Introduction to the MethComp package

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2 1 Data structures

The purpose of the MethComp package is to provide computational tools to manipulate, display and analyze data from method comparison studies. A method comparison study is a study where two methods of quantitative measurement are compared by measuring the same set of items with both methods.

There may be more than two methods, and there may be replicate measurements on each item by each method.

1 Data structures

In general we are concerned with measurements by different methods, on different items (persons, samples), possibly replicated.

Often such data are represented by a row of measurements for each item, with possible replicates listed either below or beside each other. This implicitly assumes that some replicate measurements belong together, which is not necessarily the case in all situations.

All functions in MethComp assume data to be represented in the "long" form, with one measurement on each row, and columns to indicate method, item and replicate. Specifically, we assume the following columns are available in a data frame:

- meth The measurement method. Numeric or factor.
- item Identification of item (person, sample). Numeric or factor.
- repl Replicate number. Numeric or factor.
- y The measurement by method meth on item item, replicate number repl.

There is a class, "Meth" for this kind of data frame. A dataframe is converted to a Meth object by using the Meth function on it:

```
> data( ox )
> str( ox )
'data.frame':
                     354 obs. of 4 variables:
 $ meth: Factor w/ 2 levels "CO","pulse": 1 1 1 1 1 1 1 1 1 1 ...
 $ item: num
              1 1 1 2 2 2 3 3 3 4 ...
              1 2 3 1 2 3 1 2 3 1 ...
 $ repl: num
      : num
              78 76.4 77.2 68.7 67.6 68.3 82.9 80.1 80.7 62.3 ...
> ox <- Meth( ox )
The following variables from the dataframe
"ox" are used as the Meth variables:
meth: meth
item: item
repl: repl
   y: y
```

#Replicates

```
2
                  3 #Items #Obs: 354 Values: min med max
Method
  CO
          1
                 56
                         61
                                  177
                                               22.2 78.6 93.5
  pulse
          1
              4
                 56
                         61
                                  177
                                               24.0 75.0 94.0
> summary( ox )
        #Replicates
              2
                  3 #Items #Obs: 354 Values: min med max
Method
  CO
              4
                 56
                                  177
                                               22.2 78.6 93.5
          1
                         61
                                  177
                                               24.0 75.0 94.0
          1
              4
                 56
                         61
  pulse
If these variable are not availabe in the data frame we may create them on the fly or by
giving the variable positions as arguments to the Meth function:
> data( fat )
> str( fat )
'data.frame':
                     258 obs. of 5 variables:
 $ Id: num 1 1 1 3 3 3 5 5 5 11 ...
 $ Obs: Factor w/ 2 levels "KL", "SL": 1 1 1 1 1 1 1 1 1 1 ...
 $ Rep: num 1 2 3 1 2 3 1 2 3 1 ...
 $ Sub: num 1.6 1.7 1.7 2.8 2.9 2.8 2.7 2.8 2.9 3.9 ...
 $ Vic: num 4.5 4.4 4.7 6.4 6.2 6.5 3.6 3.9 4 4.3 ...
> sc <- Meth( fat, 2, 1, 3, 4 )
The following variables from the dataframe
"fat" are used as the Meth variables:
meth: Obs
item: Id
repl: Rep
   y: Sub
       #Replicates
Method
                3 #Items #Obs: 258 Values: min med max
                                             0.39 1.7 4.2
    KL
               43
                       43
                                129
                                             0.51 1.7 4.1
    SL
               43
                       43
                                129
> str(sc)
Classes 'Meth' and 'data.frame':
                                         258 obs. of 5 variables:
 $ meth: Factor w/ 2 levels "KL", "SL": 1 1 1 1 1 1 1 1 1 1 ...
 $ item: Factor w/ 43 levels "1", "2", "3", "4", ...: 1 1 1 3 3 3 5 5 5 11 ...
 $ repl: Factor w/ 3 levels "1","2","3": 1 2 3 1 2 3 1 2 3 1 ...
     : num 1.6 1.7 1.7 2.8 2.9 2.8 2.7 2.8 2.9 3.9 ...
 $ Vic : num 4.5 4.4 4.7 6.4 6.2 6.5 3.6 3.9 4 4.3 ...
```

4 1 Data structures

```
> summary( sc )
```

```
#Replicates
```

```
Method 3 #Items #Obs: 258 Values: min med max KL 43 43 129 0.39 1.7 4.2 SL 43 43 129 0.51 1.7 4.1
```

We may even give some of them as names of the columns in the dataframe:

```
> vi <- Meth( fat, 2,1, "Rep", "Vic" )
```

```
The following variables from the dataframe
```

"fat" are used as the Meth variables:

meth: Obs
item: Id
repl: Rep
 y: Vic
 #Replicates

Method 3 #Items #Obs: 258 Values: min med max KL 43 43 129 2.0 3.9 6.5 SL 43 43 129 2.3 4.1 6.7

However, more complicated operations on the dataframe is best done on the fly using the with function (from the base package):

```
> data( hba1c )
> str( hba1c )
'data.frame':
                     835 obs. of 6 variables:
         : Factor w/ 3 levels "BR.V2", "BR.VC", ...: 2 2 2 2 2 2 2 1 1 ...
        : Factor w/ 2 levels "Cap", "Ven": 2 2 2 2 1 1 1 1 2 2 ...
 $ type
 $ item : num 12 12 12 12 12 12 12 12 12 12 ...
 $ d.samp: num 1 1 1 1 1 1 1 1 1 1 ...
 $ d.ana : num 2 3 4 5 2 3 4 5 2 3 ...
         : num 8.7 8.7 8.7 8.7 9.2 9 8.8 8.7 9.4 9.3 ...
 $ у
> hb1
      <- with( hba1c,
                Meth( meth = interaction(dev, type),
                      item = item,
```

#Replicates

Method	3	4	#Items	#0bs: 83	Values:	min med	max
BR.V2.Cap	0	38	38	153	2	5.3 8.0	12.6
BR.VC.Cap	19	19	38	133	3	5.3 8.2	12.1
${\tt Tosoh.Cap}$	0	38	38	15:	2	5.0 7.8	11.8
BR.V2.Ven	19	19	38	133	3	5.5 8.1	12.0
BR.VC.Ven	19	19	38	133	3	5.3 8.0	11.6
Tosoh.Ven	20	18	38	13:	2	5.3 8.0	12.1

repl = d.ana-d.samp,

y = y, print=TRUE))

```
> str( hb1 )
Classes 'Meth' and 'data.frame': 835 obs. of 4 variables:
$ meth: Factor w/ 6 levels "BR.V2.Cap","BR.VC.Cap",..: 5 5 5 5 2 2 2 2 4 4 ...
$ item: Factor w/ 38 levels "1","2","3","4",..: 12 12 12 12 12 12 12 12 12 12 ...
$ repl: Factor w/ 5 levels "0","1","2","3",..: 2 3 4 5 2 3 4 5 2 3 ...
$ y : num 8.7 8.7 8.7 8.7 9.2 9 8.8 8.7 9.4 9.3 ...
```

Objects of class Meth (which inherits from data.frame) has methods such as summary, plot, subset and transform. The functions mostly do not require the data to be in Meth format — if a dataframe with the right columns is supplied, it is normally converted internally to Meth format.

2 Function overview

The following is a brief overview of the functions in the MethComp package. The full documentation is in the help pages for the functions, and an illustration of the way they work can be obtained by referring to the printed manual at the end of this document or on the fly by typing e.g.:

```
> ?plot.Meth
```

which will bring up the manual page for the function plot.meth. The example code from the manual page can be run directly by:

```
> example( plot.Meth )
```

2.1 Graphical functions

BA.plot Makes a Bland-Altman plot of two methods from a data frame with method comparison data, and computes limits of agreement. The plotting is really done by a call to the function BlandAltman.

BlandAltman draws a Bland-Altman plot and computes limits of agreement.

plot.Meth Plots all methods against all others, both as a scatter plot and as a Bland-Altman plot.

bothlines Adds regression lines of y on x and vice versa to a scatter plot. Optionally, the Deming regression line can be added too.

2.2 Data manipulating functions

make.repl Generates (or replaces) a repl column in a data frame with columns meth, item and y.

perm.repl Randomly permutes replicates within (method,item) and assigns new replicate numbers.

- to.wide Transforms a data frame in the long form to the wide form where separate columns for each method are generated, with one row per (item,replicate).
- to.long Reverses the result of to.wide. The function can also generate a long form dataset from a dataset with different methods beside each other.
- summary. Meth Tabulates items by method and no. replicates for a Meth object.
- Meth.sim Simulates a dataset from a method comparison experiment for given parameters for bias, exchangeability and variance component sizes.

2.3 Analysis functions

- BA.est Estimates in the variance components models underlying the concept of limits of agreement, and returns the bias and the variance components. Assumes constant bias between methods.
- Deming Performs Deming regression, i.e. regression with errors in both variables.
- DA.reg Regresses the differences between methods on the averages and derives approximate linear conversion equations, based on [?].
- AltReg Estimates via alternating regressions in the general model. Returns estimates of mean conversion parameters and variance components.
- MCmcmc Estimates via BUGS in the general model with non-constant bias (and in the future) possibly non-constant standard deviations of the variance components. Produces a MCmcmc object, which is an mcmc.list object with some extra attributes. mcmc.list objects are handeled by the coda package, so this is required when calling MCmcmc.

2.4 Reporting functions

Some of these functions all take a MCmcmc object as input, others will postprocess the output of DA.reg, BA.est or AltReg.

The functions BA.est, AltReg return objects that have class MethComp, whereas the result of MCmcmc can be converted to an object of this type by the MethComp function. The reason for this is that the results of the MCmcmc function is output from an MCMC-simulation which we may want to monitor by special functions. The MethComp function only takes the central summaries from the MCmcmc object assuming the chains have reached convergence.

- print.MethComp Prints a table of conversion equation between methods analyzed, with prediction standard deviations.
- print.MCmcmc Prints a table of conversion equation between methods analyzed, with prediction standard deviations, but also gives summaries of the posteriors for the parameters that constitute the conversion algorithms.

- plot.MethComp, plot.MCmcmc Plots the conversion lines between methods with prediction limits.
- post.MCmcmc Plots smoothed posterior densities for the estimates. Primarily of interest for the variance components, but it has aruments to produce the posterior of the intercepts and the slopes of the conversion lines between methods too.
- check.MCmcmc Makes diagnistic plots of the traces of the chains included in the MCmcmc object.