we have a data frame containing the relevant medical history

```
medhist
```

```
##           id prior.Foo prior.Bar prior.Quuz
## 1     Barry         1         0          0
## 2     David         1         1          0
## 4 Christina         0         0          0
## 5   Esteban         0         0          0
## 6      Anna         0         0          0
```

4.2.3 Outcomes

Next, we'll look at outcomes. Since the end of study is variable, we'll have to create this variable, lets call it endofstudy. Note: if we want to start the search one day after enter we'll have to create this delayed start as a variable. Search dates are inclusive.

```r
POP$endofstudy <- POP$enter + 365
tm2 <- grepict(pattern = searchString, x = c('what1', 'what2'),
               data = RECORDS, id = 'identity',
               date = 'what.date', units = POP, units.id = 'id',
               begin = 'enter', ## earliest date to search from
               end = 'endofstudy', ## name of latest date
               verbose = FALSE)
```

For the outcomes, we probably care about more things, especially time-to-event. The event and time variables now serve as right-censored data for each outcome.

```r
tm2[, c('id', 'event', 'time', 'alias', 'match', 'match.in')]
```

```
##           id event time alias       match match.in
## 1      Barry     0  365   Foo        <NA>     <NA>
## 2  Christina     0  365   Foo        <NA>     <NA>
## 3      David     0  365   Foo        <NA>     <NA>
## 4    Esteban     0  365   Foo        <NA>     <NA>
## 5       Anna     0  365   Foo        <NA>     <NA>
## 6  Christina     1  144   Bar  bar type I    what1
## 7  Christina     1  144   Bar bar type II    what2
## 8      Barry     0  365   Bar        <NA>     <NA>
## 9      David     0  365   Bar        <NA>     <NA>
## 10   Esteban     0  365   Bar        <NA>     <NA>
## 11      Anna     0  365   Bar        <NA>     <NA>
## 12     David     1  292  Quuz        quuz    what1
## 13     Barry     0  365  Quuz        <NA>     <NA>
## 14 Christina     0  365  Quuz        <NA>     <NA>
```

16

```
## 15     Esteban     0  365  Quuz          <NA>        <NA>
## 16        Anna     0  365  Quuz          <NA>        <NA>
```

We'll assume that we only care about the first instance of each outcome.

```
tmp2 <- subset(tm2, first.id == 1,
               select = c('id', 'event', 'time', 'alias'))
(outcomes <- reshape(tmp2, idvar = 'id', timevar = c('alias'),
                     direction = 'wide'))

##             id event.Foo time.Foo event.Bar time.Bar event.Quuz time.Quuz
## 1       Barry         0      365         0      365          0       365
## 2 Christina         0      365         1      144          0       365
## 3       David         0      365         0      365          1       292
## 4     Esteban         0      365         0      365          0       365
## 5        Anna         0      365         0      365          0       365

names(outcomes) <- gsub("event", "ev", names(outcomes), fixed = TRUE)
names(outcomes) <- gsub("time", "t", names(outcomes), fixed = TRUE)
```

Now, we have a data frame containing the relevant medical history

```
outcomes

##             id ev.Foo t.Foo ev.Bar t.Bar ev.Quuz t.Quuz
## 1       Barry      0   365      0   365       0    365
## 2 Christina      0   365      1   144       0    365
## 3       David      0   365      0   365       1    292
## 4     Esteban      0   365      0   365       0    365
## 5        Anna      0   365      0   365       0    365
```

4.3 Other output

I'm thinking. . . maybe not use these?

4.3.1 Wide and stacked

With a wide, stacked output, we get one row per individual and search. We get some information on the first match - all information from the long stacked format, except `first.id` and `first.id_date` - and some summary information on all matches:

- `events` which counts the matches,

- `matches` which concatenates the matches, and

- `matches.info` which stores a concatenation of *match*:*math.in*:*date* for all matches.

```
tm3 <- grepict(pattern = searchString, x = c('what1', 'what2'),
                data = RECORDS, id = 'identity', date = 'what.date',
                units = POP, units.id = 'id', begin = 'enter',
                end = 'endofstudy',
                long = FALSE, ## wide output format
                stack = TRUE, ## stack
                verbose = FALSE
)
str(tm3)

## 'data.frame': 15 obs. of  13 variables:
##  $ id          : Factor w/ 5 levels "Anna","Barry",..: 1 2 3 4 5 1 2 3 4 5 ...
##  $ begin       : Date, format: "2010-01-01" "2010-02-01" ...
##  $ end         : Date, format: "2011-01-01" "2011-02-01" ...
##  $ date        : Date, format: "2011-01-01" "2011-02-01" ...
##  $ event       : num  0 0 0 0 0 0 0 1 0 0 ...
##  $ time        : num  365 365 365 365 365 365 365 144 365 365 ...
##  $ match       : chr  NA NA NA NA ...
##  $ match.in    : chr  NA NA NA NA ...
##  $ pattern     : chr  "foo" "foo" "foo" "foo" ...
##  $ alias       : chr  "Foo" "Foo" "Foo" "Foo" ...
##  $ events      : num  0 0 0 0 0 0 0 2 0 0 ...
##  $ matches     : chr  NA NA NA NA ...
##  $ matches.info: chr  NA NA NA NA ...
```

Also, selected info tabulated below.

```
val <- c('id', 'alias', 'event', 'time', 'events', 'matches.info')
tm3[, val]
```

```
##             id   alias event time events
## 1       Anna    Foo     0  365      0
## 2      Barry    Foo     0  365      0
## 3  Christina    Foo     0  365      0
## 4      David    Foo     0  365      0
## 5    Esteban    Foo     0  365      0
## 6       Anna    Bar     0  365      0
## 7      Barry    Bar     0  365      0
## 8  Christina    Bar     1  144      2
## 9      David    Bar     0  365      0
## 10   Esteban    Bar     0  365      0
## 11      Anna   Quuz     0  365      0
## 12     Barry   Quuz     0  365      0
## 13 Christina   Quuz     0  365      0
## 14     David   Quuz     1  292      1
## 15   Esteban   Quuz     0  365      0
##                                                matches.info
## 1                                                      <NA>
## 2                                                      <NA>
## 3                                                      <NA>
## 4                                                      <NA>
## 5                                                      <NA>
## 6                                                      <NA>
## 7                                                      <NA>
## 8  bar type I:what1:2010-07-23 bar type II:what2:2010-07-23
## 9                                                      <NA>
## 10                                                     <NA>
## 11                                                     <NA>
## 12                                                     <NA>
## 13                                                     <NA>
## 14                                      quuz:what1:2011-01-18
## 15                                                     <NA>
```

4.3.2 Wide and unstacked

With a wide and unstacked output, we get all variables (from the wide stacked output) *for each search term* - except `id`, `begin` and `end` which are the same for all rows - with the name of the search term as suffix (or a naming scheme if no names are supplied).

Below you can see the structure

```
tm4 <- grepict(pattern = searchString, x = c('what1', 'what2'),
               data = RECORDS, id = 'identity', date = 'what.date',
               units = POP, units.id = 'id', begin = 'enter',
               end = 'endofstudy',
               long = FALSE, ## wide output format
               stack = FALSE, ## don't stack
               verbose = FALSE
)
str(tm4)

## 'data.frame': 5 obs. of  33 variables:
##  $ id               : Factor w/ 5 levels "Anna","Barry",..: 1 2 3 4 5
##  $ begin            : Date, format: "2010-01-01" "2010-02-01" ...
##  $ end              : Date, format: "2011-01-01" "2011-02-01" ...
##  $ date.Foo         : Date, format: "2011-01-01" "2011-02-01" ...
##  $ event.Foo        : num  0 0 0 0 0
##  $ time.Foo         : num  365 365 365 365 365
##  $ match.Foo        : logi  NA NA NA NA NA
##  $ match.in.Foo     : logi  NA NA NA NA NA
##  $ pattern.Foo      : chr  "foo" "foo" "foo" "foo" ...
##  $ alias.Foo        : chr  "Foo" "Foo" "Foo" "Foo" ...
##  $ events.Foo       : num  0 0 0 0 0
##  $ matches.Foo      : logi  NA NA NA NA NA
##  $ matches.info.Foo : logi  NA NA NA NA NA
##  $ date.Bar         : Date, format: "2011-01-01" "2011-02-01" ...
##  $ event.Bar        : num  0 0 1 0 0
##  $ time.Bar         : num  365 365 144 365 365
##  $ match.Bar        : Factor w/ 15 levels "","bar type I",..: NA NA 2 NA NA
##  $ match.in.Bar     : Factor w/ 2 levels "what1","what2": NA NA 1 NA NA
##  $ pattern.Bar      : chr  "bar" "bar" "bar" "bar" ...
##  $ alias.Bar        : chr  "Bar" "Bar" "Bar" "Bar" ...
##  $ events.Bar       : num  0 0 2 0 0
##  $ matches.Bar      : chr  NA NA "bar type I bar type II" NA ...
##  $ matches.info.Bar : chr  NA NA "bar type I:what1:2010-07-23 bar type II:what2:2010-07-2
##  $ date.Quuz        : Date, format: "2011-01-01" "2011-02-01" ...
##  $ event.Quuz       : num  0 0 0 1 0
##  $ time.Quuz        : num  365 365 365 292 365
##  $ match.Quuz       : Factor w/ 8 levels "","bar type I",..: NA NA NA 8 NA
```

20

```
##  $ match.in.Quuz    : Factor w/ 2 levels "what1","what2": NA NA NA 1 NA
##  $ pattern.Quuz     : chr  "quuz" "quuz" "quuz" "quuz" ...
##  $ alias.Quuz       : chr  "Quuz" "Quuz" "Quuz" "Quuz" ...
##  $ events.Quuz      : num  0 0 0 1 0
##  $ matches.Quuz     : chr  NA NA NA "quuz" ...
##  $ matches.info.Quuz: chr  NA NA NA "quuz:what1:2011-01-18" ...
```

Also, selected info on the Bar- and Quuz outcome tabulated.

```
val <- c('id', names(tm4)[grepl("event|time", names(tm4))])
tm4[, val[!grepl("Foo", val)]]
```

```
##           id event.Bar time.Bar events.Bar event.Quuz time.Quuz events.Quuz
## 1      Anna         0      365          0          0       365           0
## 2     Barry         0      365          0          0       365           0
## 3 Christina         1      144          2          0       365           0
## 4     David         0      365          0          1       292           1
## 5    Esteban         0      365          0          0       365           0
```

5   Other functions

5.1   `cdate`

`cdate` is a function to handle dates of the form "20010700" or "20010000", which can appear as dates of death when the precise date is unknown. If nothing else is known this function will replace an unknown

- day of the month with the midpoint of that month, and

- month (and day) with the midpoint of that year.

Sometimes, there is another date for an individual when the individual was known to be alive. In the applications of most interest to the author it is an admission, or discharge, date to a hospital, and thus we believe that the individual did *not* die at the hospital (else the date of death would be known), therefore the replacement date will be the midpoint of whatever remains of the unknown period, e.g if we encounter date of death as "20010100" with a known hospital discharge at 2001-01-21, we will interpret it as 2001-01-26.

Examples:

```
cdate(x = c("2001/01/01", "2001/01/00", "2001/00/00"), sep = "/")

## [1] "2001-01-01" "2001-01-16" "2001-07-02"

cdate(x = c("20010101", "20010100", "20010000"),
      low.bound = as.Date(c("1999-01-01", "2001-01-21",
                            "2001-06-20")))

## [1] "2001-01-01" "2001-01-26" "2001-09-25"
```