# hyperSpec Introduction

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#### Reproducing the Examples in this Vignette

All spectra used in this manual are installed automatically with hyperSpec. Note that some definitions are executed in vignette.defs.

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#### 1. Introduction

inline:

hyperSpec is a R package that allows convenient handling of hyperspectral data sets, i. e. data sets combining spectra with further data on a per-spectrum basis. The spectra can be anything that is recorded over a common discretized axis.

This vignette gives an introduction on basic working techniques using the R package *hyperSpec*. This is done mostly from a spectroscopic point of view: rather than going through the functions provided by *hyperSpec*, it is organized in spectroscopic tasks. However, the functions discussed are printed on the margin for a quick overview.

hyperSpec comes with five data sets,

available

chondro a Raman map of chondrocytes in cartilage,

flu a set of fluorescence spectra of a calibration series, and

laser a time series of an unstable laser emission

paracetamol a Raman spectrum of paracetamol (acetaminophene) ranging from 100 to 3200 cm<sup>-1</sup> with overlapping wavelength ranges.

barbiturates GC-MS spectra with differing wavelength axes as a list of 286 hyperSpec objects.

In this vignette, the data sets are used to illustrate appropriate procedures for different tasks and different spectra. In addition, the first three data sets are accompanied by their own vignettes showing exemplary work flows for the respective data type.

This document describes how to accomplish typical tasks in the analysis of spectra. It does not give a complete reference on particular functions. It is therefore recommended to look up the methods in R's help system using ? command.

A complete list of the functions available in hyperSpec is given in appendix A (p. 37).

#### 1.1. Notation and Terms

Throughout the documentation of the package, the following terms are used:

wavelength: spectral abscissa

frequency, wavenumbers, chemical shift, Raman shift,  $\frac{m}{z}$ , etc.

intensity: spectral ordinate

transmission, absorbance,  $\frac{e^-}{s}$ , intensity, ...

extra data: further information/data belonging to each spectrum

spatial information (spectral images, maps, or profiles), temporal information (kinetics, time series), concentrations (calibration series), class membership information, etc.

hyperSpec object may contain arbitrary numbers of extra data columns.

In R, slots of a S4 class are accessed by the @ operator. In this vignette, the notation @xxx will thus mean "slot xxx of an object". Likewise, named elements of a list and columns of a data.frame are accessed by the \$ operator, and \$xxx will be used for "column xxx", and as an abbreviation for "column xxx of the data.frame in slot data of the object" (the structure of hyperSpec objects is discussed in section 4, p. 5).

#### 2. Remarks on R

#### 2.1. Generic Functions

Generic Functions are functions that apply to a wide range of data types or classes, e. g. plot, print, mathematical operators, etc. These functions can be implemented in a specialized way by each class. hyperSpec implements with a variety of such functions, see appendix A (p. 37).

#### 2.2. Functionality Can be Extended at Runtime

R's concept of functions offers much flexibility. Functions may be added or changed by the user in his *workspace* at any time. This is also true for methods belonging to a certain class. Neither restart of R nor reloading of the package or anything the like is needed. If the original function resides in a namespace (as it is the case for all functions in *hyperSpec*), the original function is not deleted. It is just masked by the user's new function but stays accessible via the :: operator.

The same is true for "normal" variables: You may create changed copies of the example data sets, work with these and then "reset" to *hyperSpec*'s version of the data set by removing the object in your workspace.

This offers the opportunity of easily writing specialized functions that are adapted to specific tasks. hyperSpec's vignettes use this to set up special versions of the lattice graphics functions that are already wrapped in print (see also R FAQ: Why do lattice/trellis graphics not work?) and allow the code in the code chunks of the vignettes to be exactly what one would type during an interactive R session. For the code, check the vignettes.defs file accompanying all hyperSpec vignettes.

#### 2.3. Validity Checking

S4 classes have a mechanism to define and enforce that the data actually stored in the object is appropriate for this class. In other words, there is a mechanism of *validity checking*.

The functions provided by *hyperSpec* check the validity of *hyperSpec* objects at the beginning, and – if the validity could be broken by inappropriate arguments – also before leaving the function.

validObject, chk.hy

It is highly recommended to use validity checking also for user-defined functions. In addition, nongeneric functions should first ensure that the argument actually is a *hyperSpec* object. The two tasks are accomplished by:

```
> chk.hy (object)
> validObject (object)
```

The first line checks whether object is a *hyperSpec* object, the second checks its validity. Both functions return TRUE if the checks succeed, otherwise they raise an error and stop.

#### 2.4. Special Function Names

#### 2.4.1. The Names of Operators

Operators such as +, -,\*, %%, etc. are in fact functions in R. Thus they can be handed over as arguments to other functions (particularly to the vectorization functions \*apply, sweep, etc.). In this case the name of the function must be quoted: `-` is the recommended style (although "-" will often work as well), e.g.:

```
> sweep (flu, 2, mean, `-`)
These functions can also be called in a more function-like style (prefix notation):
> `+` (3, 5)
[1] 8
```

#### 2.4.2. Assignment Functions

R allows the definition of functions that do an assignment (set some part of the object), such as:

```
> wl (flu) <- new.wavelength.values
an assignment to variable wl: `wl<-`.</pre>
```

#### 3. Loading and the package and configuration

```
To load hyperSpec, use > library ("hyperSpec")
```

The global behaviour of *hyperSpec* can be configured via options. The values of the options are retrieved with hy.getOptions and hy.getOption, and changed with hy.setOptions. Table 1 gives an overview.

#### 4. The structure of hyperSpec objects

hyperSpec is a S4 (or new-style) class. Four slots contain the parts of the object:

Cwavelength containing a numeric vector with the wavelength axis of the spectra.

@data a data.frame with the spectra and all further information belonging to the spectra

@label a list with appropriate labels (particularly for axis annotations)

name	default value (range)	description	used by
debuglevel	0 (1L 2L)	amount of debugging information produced	spc.identify, map.identify, various file import functions
gc	FALSE	triggers frequent calling of gc ()	read.ENVI, new ("hyperSpec")
tolerance	$\sqrt{ .   exttt{Machine\$.dou}}$	bltelepsnce for numerical comparisons	file import func- tions (removing empty spectra), normalize01
file.remove.empRyspRc		automatic removing of empty spectra	<pre>file import func- tions, see vignette ("fileio")</pre>
file.keep.nar	meTRUE	automatic recording of file name in column \$filename	file import functions, see vignette ("fileio")

 $\textbf{Table 1} \quad \textit{hyperSpec} \text{ options. Please refer to the documentation of the respective functions for details about the effect of the options.}$ 

slot	get	set
@wavelength	wl	wl<-
@data	[, [[, $\$$ , as.data.frame, as.long.df,	[<-, [[<-, \$<-
@label	labels	labels<-

Table 2 Get and set functions for the slots of hyperSpec objects

While the parts of the *hyperSpec* object can be accessed directly, it is good practice to use the functions provided by *hyperSpec* to handle the objects rather than accessing the slots directly (tab. 2). This also ensures that proper (*valid*) objects are returned. In some cases, however, direct access to the slots can considerably speed up calculations, see section 14 (p. 34).

Most of the data is stored in @data. This data.frame has one special column, \$spc. It is the column that actually contains the spectra. The spectra are stored in a matrix inside this column, as illustrated in figure 1. Even if there are no spectra, \$spc must still be present. It is then a matrix with zero columns.

Slot @label contains an element for each of the columns in @data plus one holding the label for the wavelength axis, .wavelength. They are accessed by their names which must be the same for columns of @data and the list elements. The elements of the list may be anything suitable for axis annotations, i.e. they should be either character strings or expressions for "pretty" axis annotations (see e.g. figure 7 on page 26). To get familiar with expressions for axis annotation, see ? plotmath and demo (plotmath).

#### 5. Functions provided by hyperSpec

Table A (p. 37) in the appendix gives an overview of the functions implemented by hyperSpec.

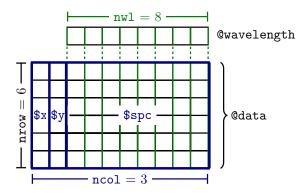


Figure 1 The structure of the data in a hyperSpec object.

#### 6. Obtaining Basic Information about hyperSpec Objects

As usual, the *print* and *show* methods display information about the object, and *summary* yields some additional details about the data handling done so far:

print, show, summary

> chondro

```
hyperSpec object
   875 spectra
   4 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (875 rows x 4 columns)
   1. y: y [numeric] -4.77 -4.77
   2. x: x [numeric] -11.55 -10.55 ... 22.45
   3. clusters: clusters [factor] matrix matrix ... lacuna + NA
   4. spc: I / a.u. [matrix300] 501.82 500.46 ... 169.29
> summary (chondro)
hyperSpec object
   875 spectra
   4 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (875 rows x 4 columns)
   1. y: y [numeric] -4.77 -4.77 ... 19.23
   2. x: x [numeric] -11.55 -10.55 ... 22.45
   3. clusters: clusters [factor] matrix matrix ... lacuna + NA
   4. spc: I / a.u. [matrix300] 501.82 500.46 ... 169.29
```

The data set chondro consists of 875 spectra with 300 data points each, and 4 data columns: two for the spatial information, one factor with the results of a cluster analysis plus \$spc. These information nwl, dim can be directly obtained by

```
> nrow (chondro)
[1] 875
> nwl (chondro)
[1] 300
> ncol (chondro)
[1] 4
```

```
> dim (chondro)
nrow ncol nwl
875    4 300
The names of the columns in @data are accessed by
> colnames (chondro)
```

11 x 11

[1] "y"

colnames, rownames, dimnames, wl

Likewise, rownames returns the names assigned to the spectra, and dimnames yields a list of these three vectors (including also the column names of \$spc). The column names of the spectra matrix contain the wavelengths as character, while wl (see section 9.5.4, p. 15) yields the numeric vector of wavelengths.

Extra data column names and rownames of the object may be set by colnames<- and rownames<-, respectively. colnames<- renames the labels as well.

colnames<-,
rownames<-</pre>

#### 7. Creating a hyperSpec Object, Data Import and Export

"clusters" "spc"

hyperSpec comes with filters for a variety of file formats. These are discussed in detail in a separate vignette accessible via vignette ("fileio").

#### 7.1. Creating a hyperSpec Object from Spectra Matrix and Wavelength Vector

If the data is in R's workspace, a hyperSpec object is created by:

```
> spc <- new ("hyperSpec", spc = spectra.matrix, wavelength = wavelength.vector, data = extra.data)
```

The most frequently needed arguments are:

spc the spectra matrix

wavelength the wavelength axis vector

data the extra data (can already contain the spectra matrix in column \$spc)

a list with the proper labels. Do not forget the wavelength axis label in \$.wavelength and the spectral intensity axis label in \$spc.

More information about converting existing data into hyperSpec objects can be found in vignette ("fileio").

#### 7.2. Creating Random Spectra

If *mvtnorm* is available, multivariate normally distributed spectra can be generated from mean and covariance matrix using rmmvnorm (fig. 2a). Note that the *hyperSpec* function's name has an additional "m": it already takes care of multiple groups. Mean spectra and pooled covariance matrix can be calculated using pooled.cov:

rmmvnorm

pooled.cov

```
> pcov <- pooled.cov (chondro, chondro$clusters)
> rnd <- rmmvnorm (rep (10, 3), mean = pcov$mean, sigma = pcov$COV)
> cluster.cols <- c ("dark blue", "orange", "#CO2020")
> plot (rnd, col = cluster.cols [rnd$.group])
```

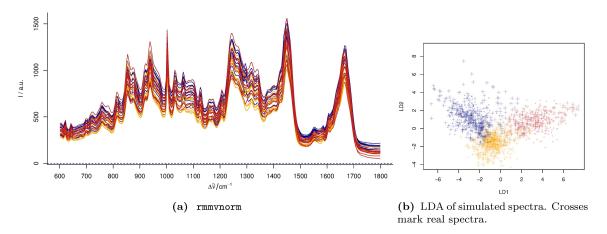


Figure 2 Multivariate normally distributed random spectra.

fig. 2b shows the linear discriminant analysis (LDA) scores of such simulated specta in comparison to the real spectra in the chondro object:

If individual covariance matrices should be used for each group, *sigma* should be an array with the 3rd dimension corresponding to the group.

#### 8. The Logbook

#### **Deprecated**

The logbook has now been REMOVED.

#### 9. Access to the data

The main functions to retrieve the data of a hyperSpec object are [] and [[]].

[], [[]]

The difference between these functions is that [] returns a *hyperSpec* object, whereas the result of [[]] is a *data.frame* if extra data columns were selected or otherwise the spectra matrix. Single extra data columns may be retrieved by \$.

In order to change data, use  $[] \leftarrow$ ,  $[[]] \leftarrow$ , and  $\leftarrow$  (see 9.4 and 9.3).

[]<-, [[]]<-,

#### 9.1. Access Functions and Abbreviations for Parts of the hyperSpec Object's Data

hyperSpec comes with three abbreviation functions for easy access to the data:

[] [[]] \$. \$.. []<-[[]]<- \$<- x [[]] returns the spectra matrix (x\$spc).

x = [i, j, l] the cut spectra matrix is returned if wavelengths are specified in l.

x = [i, j, l] If data columns are selected (second index), the result is a data-frame.

x = [i, , l] <- Also, parts of the spectra matrix can be set (only indices for spectra and wavelength are allowed for this function).

 $x[i, j] \leftarrow sets parts of x@data.$ 

x \$. returns the complete data frame x@data, with the spectra in column \$spc.

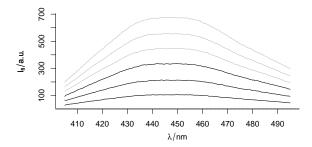
x \$.. returns the extra data (x@data without x\$spc).

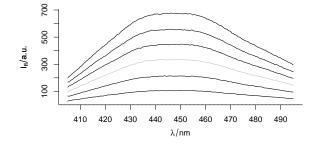
 $x \$ .. <- sets the extra data (x@data without x\$spc). The columns must match exactly in this case.

#### 9.2. Selecting and Deleting Spectra

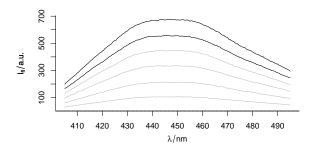
The extraction function [] takes the spectra as first argument (For detailed help: see? `[`). It may be a vector giving the indices of the spectra to extract (select), a vector with negative indices indicating which spectra should be deleted, or a logical. Note that a matrix given to [] will be treated as a vector.

```
> plot (flu, col = "gray")
> plot (flu [1 : 3], add = TRUE)
```





```
> plot (flu, col = "gray")
> plot (flu [flu$c > 0.2], add = TRUE)
```



#### 9.2.1. Random Samples

```
A random subset of spectra is conveniently selected by sample:
                                                                                                         sample
> sample (chondro, 3)
hyperSpec object
   3 spectra
   4 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (3 rows x 4 columns)
   1. y: y [numeric] 9.23 -0.77 16.23
   2. x: x [numeric] 1.45 3.45 1.45
   3. clusters: clusters [factor] cell
                                        matrix matrix
   4. spc: I / a.u. [matrix300] 282.44 306.41 ... 119.63
If appropriate indices into the spectra are needed instead, use isample:
                                                                                                         isample
```

seq

```
> isample (chondro, 3)
```

[1] 719 170 118

#### 9.2.2. Sequences

```
Sequences of every n<sup>th</sup> spectrum or the like can be retrieved with seq:
```

```
> seq (chondro, length.out = 3, index = TRUE)
```

```
[1] 1 438 875
```

```
> seq (chondro, by = 100)
```

```
hyperSpec object
   9 spectra
   4 data columns
```

300 data points / spectrum wavelength: Delta \* tilde(nu)/cm^-1 [numeric] 602 606 ... 1798

data: (9 rows x 4 columns) 1. y: y [numeric] -4.77 -2.77 ... 17.23

2. x: x [numeric] -11.55 18.45 ... 18.45 3. clusters: clusters [factor] matrix matrix ... lacuna

4. spc: I / a.u. [matrix300] 501.82 400.94 ... 124.64

Here, indices may be requested using index = TRUE.

#### 9.3. Selecting Extra Data Columns

The second argument of the extraction functions [] and [[]] specifies the (extra) data columns. They can be given like any column specification for a *data.frame*, i. e. numeric, logical, or by a vector of the column names:

```
> colnames (chondro)
[1] "y"
              "x"
                         "clusters" "spc"
> chondro [[1 : 3, 1]]
1 -4.77
2 - 4.77
3 -4.77
> chondro [[1 : 3, -4]]
           x clusters
1 -4.77 -11.55 matrix
2 -4.77 -10.55
               matrix
3 -4.77 -9.55 matrix
> chondro [[1 : 3, "x"]]
1 -11.55
2 -10.55
3 -9.55
> chondro [[1 : 3, c (TRUE, FALSE)]]
                                          # note the recycling!
     y clusters
1 -4.77 matrix
2 -4.77
         matrix
To select one column, the $ operator is more convenient:
                                                                                                    $
> flu$c
[1] 0.05 0.10 0.15 0.20 0.25 0.30
hyperSpec supports command line completion for the $ operator.
The extra data may also be set this way:
                                                                                                    $<-
> flu$n <- list (1 : 6, label = "sample no.")
```

# 9.4. More on the [[]] and [[]] <- Operators: Accessing Single Elements of the Spectra Matrix

[[]] works mostly analogous to []. In addition, however, these two functions also accept index matrices of size  $n \times 2$ . In this case, a vector of values from the spectra matrix is returned.

This function will append new columns, if necessary.

```
> chondro [[indexmatrix, wl.index = TRUE]]
[1] 501.82 507.81 456.03
> diag (chondro [[1 : 3, , min ~ min + 2i]])
[1] 501.82 507.81 456.03
[[]] \leftarrow also accepts index matrices of size n \times 2.
> indexmatrix <- matrix (c (1 : 3, 1 : 3), ncol = 2)</pre>
> indexmatrix
    [,1] [,2]
[1,]
       1
[2,]
       2
            2
       3
[3,]
> chondro [[indexmatrix, wl.index = TRUE]]
[1] 501.82 507.81 456.03
> diag (chondro [[1 : 3, , min ~ min + 2i]])
[1] 501.82 507.81 456.03
```

#### 9.5. Wavelengths

#### 9.5.1. Converting Wavelengths to Indices and vice versa

Spectra in hyperSpec have always discretized wavelength axes, they are stored in a matrix with each column corresponding to one wavelength. hyperSpec provides two functions to convert the respective column indices into wavelengths and vice versa: i2wl and wl2i.

using a formula. The basic syntax is start  $\sim$  end. This yields a vector index of start: index of end.

If the wavelengths are given as a numeric vector, they are each converted to the corresponding wavelength. In addition there is a more sophisticated possibility of specifying wavelength ranges

wl2i i2wl

The result of the formula conversion differs from the numeric vector conversion in three ways:

- The colon operator for constructing vectors accepts only integer numbers, the tilde (for formulas) does not have this restriction.
- If the vector does not take into account the spectral resolution, one may get only every  $n^{\text{th}}$  point or repetitions of the same index:

```
> wl2i (flu, 405 : 410)
[1] 1 3 5 7 9 11
> wl2i (flu, 405 ~ 410)
[1] 1 2 3 4 5 6 7 8 9 10 11
> wl2i (chondro, 1000 : 1010)
[1] 100 101 101 101 101 102 102 102 102 103 103
> wl2i (chondro, 1000 ~ 1010)
[1] 100 101 102 103
```

• If the object's wavelength axis is not ordered, the formula approach will give weird results. In that (probably rare) case, use orderwl first to obtain an object with ordered wavelength axis.

start and end may contain the special variables min and max that correspond to the lowest and highest wavelengths of the object:

```
> wl2i (flu, min ~ 410)
[1] 1 2 3 4 5 6 7 8 9 10 11
```

Often, specifications like wavelength  $\pm n$  data points are needed. They can be given using complex numbers in the formula. The imaginary part is added to the index calculated from the wavelength in the real part:

```
> wl2i (flu, 450 - 2i ~ 450 + 2i)
[1] 89 90 91 92 93
> wl2i (flu, max - 2i ~ max)
[1] 179 180 181
```

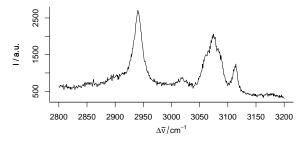
To specify several wavelength ranges, use a list containing the formulas and vectors<sup>1</sup>:

This mechanism also works for the wavelength arguments of [], [[]], and plotspc.

#### 9.5.2. Selecting Wavelength Ranges

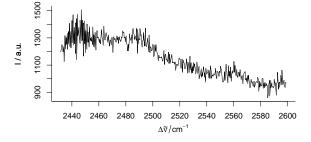
Wavelength ranges can easily be selected using []'s third argument:

```
> plot (paracetamol [,, 2800 ~ 3200])
```



By default, the values given are treated as wavelengths. If they are indices into the columns of the spectra matrix, use wl.index = TRUE:

```
> plot (paracetamol [,, 2800 : 3200, wl.index = TRUE])
```



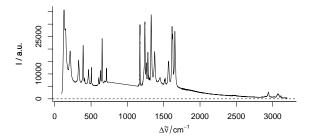
Section 9.5.1 (p. 13) details into the different possibilities of specifying wavelengths.

<sup>&</sup>lt;sup>1</sup>Formulas are combined to a list by c.

#### 9.5.3. Deleting Wavelength Ranges

Deleting wavelength ranges may be accomplished using negative index vectors together with wl.index = TRUE.

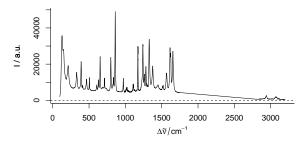
> plot (paracetamol [,, -(500 : 1000), wl.index = TRUE])



However, this mechanism works only if the proper indices are known.

If the range to be cut out is rather known in the units of the wavelength axis, it is easier to select the remainder of the spectrum instead. To delete the spectral range from 1750 to  $2800\,\mathrm{cm^{-1}}$  of the paracetamol spectrum one can thus use:

> plot (paracetamol [,, c (min ~ 1750, 2800 ~ max)])



(It is possible to produce a plot of this data where the cut range is actually omitted and the wavelength axis is optionally cut in order to save space. For details see the "plotting" vignette).

#### 9.5.4. Changing the Wavelength Axis

Sometimes wavelength axes need to be transformed, e.g. converting from wavelengths to frequencies. In this case, retrieve the wavelength axis vector with w1, convert each value of the resulting vector and assign the result with w1<-. Also the label of the wavelength axis may need to be adjusted.

wl, wl<-

As an example, convert the wavelength axis of laser to frequencies. As the wavelengths are in nanometers, and the frequencies are easiest expressed in terahertz, an additional conversion factor of 1000 is needed:

#### > laser

```
hyperSpec object
84 spectra
2 data columns
36 data points / spectrum
wavelength: lambda/nm [numeric] 404.58 404.62 ... 405.82
data: (84 rows x 2 columns)
1. t: t / s [numeric] 0 2 ... 5722
2. spc: I / a.u. [matrix36] 164.65 179.72 ... 112.09
```

```
> wavelengths <- wl (laser)
> frequencies <- 2.998e8 / wavelengths / 1000
> wl (laser) <- frequencies
> labels (laser, ".wavelength") <- "f / THz"
> laser
hyperSpec object
   84 spectra
   2 data columns
   36 data points / spectrum
wavelength: f / THz [numeric] 741.01 740.95 ... 738.76
data: (84 rows x 2 columns)
   1. t: t / s [numeric] 0 2 ... 5722
   2. spc: I / a.u. [matrix36] 164.65 179.72 ... 112.09
> rm (laser)
There are other possibilities of invoking wl<- including the new label, e.g.
> wl (laser, "f / THz") <- frequencies
and
> wl (laser) <- list (wl = frequencies, label = "f / THz")</pre>
see ?`wl<-` for more information.
```

#### 9.5.5. Ordering the Wavelength Axis

If the wavelength axis of an object needs reordering (e.g. after collapse), orderwl can be used:

```
> barb <- collapse (barbiturates [1 : 3])
> wl (barb)

[1] 160.90 158.85 147.00 140.90 133.05 130.90 119.95 119.15 118.05 116.95 112.90 106.00 105.10
[14] 98.95 96.95 91.00 85.05 83.05 77.00 71.90 71.10 70.00 69.00 57.10 56.10 55.00
[27] 43.85 43.05 41.10 40.10 39.00 32.15 31.15 30.05 29.05 28.15 27.05 132.95 131.00
[40] 120.05 119.05 117.95 113.00 105.90 82.95 72.00 69.10 56.00 44.05 40.00 30.15 28.05
[53] 27.15 84.15 68.90 55.10 43.95

> barb <- orderwl (barb)

[1] 27.05 27.15 28.05 28.15 29.05 30.05 30.15 31.15 32.15 39.00 40.00 40.10 41.10
[14] 43.05 43.85 43.95 44.05 55.00 55.10 56.00 56.10 57.10 68.90 69.00 69.10 70.00
[27] 71.10 71.90 72.00 77.00 82.95 83.05 84.15 85.05 91.00 96.95 98.95 105.10 105.90
[40] 106.00 112.90 113.00 116.95 117.95 118.05 119.05 119.15 119.95 120.05 130.90 131.00 132.95
[53] 133.05 140.90 147.00 158.85 160.90</pre>
```

#### 9.6. Conversion to Long- and Wide-Format data.frames

as.data.frame

orderwl

```
as.data.frame extracts the @data slot as a data.frame:
```

```
> flu <- flu [,,400 \tilde{} 407] # make a small and handy version of the flu data set > as.data.frame (flu)
```

```
> colnames (as.data.frame (flu))
[1] "file" "spc" "c"
                        "n"
> as.data.frame (flu) $ spc
        405
             405.5
                        406
                             406.5
                                       407
     27.150
             32.345
                     33.379
                            34.419
                                    36.531
[1,]
[2,]
     66.801 63.715 66.712 69.582 72.530
[3,] 93.144 103.068 106.194 110.186 113.249
[4,] 130.664 139.998 143.798 148.420 152.133
[5,] 167.267 171.898 177.471 184.625 189.752
[6,] 198.430 209.458 215.785 224.587 232.528
Note that the spectra matrix is still a matrix inside column $spc.
as.data.frame and the abbreviations $. and $.. retrieve the usual wide format data.frames:
```

\$.. \$.. > flu\$.

```
file spc.405 spc.405.5 spc.406 spc.406.5 spc.407
1 rawdata/flu1.txt 27.150
                             32.345 33.379
                                               34.419 36.531 0.05 1
2 rawdata/flu2.txt 66.801
                             63.715 66.712
                                               69.582 72.530 0.10 2
3 rawdata/flu3.txt 93.144
                            103.068 106.194
                                              110.186 113.249 0.15 3
4 rawdata/flu4.txt 130.664
                            139.998 143.798
                                             148.420 152.133 0.20 4
5 rawdata/flu5.txt 167.267
                            171.898 177.471
                                             184.625 189.752 0.25 5
6 rawdata/flu6.txt 198.430
                            209.458 215.785
                                              224.587 232.528 0.30 6
> flu$..
             file
```

```
1 rawdata/flu1.txt 0.05 1
2 rawdata/flu2.txt 0.10 2
3 rawdata/flu3.txt 0.15 3
4 rawdata/flu4.txt 0.20 4
5 rawdata/flu5.txt 0.25 5
6 rawdata/flu6.txt 0.30 6
```

If another subset of column needs to be extracted, use [[]]:

```
> flu [[, c ("c", "spc")]]
```

```
c spc.405 spc.405.5 spc.406 spc.406.5 spc.407
1 0.05 27.150
                 32.345 33.379
                                   34.419 36.531
       66.801
2 0.10
                 63.715 66.712
                                   69.582 72.530
3 0.15
       93.144
                103.068 106.194
                                  110.186 113.249
4 0.20 130.664
                139.998 143.798
                                  148,420 152,133
5 0.25 167.267
                171.898 177.471
                                  184.625 189.752
                209.458 215.785
                                  224.587 232.528
6 0.30 198.430
```

This can be combined with extracting certain spectra and wavelengths, see below in subsection "Conversion to Matrix" on page 18.

The transpose of a wide format data.frame can be obtained by as.t.df. For further examples, see as.t.df the discussion of ggplot2 in vignette ("plotting").

[[]]

#### > as.t.df (apply (flu, 2, mean\_pm\_sd))

```
.wavelength mean.minus.sd
                                       mean mean.plus.sd
spc.405
                405.0
                              49.958 113.91
                                                   177.86
spc.405.5
                405.5
                              53.396 120.08
                                                   186.77
spc.406
                              55.352 123.89
                                                   192.43
                406.0
spc.406.5
                406.5
                              57.310 128.64
                                                   199.96
                              59.513 132.79
                                                   206.06
spc.407
                407.0
```

Some functions need the data being an unstacked or long-format data.frame. as.long.df is the appropriate conversion function.

```
> head (as.long.df (flu), 20)
    .wavelength
                                    file
                    spc
          405.0 27.150 rawdata/flu1.txt 0.05 1
2
          405.0 66.801 rawdata/flu2.txt 0.10 2
3
          405.0 93.144 rawdata/flu3.txt 0.15 3
4
          405.0 130.664 rawdata/flu4.txt 0.20 4
          405.0 167.267 rawdata/flu5.txt 0.25 5
          405.0 198.430 rawdata/flu6.txt 0.30 6
6
1.1
          405.5 32.345 rawdata/flu1.txt 0.05 1
2.1
          405.5 63.715 rawdata/flu2.txt 0.10 2
3.1
          405.5 103.068 rawdata/flu3.txt 0.15 3
          405.5 139.998 rawdata/flu4.txt 0.20 4
4.1
5.1
          405.5 171.898 rawdata/flu5.txt 0.25 5
6.1
          405.5 209.458 rawdata/flu6.txt 0.30 6
1.2
          406.0 33.379 rawdata/flu1.txt 0.05 1
          406.0 66.712 rawdata/flu2.txt 0.10 2
2.2
3.2
          406.0 106.194 rawdata/flu3.txt 0.15 3
4.2
          406.0 143.798 rawdata/flu4.txt 0.20 4
5.2
          406.0 177.471 rawdata/flu5.txt 0.25 5
6.2
          406.0 215.785 rawdata/flu6.txt 0.30 6
          406.5 34.419 rawdata/flu1.txt 0.05 1
1.3
          406.5 69.582 rawdata/flu2.txt 0.10 2
2.3
9.7. Conversion to Matrix
```

as.matrix, [[]]

The spectra matrix is extracted by as.matrix, the convenient abbreviation is [[]]:

```
> flu [[]]
```

```
405
              405.5
                        406
                              406.5
                                        407
     27.150 32.345 33.379
[1,]
                             34.419
                                     36.531
[2,] 66.801 63.715 66.712 69.582 72.530
[3,] 93.144 103.068 106.194 110.186 113.249
[4,] 130.664 139.998 143.798 148.420 152.133
[5,] 167.267 171.898 177.471 184.625 189.752
[6,] 198.430 209.458 215.785 224.587 232.528
> class (flu [[]])
```

[1] "matrix"

> rm (flu)

containing parts of the spectra matrix:

```
> flu [[1:3,, 406 ~ 407]]
        406
             406.5
                        407
     33.379
             34.419
                     36.531
[2,] 66.712 69.582 72.530
[3,] 106.194 110.186 113.249
```

If indices for the columns to extract are given, a data.frame is returned instead of a matrix:

```
> flu [[1:3, c ("file", "spc"), 406 ~ 407]]
             file spc.406 spc.406.5 spc.407
1 rawdata/flu1.txt 33.379
                             34.419 36.531
2 rawdata/flu2.txt 66.712
                             69.582 72.530
3 rawdata/flu3.txt 106.194
                            110.186 113.249
```

#### 10. Combining and Decomposing hyperSpec Objects

#### 10.1. Binding Objects together

hyperspec Objects can be bound together, either by columns (cbind) to append a new spectral range or by row (rbind) to append new spectra:

```
> dim (flu)
nrow ncol nwl
6  3 181
> dim (cbind (flu, flu))
nrow ncol nwl
6  3 362
> dim (rbind (flu, flu))
nrow ncol nwl
12  3 181
```

There is also a more general function, bind, taking the direction ("r" or "c") as first argument followed by the objects to bind either in separate arguments or in a list.

As usual for **rbind** and **cbind**, the objects that should be bound together must have the same rows and columns, respectively.

For binding row-wise (rbind), collapse is more flexible but also faster.

collapse

#### 10.2. Binding Objects that do not Share the Same Extra Data and/or Wavelength Axis

collapse combines objects that should be bound together by row, but they do not share the columns and/or spectral range. The resulting object has all columns from all input objects, and all wavelengths from the input objects. If an input object does not have a particular column or wavelength, its value in the resulting object is NA.

collapse

The barbiturates data is a list of 286 hyperSpec objects, each containing one mass spectrum. The spectra have between 4 and 101 data points each.

```
> barb <- collapse (barbiturates)
> wl (barb) [1 : 25]

[1] 160.90 158.85 147.00 140.90 133.05 130.90 119.95 119.15 118.05 116.95 112.90 106.00 105.10
[14] 98.95 96.95 91.00 85.05 83.05 77.00 71.90 71.10 70.00 69.00 57.10 56.10
```

The resulting object does not have an ordered wavelength axis. This can be obtained in a second step:

```
> barb <- orderwl (barb)</pre>
> barb [[1:3, , min ~ min + 10i]]
    25.95 26.05 26.15 26.95 27.05 27.15 28.05 28.15 29.05 29.15 29.95
           NA NA NA 562 NA
Γ1. ]
       NA
                                       NA 11511 6146
                                                              NA
[2,]
       NΑ
            NΑ
                 NΑ
                       NA
                            NΑ
                                 618 10151
                                            NA 5040
                                                        NA
                                                              NA
                            638
[3,]
            NA
                  NA
                       NA
                                  NA
                                        NA 10722 5253
```

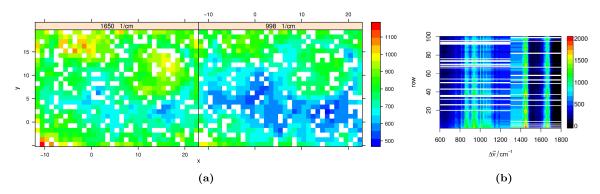


Figure 3 (a) For both spectral ranges some spectra are missing. (b) The missing parts of the spectra are filled with NA.

#### 10.3. Binding Objects that do not Share the Same Spectra

merge adds a new spectral range (like cbind), but works also if spectra are missing in one of the objects. The arguments by, by.x, and by.y specify which columns should be used to decide which spectra are the same. The arguments all, all.x, and all.y determine whether spectra should be kept for the result set if they appear in only one of the objects. For details, see also the help on the base function merge.

As an example, let's construct a version of the chondro data like being taken as two maps with different spectral ranges. In each data set, some spectra are missing.

```
> chondro.low <- sample (chondro [,, 600 ~ 1200], 700)
> nrow (chondro.low)
[1] 700
> chondro.high <- sample (chondro [,, 1400 ~ 1800], 700)
> nrow (chondro.high)
```

As all extra data columns are the same, no special declarations are needed for merging the data:

```
> chondro.merged <- merge (chondro.low, chondro.high)
> nrow (chondro.merged)
```

[1] 557

By default, the result consists of only those spectra, where *both* spectral ranges were available. To keep all spectra replacing missing parts by NA (see fig. 3):

```
> chondro.merged <- merge (chondro.low, chondro.high, all = TRUE)
> nrow (chondro.merged)
```

[1] 843

> merged <- merge (chondro [1:7,, 610  $\tilde{}$  620], chondro [5:10,, 615  $\tilde{}$  625], all = TRUE) > merged\$.

```
x clusters .nx .ny spc.610 spc.614 spc.618 spc.614 spc.618 spc.622 spc.626
-4.77 -11.55
                                488.63
                                       466.18
                                                492.00
               matrix
                        1 NA
                                                             NA
                                                                      NA
                                                                              NA
                                                                                      NA
-4.77 -10.55
                                                                      NA
               matrix
                         2 NA
                                489.48
                                        465.05
                                                490.53
                                                             NA
                                                                              NΑ
                                                                                      NΑ
                         3
                            NA
                                456.03
                                        436.62
                                                 458.06
                                                             NA
                                                                      NA
                                                                              NA
                                                                                       NA
       -9.55
               {\tt matrix}
                                464.82
                                                 470.02
       -8.55
                         4
                           NA
                                        444.85
               matrix
                                                             NA
                                                                      NA
                                                                              NA
                                                                                       NA
```

```
1 428.66
                                         410.80
                                                 433.12 410.80 433.12
  -4.77
         -7.55
                 matrix
                          5
                                                                         461.19
  -4.77
         -6.55
                              2
                                 426.07
                                         407.86
                                                 431.21
                                                         407.86
                                                                         458.15
                                                                                 394.18
6
                          6
                                                                 431.21
                 matrix
  -4.77
         -5.55
                          7
                              3
                                 412.37
                                         396.50
                                                 421.27
                                                         396.50
                                                                 421.27
                                                                         445.54
                                                                                 382.72
                 lacuna
8
  -4.77
         -4.55
                 lacuna
                         NA
                              4
                                     NA
                                             NA
                                                     NA
                                                         381.95
                                                                 406.25
                                                                         429.67
                                                                                 368.46
  -4.77
         -3.55
                 lacuna
                         NΑ
                              5
                                     NΑ
                                             NΑ
                                                     NΑ
                                                         397.51
                                                                 423.30
                                                                         446.15
                                                                                 381.87
         -2.55
                                                         377.39
                                                                 402.23
                 lacuna
                              6
                                     NA
                                             NA
                                                     NA
                                                                         424.19
```

If the spectra overlap, the result will have both data points. In the example here one could easily delete duplicate wavelengths. For real data, however, the duplicated wavelength will hardly ever contain the same values. The appropriate method to deal with this situation depends on the data at hand, but it will usually be some kind of spectral interpolation.

One possibility is removing duplicated wavelengths by using the mean intensity. This can conveniently be done by using approx using method = "constant". For duplicated wavelengths, the intensities will be combined by the tie function. This already defaults to the mean, but we need na.rm = TRUE.

Thus, the function to calculate the new spectral intensities is

```
> approxfun <- function (y, wl, new.wl){
    approx (wl, y, new.wl, method = "constant",
             ties = function (x) mean (x, na.rm = TRUE)
             )$v
+ }
which can be applied to the spectra:
> merged <- apply (merged, 1, approxfun,
                    wl = wl (merged), new.wl = unique (wl (merged)),
                    new.wavelength = "new.wl")
> merged$.
             x clusters .nx .ny spc.610 spc.614 spc.618 spc.622 spc.626
   -4.77 -11.55
                 matrix
                          1 NA 488.63
                                        466.18
                                                 492.00
                                                             NA
   -4.77 -10.55
                          2 NA
                                 489.48
                                         465.05
                                                 490.53
                                                             NA
                                                                      NA
2
                 matrix
3
  -4.77
         -9.55
                 {\tt matrix}
                          3
                             NA
                                 456.03
                                         436.62
                                                 458.06
                                                             NA
                                                                     NA
   -4.77
         -8.55
                 matrix
                          4 NA
                                 464.82
                                         444.85
                                                  470.02
                                                             NA
                                                                 397.38
         -7.55
                          5
                                 428.66
                                         410.80
                                                         461.19
5
  -4.77
                 matrix
                              1
                                                 433.12
   -4.77
          -6.55
                 matrix
                          6
                              2
                                 426.07
                                         407.86
                                                  431.21
                                                          458.15
                                                                 394.18
   -4.77
         -5.55
                          7
                              3
                                 412.37
                                         396.50
                                                 421.27
                                                         445.54
                                                                 382.72
                 lacuna
                                         381.95
8
  -4.77
         -4.55
                 lacuna NA
                              4
                                     NΑ
                                                 406.25
                                                         429.67
                                                                 368.46
   -4.77
          -3.55
                  lacuna
                         NA
                              5
                                     NA
                                         397.51
                                                  423.30
                                                          446.15
                                                                 381.87
10 -4.77
         -2.55
                  lacuna
                         NA
                              6
                                     NA
                                         377.39
                                                 402.23
                                                         424.19
                                                                 362.43
```

#### 10.4. Matrix Multiplication

Two hyperSpec objects can be matrix multiplied by %\*%. For an example, see the principal component analysis below (section 13.1 on page 30).

#### 10.5. Decomposition

Matrix decompositions are common operations during chemometric data analysis. The results, e.g. of a principal component analysis are two matrices, the so-called scores and loadings. The results can have either the same number of rows as the spectra matrix they were calculated from (scores-like), or they have as many wavelengths as the spectra (loadings-like).

Both types of result objects can be "re-imported" into hyperSpec objects with function decomposition. decomposition

A scores-like object retains all per-spectrum information (i.e. the extra data) while the spectra matrix and wavelength vector are replaced. A loadings-like object retains the wavelength information, while extra data is deleted (set to NA) unless the value is constant for all spectra.

A demonstration can be found in the principal component analysis example (section 13.1) on page 30.

#### 11. Plotting

hyperSpec offers a variety of possibilities to plot spectra, spectral maps, the spectra matrix, time series, depth profiles, etc.. This all is discussed in a separate document: see vignette ("plotting").

#### 12. Spectral (Pre)processing

#### 12.1. Cutting the Spectral Range

[] [[]]

The extraction functions [] and [[]] can be used to cut the spectra: Their third argument takes wavelength specifications as discussed above and also logicals (i.e. vectors specifying with TRUE/FALSE for each column of \$spc whether it should be included or not.

[] returns a hyperSpec object, [[]] the spectra matrix \$spc (or the data.frame @data if in addition data columns were specified) only.

```
> flu [,, min ~ 408.5]
hyperSpec object
   6 spectra
   3 data columns
   8 data points / spectrum
wavelength: lambda/nm [numeric] 405.0 405.5 ... 408.5
data: (6 rows x 3 columns)
   1. file: [factor] rawdata/flu1.txt rawdata/flu2.txt ... rawdata/flu6.txt
   2. spc: I[f1]/"a.u." [matrix8] 27.150 66.801 ... 256.89
   3. c: c / (mg / 1) [numeric] 0.05 0.10 ... 0.3
> flu [[,, c (min ~ min + 2i, max - 2i ~ max)]]
         405
              405.5
                        406
                                494
                                      494.5
     27.150
             32.345
                     33.379
                             47.163
                                     46.412
                                             45,256
[1,]
     66.801 63.715 66.712 96.602
[2.]
                                     96.206 94.610
[3,] 93.144 103.068 106.194 149.539 148.527 145.793
[4,] 130.664 139.998 143.798 201.484 198.867 195.867
[5,] 167.267 171.898 177.471 252.066 248.067 246.952
[6,] 198.430 209.458 215.785 307.519 302.325 294.649
```

#### 12.2. Shifting Spectra

Sometimes, spectra need to be aligned along the spectral axis.

In general, two options are available for shifting spectra along the wavelength axis.

- 1. The wavelength axis can be shifted, while the intensities stay unaffected.
- 2. the spectra are interpolated onto a new wavelength axis, while the nominal wavelengths stay.

The first method is very straightforward (see fig 4a):

```
> tmp <- chondro
> wl (tmp) <- wl (tmp) - 10
```

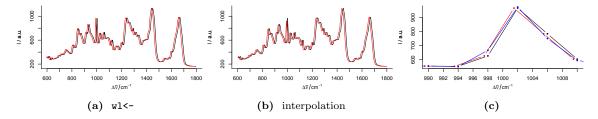


Figure 4 Shifting the Spectra along the Wavelength Axis. (a) Changing the wavelength values. (b) Interpolation. (c) Detail view of the phenylalanine band: shifting by wl<- (red) does not affect the intensities, while the spectrum is slightly changed by interpolations (blue).

but it cannot be used if each spectrum (or groups of spectra) are shifted individually.

In that case, interpolation is needed. R offers many possibilities to interpolate (e.g. approx for constant / linear approximation, spline for spline interpolation, loess can be used to obtain smoothed approximations, etc.). The appropriate interpolation strategy will depend on the spectra, and hyperSpec therefore leaves it up to the user to select a sensible interpolation function.

As an example, we will use natural splines to do the interpolation. It is convenient to set it up as a function:

```
> interpolate <- function (spc, shift, wl){
+    spline (wl + shift, spc, xout = wl, method = "natural")$y
+ }
This function can now be applied to a set of spectra (see fig 4b):
> tmp <- apply (chondro, 1, interpolate, shift = -10, wl = wl (chondro))
If different spectra need to be offset by different shift, use a loop<sup>2</sup>
> shifts <- rnorm (nrow (chondro))
> tmp <- chondro [[]]
> for (i in seq_len (nrow (chondro)))
+    tmp [i, ] <- interpolate (tmp [i, ], shifts [i], wl = wl (chondro))
> chondro [[]] <- tmp</pre>
```

#### 12.2.1. Calculating the Shift

Often, the shift in the spectra is determined by aligning a particular signal. This strategy works best with spectrally oversampled data that allows accurate determination of the signal position.

For the chondro data, let's use the maximum of the phenylalanine band between 990 and 1020 cm<sup>-1</sup>. As just the very maximum is too coarse, we'll use the maximum of a square polynomial fitted to the maximum and its two neighbours.

```
> find.max <- function (y, x){
+  pos <- which.max (y) + (-1:1)
+  X <- x [pos] - x [pos [2]]
+  Y <- y [pos] - y [pos [2]]
+
+  X <- cbind (1, X, X^2)</pre>
```

<sup>&</sup>lt;sup>2</sup>sweep cannot be used here, and while there is the possibility to use sapply or mapply, they are not faster than the for loop in this case. Make sure to work on a copy of the spectra matrix, as that is much faster than row-wise extracting and changing the spectra by [[ and [[<-.

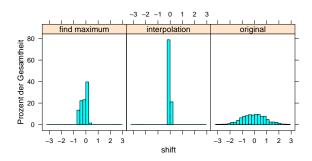


Figure 5 The shifts used to disturb the chondrocyte data (original), and the remaining shift after correction with the two methods discussed here.

```
coef <- qr.solve (X, Y)</pre>
      coef [2] / coef [3] / 2 + x [pos [2]]
+ }
> bandpos <- apply (chondro [[,, 990 ~ 1020]], 1, find.max, wl (chondro [,, 990 ~ 1020]))
> refpos <- find.max (colMeans (chondro[[,, 990 ~ 1020]]), wl (chondro [,, 990 ~ 1020]))</pre>
> shift1 <- refpos - bandpos
A second possibility is to optimize the shift. For this strategy, the spectra must be sufficiently similar,
while low spectral resolution is compensated by using larger spectral windows.
> chondro <- chondro - spc.fit.poly.below (chondro [,,min+3i ~ max - 3i], chondro)</pre>
Fitting with npts.min = 15
> chondro <- sweep (chondro, 1, rowMeans (chondro [[]], na.rm = TRUE), "/")</pre>
> targetfn <- function (shift, wl, spc, targetspc){</pre>
    error <- spline (wl + shift, spc, xout = wl)$y - targetspc
    sum (error^2)
+ }
> shift2 <- numeric (nrow (chondro))</pre>
> tmp <- chondro [[]]
> target <- colMeans (chondro [[]])</pre>
> for (i in 1 : nrow (chondro))
    shift2 [i] <- unlist (optimize (targetfn, interval = c (-5, 5), wl = chondro@wavelength,
                                     spc = tmp[i,], targetspc = target)$minimum)
```

Figure 5 shows that the second correction method works better for the chondrocyte data. This was expected, as the spectra are hardly or not oversampled, but are very similar to each other.

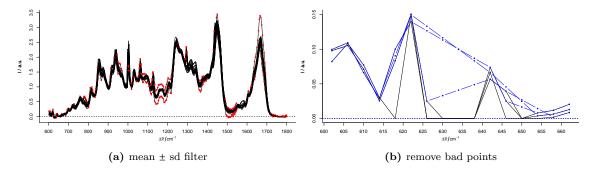
#### 12.3. Removing Bad Data

#### 12.3.1. Bad Spectra

Occasionally, one may want to remove spectra because of too low or too high signal.

E.g. for infrared spectra one may state that the absorbance maximum should be, say, between 0.1 and 1. *hyperSpec*'s comparison operators return a logical matrix of the size of the spectra that is suitable for later indexing:

```
> ir.spc <- chondro / 1500 ## fake IR data
> high.int <- apply (ir.spc > 1, 1, any) # any point above 1 is bad
> low.int <- apply (ir.spc, 1, max) < 0.1 # the maximum should be at least 0.1
> ir.spc <- ir.spc [! high.int & ! low.int]</pre>
```



 ${\bf Figure} \,\, {\bf 6} \quad {\rm filtering} \,\, {\rm data}$ 

#### 12.3.2. Removing Spectra outside mean $\pm n$ sd

#### 12.3.3. Bad Data Points

Assume the data contains once in a while a detector readout of 0:

```
> spc <- chondro [1 : 3,, min ~ min + 15i]
> spc [[cbind (1:3, sample (nwl (spc), 3)), wl.index = TRUE]] <- 0
> spc [[]]
                                                618
          602
                   606
                             610
                                      614
                                                         622
                                                                     626
                                                                                630
                                                                                          634
[1,] 0.097953 0.10528 0.071208 0.025446 0.083304 0.14991 -0.0031208 -0.050398 -0.056964 -0.0252394
[2,] 0.099638 0.10940 0.066150 0.028939 0.100018 0.13981 -0.0140269 -0.049880 -0.055121 -0.0071423
 \hbox{\tt [3,]} \ \ 0.082254 \ \ 0.10830 \ \ 0.077309 \ \ 0.030292 \ \ 0.000000 \ \ 0.15056 \quad \  0.0250632 \ \ -0.053219 \ \ -0.063573 \ \ -0.0405063 
           642
                    646
                                650
                                             654
                                                        658
                                                                   662
[1,] 0.066517 0.000000 -0.0035541 0.00374453 0.0063589 0.0128233
[2,] 0.073752 0.025309 0.0000000 0.00796657 0.0117685 0.0204543
[3,] 0.056744 0.039410 -0.0035429 -0.00099433 0.0018429 0.0086044
```

We can set these points to NA, again using that the comparison returns a suitable logical matrix:

```
> spc [[spc < 1e-4]] <- NA
> spc [[]]
                 606
                          610
                                   614
                                            618
                                                    622
                                                             626 630 634 638
                                                                                           646 650
[1,] 0.097953 0.10528 0.071208 0.025446 0.083304 0.14991
                                                              NA NA NA NA 0.066517
                                                                                            NA NA
[2,] 0.099638 0.10940 0.066150 0.028939 0.100018 0.13981
                                                                          NA 0.073752 0.025309
                                                              NA
                                                                  NA
                                                                     NA
                                                                                                NA
[3,] 0.082254 0.10830 0.077309 0.030292
                                             NA 0.15056 0.025063 NA
                                                                     NA
                                                                          NA 0.056744 0.039410
          654
                    658
                              662
[1,] 0.0037445 0.0063589 0.0128233
[2,] 0.0079666 0.0117685 0.0204543
[3,]
           NA 0.0018429 0.0086044
```

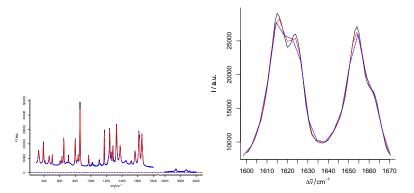


Figure 7 Smoothing interpolation by spc.loess with new data point spacing of 2 cm<sup>-1</sup> (red) and spc.bin (blue). The magnification on the right shows how interpolation may cause a loss in signal height.

Depending on the type of analysis, one may wants to replace the NAs by interpolating the neighbour values. So far, *hyperSpec* provides three functions that can interpolate the NAs: : spc.NA.linapprox, spc.loess, and spc.bin with na.rm = TRUE (the latter two are discussed below).

```
spc.NA.linapprox,
spc.loess,
spc.bin
```

```
> spc.corrected <- spc.NA.linapprox (spc)
> spc.corrected [[]]
          602
                  606
                           610
                                    614
                                             618
                                                     622
                                                              626
                                                                       630
                                                                                 634
                                                                                          638
[1,] 0.097953 0.10528 0.071208 0.025446 0.083304 0.14991 0.133229 0.116551 0.099873 0.083195
[2,] 0.099638 0.10940 0.066150 0.028939 0.100018 0.13981 0.126601 0.113389 0.100176 0.086964
[3,] 0.082254 0.10830 0.077309 0.030292 0.090425 0.15056 0.025063 0.032983 0.040903 0.048824
          642
                   646
                            650
                                      654
                                                658
                                                          662
[1,] 0.066517 0.045593 0.024669 0.0037445 0.0063589 0.0128233
[2,] 0.073752 0.025309 0.016638 0.0079666 0.0117685 0.0204543
[3,] 0.056744 0.039410 0.026888 0.0143654 0.0018429 0.0086044
```

#### 12.3.4. Spikes in Raman Spectra

...coming soon...

#### 12.4. Smoothing Interpolation

spc.bin

Spectra acquired by grating instruments are frequently interpolated onto a new wavelength axis, e.g. because the unequal data point spacing should be removed. Also, the spectra can be smoothed: reducing the spectral resolution allows to increase the signal to noise ratio. For chemometric data analysis reducing the number of data points per spectrum may be crucial as it reduces the dimensionality of the data.

hyperSpec provides two functions to do so: spc.bin and spc.loess.

spc.bin bins the spectral axis by averaging every by data points.

```
> plot (paracetamol, wl.range = c (300 ~ 1800, 2800 ~ max), xoffset = 850)
> p <- spc.loess (paracetamol, c(seq (300, 1800, 2), seq (2850, 3150, 2)))
> plot (p, wl.range = c (300 ~ 1800, 2800 ~ max), xoffset = 850, col = "red", add = TRUE)
> b <- spc.bin (paracetamol, 4)
> plot (b, wl.range = c (300 ~ 1800, 2800 ~ max), xoffset = 850,
+ lines.args = list (pch = 20, cex = .3, type = "p"), col = "blue", add = TRUE)
```

spc.loess applies R's loess function for spectral interpolation. Figure 7 shows the result of interpolating from 300 to 1800 and 2850 to 3150 cm<sup>-1</sup> with 2 cm<sup>-1</sup> data point distance. This corresponds to a spectral resolution of about 4 cm<sup>-1</sup>, and the decrease in spectral resolution can be seen at the sharp bands where the maxima are not reached (due to the fact that the interpolation wavelength axis does not necessarily hit the maxima. The original spectrum had 4064 data points with unequal data point spacing (between 0 and 1.4 cm<sup>-1</sup>). The interpolated spectrum has 902 data points.

#### 12.5. Background Correction

sweep

To subtract a background spectrum of each of the spectra in an object, use sweep (spectra, 2, background.spectrum, "-").

#### 12.6. Offset Correction

apply sweep

Calculate the offsets and sweep them off the spectra:

```
> offsets <- apply (chondro, 1, min)
> chondro.offset.corrected <- sweep (chondro, 1, offsets, "-")</pre>
```

If the offset is calculated by a function, as here with the min, hyperSpec's sweep method offers a shortcut: sweep's STATS argument may be the function instead of a numeric vector:

```
> chondro.offset.corrected <- sweep (chondro, 1, min, "-")
```

#### 12.7. Baseline Correction

hyperSpec comes with two functions to fit polynomial baselines.

spc.fit.poly
spc.fit.poly.below

spc.fit.poly fits a polynomial baseline of the given order. A least-squares fit is done so that the function may be used on rather noisy spectra. However, the user must supply an object that is cut appropriately. Particularly, the supplied wavelength ranges are not weighted.

spc.fit.poly.below tries to find appropriate support points for the baseline iteratively.

Both functions return a hyperSpec object containing the fitted baselines. They need to be subtracted afterwards:

```
> bl <- spc.fit.poly.below (chondro)
Fitting with npts.min = 15
> chondro <- chondro - bl</pre>
```

For details, see vignette (baselinebelow).

Package baseline [1] offers many more functions for baseline correction. The baseline function works on the spectra matrix, which is extracted by [[]]. The result is a baseline object, but can easily be re-imported into the hyperSpec object:

```
> corrected <- hyperSpec::chondro [1] # start with the unchanged data set
> require ("baseline")
> bl <- baseline (corrected [[]], method = "modpolyfit", degree = 4)
> corrected [[]] <- getCorrected (bl)</pre>
```

Fig. 8 shows the result for the first spectrum of chondro.

```
> rm (bl, chondro)
```

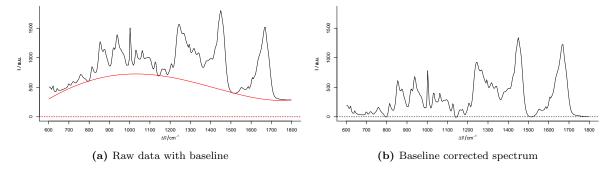


Figure 8 Baseline correction using the baseline package: the first spectrum of chondro with baseline (left) and after baseline correction (right) with method "modpolyfit".

#### 12.8. Intensity Calibration

#### 12.8.1. Correcting by a constant, e.g. Readout Bias

CCD cameras often operate with a bias, causing a constant value for each pixel. Such a constant can be immediately subtracted:

spectra - constant

#### 12.8.2. Correcting Wavelength Dependence

sweep

For each of the wavelengths the same correction needs to be applied to all spectra.

1. There might be wavelength dependent offsets (background or dark spectra). They are subtracted:

```
sweep (spectra, 2, offset.spectrum, "-")
```

2. A multiplicative dependency such as a CCD's photon efficiency: sweep (spectra, 2, photon.efficiency, "/")

#### 12.8.3. Spectra Dependent Correction

sweep

If the correction depends on the spectra (e.g. due to inhomogeneous illumination while collecting imaging data, differing optical path length, etc.), the *MARGIN* of the **sweep** function needs to be 1 or SPC:

- Pixel dependent offsets are subtracted: sweep (spectra, SPC, pixel.offsets, "-")
- A multiplicative dependency: sweep (spectra, SPC, illumination.factors, "\*")

#### 12.9. Normalization

apply sweep

Again, sweep is the function of choice. E.g. for area normalization, use:

```
> chondro <- sweep (chondro, 1, mean, "/")</pre>
```

(using the mean instead of the sum results in conveniently scaled spectra with intensities around 1.)

If the calculation of the normalization factors is more elaborate, use a two step procedure:

- Calculate appropriate normalization factors
   You may calculate the factors using only a certain wavelength range, thereby normalizing on
   a particular band or peak.
- 2. Again, sweep the factor off the spectra:
   normalized <- sweep (spectra, 1, factors, "\*")</pre>

```
> factors <- 1 / apply (chondro [, , 1600 ~ 1700], 1, mean)
> chondro <- sweep (chondro, 1, factors, "*")</pre>
```

For minimum-maximum-normalization, first do an offset- or baseline correction, then normalize using max.

#### 12.10. Centering and Variance Scaling the Spectra

scale

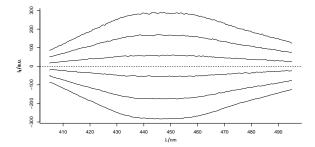
Centering means that the mean spectrum is subtracted from each of the spectra. Many data analysis techniques, like principal component analysis, partial least squares, etc., work much better on centered data. From a spectroscopic point of view it depends on the particular data set whether centering does make sense or not.

Variance scaling is often used in multivariate analysis to adjust the influence and scaling of the variates (that are typically different physical values). However, spectra already do have the same scale of the same physical value. Thus one has to trade off the the expected numeric benefit with the fact that for wavelengths with low signal the noise level will "explode" by variance scaling. Scaling usually makes sense only for centered data.

Both tasks are carried out by the same method in R, scale, which will by default both mean center and variance scale the spectra matrix.

To center the flu data set, use:

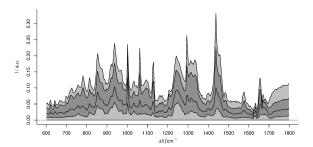
> flu.centered <- scale (flu, scale = FALSE)
> plot (flu.centered)



On the other hand, the chondro data set consists of Raman spectra, so the spectroscopic interpretation of centering is getting rid of the the average chemical composition of the sample. But: what is the meaning of the "average spectrum" of an inhomogeneous sample? In this case it may be better to subtract the minimum spectrum (which will hopefully have almost the same benefit on the data analysis) as it is the spectrum of that chemical composition that is underlying the whole sample.

One more point to consider is that the actual minimum spectrum will pick up (negative) noise. In order to avoid that, using e.g. the 5<sup>th</sup> percentile spectrum is more suitable:

> chondro <- scale (chondro, center = quantile (chondro, 0.05), scale = FALSE)
> plot (chondro, "spcprctl5")



See section 14 (p. 14) for some tips to speed up these calculations.

#### 12.11. Multiplicative Scatter Correction (MSC)

pls::msc

MSC can be done using msc from package pls[2]. It operates on the spectra matrix:

- > require (pls)
- > chondro.msc <- chondro
- > chondro.msc [[]] <- msc (chondro [[]])</pre>

#### 12.12. Spectral Arithmetic

+ - \* / ^ log log10

Basic mathematical functions are defined for *hyperSpec* objects. You may convert spectra: absorbance.spectra = - log10 (transmission.spectra)

In this case, do not forget to adapt the label:

labels

> labels (absorbance.spectra)\$spc <- "A"</pre>

Be careful: R's log function calculates the natural logarithm if no base is given.

The basic arithmetic operators work element-wise in R. Thus they all need either a scalar, or a matrix (or *hyperSpec* object) of the correct size.

Matrix multiplication is done by **%\*%**, again each of the operands may be a matrix or a *hyperSpec* %\*% object, and must have the correct dimensions.

#### 13. Data Analysis

## 13.1. Data Analysis Methods using a data.frame e.g. Principal Component Analysis with prcomp

\$.

The \$. notation is handy, if a data analysis function expects a *data.frame*. The column names can then be used in the formula:

```
> pca <- prcomp (~ spc, data = chondro$., center = FALSE)
```

Many modeling functions call as.data.frame on their data argument. In that case, the conversion is done automatically:

> pca <- prcomp (~ spc, data = chondro, center = FALSE)

Results of such a decomposition can be put again into *hyperSpec* objects. This allows to plot e.g. decomposition the loading like spectra, or score maps, see figure 9.

```
> scores <- decomposition (chondro, pca$x, label.wavelength = "PC",
                             label.spc = "score / a.u.")
> scores
hyperSpec object
   875 spectra
   4 data columns
   300 data points / spectrum
wavelength: PC [integer] 1 2 ... 300
data: (875 rows x 4 columns)
   1. y: y [numeric] -4.77 -4.77 ... 19.23
   2. x: x [numeric] -11.55 -10.55 ... 22.45
   3. clusters: clusters [factor] matrix matrix ... lacuna + NA
   4. spc: spc [matrix300] -0.43543 -0.92192 ... -1.1102e-16
The loadings can be similarly re-imported:
> loadings <- decomposition (chondro, t(pca$rotation), scores = FALSE,
                               label.spc = "loading I / a.u.")
> loadings
hyperSpec object
   300 spectra
   1 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (300 rows x 1 columns)
   1. spc: spc [matrix300] -0.0258979 -0.0014762 ... 0.013207
```

There is, however, one important difference. The loadings are thought of as values computed from all spectra together. Thus no meaningful extra data can be assigned for the loadings object (at least not if the column consists of different values). Therefore, the loadings object lost all extra data (see above).

retain.columns triggers whether columns that contain different values should be dropped. If it is set to TRUE, the columns are retained, but contain NAs:

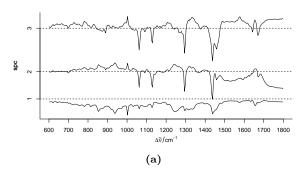
If an extra data column does contain only one unique value, it is retained anyways:

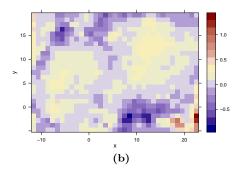
```
> chondro$measurement <- 1
> loadings <- decomposition (chondro, t(pca$rotation), scores = FALSE,
+ label.spc = "loading I / a.u.")
> loadings[1]$..

    measurement
PC1 1
```

#### 13.1.1. PCA as Noise Filter

Principal component analysis is sometimes used as a noise filtering technique. The idea is that the relevant differences are captured in the first components while the higher components contain noise only. Thus the spectra are reconstructed using only the first p components.





%\*%

Figure 9 (a) The first three loadings: plot (loadings [1 : 3], stacked = TRUE). (b) The third score map: plotmap (scores [, , 3]).

This reconstruction is in fact a matrix multiplication:

$$spectra^{(nrow \times nwl)} = scores^{(nrow \times p)}loadings^{(p \times nwl)}$$

Note that this corresponds to a model based on the Beer-Lambert law:

$$A_n(\lambda) = c_{n,i}\epsilon(i,\lambda) + error$$

The matrix formulation puts the n spectra into the rows of A and c, while the i pure components appear in the columns of c and rows of the absorbance coefficients  $\epsilon$ .

For an ideal data set (constituents varying independently, sufficient signal to noise ratio) one would expect the principal component analysis to extract something like the concentrations and pure component spectra.

If we decide that only the first 10 components actually carry spectroscopic information, we can reconstruct spectra with better signal to noise ratio:

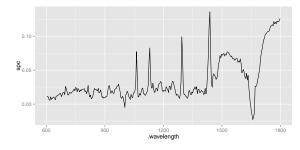
> smoothed <- scores [,, 1:10] %\*% loadings [1:10]

Keep in mind, though, that we cannot be sure how much useful information was discarded with the higher components. This kind of noise reduction may influence further modeling of the data. Mathematically speaking, the rank of the new  $875 \times 300$  spectra matrix is only 10.

# 13.2. Data Analysis using long-format data.frame e.g. plotting with ggplot2

Some functions need the data being an *unstacked* or *long-format data.frame*. as.long.df is the appropriate conversion function.

- > require (ggplot2)
- > ggplot (as.long.df (chondro [1]), aes  $(x = .wavelength, y = spc)) + geom_line ()$



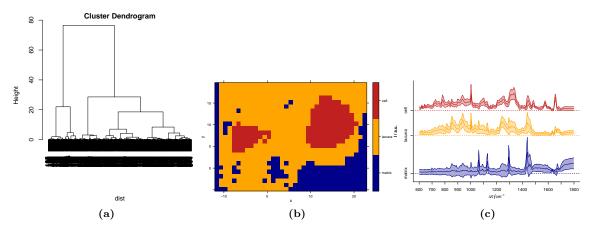


Figure 10 The results of the cluster analysis: (a) the dendrogram (b) the map of the 3 clusters (c) the mean spectra.

# 13.3. Data Analysis Methods using a matrix e.g. Hierarchical Cluster Analysis

Some functions expect their input data in a matrix, so either as.matrix (object) or the abbreviation object [[]] can be used:

> dist <- pearson.dist (chondro [[]])</pre>

Again, many such functions coerce the data to a matrix automatically, so the *hyperSpec* object can be handed over:

- > dist <- pearson.dist (chondro)
  > dendrogram <- hclust (dist, method = "ward")</pre>
- > plot (dendrogram)

In order to plot a cluster map, the cluster membership needs to be calculated from the dendrogram.

First, cut the dendrogram so that three clusters result:

> chondro\$clusters <- as.factor (cutree (dendrogram, k = 3))</pre>

As the cluster membership was stored as factor, the levels can be meaningful names, which are displayed in the color legend.

> levels (chondro\$clusters) <- c ("matrix", "lacuna", "cell")</pre>

Then the result may be plotted (figure 10b):

### 13.4. Calculating group-wise Sum Characteristics,

#### e.g. Cluster Mean Spectra

aggregate applies the function given in FUN to each of the groups of spectra specified in by.

aggregate

[[]]

So we may plot the cluster mean spectra:

```
> means <- aggregate (chondro, by = chondro$clusters, mean_pm_sd)
> plot (means, col = cluster.cols, stacked = ".aggregate", fill = ".aggregate")
```

#### 13.5. Splitting an Object, and Binding a List of hyperSpec Objects

split

A hyperSpec object may also be split into a list of hyperSpec objects:

```
> clusters <- split (chondro, chondro$clusters)</pre>
> clusters
$matrix
hyperSpec object
   187 spectra
   5 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (187 rows x 5 columns)
   1. y: y [numeric] -4.77 -4.77 ... 19.23
   2. x: x [numeric] -11.55 -10.55 ... -11.55
   3. clusters: clusters [factor] matrix matrix ... matrix
   4. spc: I / a.u. [matrix300] 0.011964 0.022204 ... 0.13706
   5. measurement: measurement [numeric] 1 1 ... 1
$lacuna
hyperSpec object
   546 spectra
   5 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (546 rows x 5 columns)
   1. y: y [numeric] -4.77 -4.77 ... 19.23
   2. x: x [numeric] -8.55 -7.55 ... 22.45
   3. clusters: clusters [factor] lacuna lacuna ... lacuna
   4. spc: I / a.u. [matrix300] 0.038900 0.031386 ... 0.049803
   5. measurement: measurement [numeric] 1 1 ... 1
$cell
hyperSpec object
   142 spectra
   5 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (142 rows x 5 columns)
   1. y: y [numeric] 4.23 4.23 ... 16.23
   2. x: x [numeric] -7.55 -6.55 ... 14.45
   3. clusters: clusters [factor] cell cell ... cell
   4. spc: I / a.u. [matrix300] 0.024574 0.027541 ... 0.017377
   5. measurement: measurement [numeric] 1 1 ... 1
```

Splitting can be reversed by **rbind** (see section 10.1, page 19). Another, similar way to combine a number of *hyperSpec* objects with different wavelength axes or extra data columns is **collapse** (see section 10.2, page 19).

#### 14. Speed and Memory Considerations

While most of hyperSpec's functions work at a decent speed for interactive sessions (of course depending on the size of the object), iterated (repeated) calculations as for bootstrapping or iterated cross validation may ask for special speed considerations.

As an example, let's again consider the code for shifting the spectra:

```
> tmp <- chondro [1 : 50]
> shifts <- rnorm (nrow (tmp))
> system.time ({
```

```
+ for (i in seq_len (nrow (tmp)))
+ tmp [[i]] <- interpolate (tmp [[i]], shifts [i], wl = wl (tmp))
+ })

user system elapsed
0.060 0.001 0.061</pre>
```

Calculations that involve a lot of subsetting (i.e. extracting or changing the spectra matrix or extra data) can be sped up considerably if the required parts of the *hyperSpec* object are extracted beforehand. This is somewhat similar to model fitting in R in general: many model fitting functions in R are much faster if the formula interface is avoided and the appropriate *data.frames* or matrices are handed over directly.

```
> tmp <- chondro [1 : 50]
> system.time ({
+    tmp.matrix <- tmp [[]]
+    wl <- wl (tmp)
+    for (i in seq_len (nrow (tmp)))
+       tmp.matrix [i, ] <- interpolate (tmp.matrix [i, ], shifts [i], wl = wl)
+    tmp [[]] <- tmp.matrix
+ })

    user system elapsed
    0.007    0.000    0.005</pre>
```

#### Additional packages.

matrixStats implements fast functions to calculate summary statistics for each row or each column of a matrix. This functionality can be enabled for hyperSpec by installing package hyperSpec.matrixStats which is available in hyperSpec's development repository at http://hyperSpec.r-forge.r-project.org/

scale implements a fast replacement for base::scale implemented in C++. It is not yet publicly available, but you can email me (Claudia Beleites <chemometrie@beleites.de>) for the source package.

**Compiled code.** R provides interfaces to Fortran and C code, see the manual "Writing R Extensions". Rcpp[3, 4] allows to conveniently integrate C++ code. inline[5] adds another layer of convenience: inline definition of functions in C, C++, or Fortran.

An intermediate level is byte compilation of R code, which is done by *compiler*[6].

**Memory use.** In general, it is recommended not to work with variables that are more than approximately a third of the available RAM in size. Particularly the import of raw spectroscopic data can consume large amounts of memory. At certain points, *hyperSpec* provides switches that allow working with data sets that are actually close to this memory limit.

The initialization method new ("hyperSpec", ...) takes particular care to avoid unneccessary copies of the spectra matrix. In addition, frequent calls to gc () can be requested by hy.setOption (gc = TRUE). The same behaviour is triggered in read.ENVI and its derivatives (read.ENVI and read.ENVI.Nicolet). The memory consumption of scan.txt.Renishaw can be lowered by importing the data in chunks (argument nlines).

("hyperSpec"),
read.ENVI\*,
scan.txt.Renishaw

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#### References

- [1] Kristian Hovde Liland and BjÄ,rn-Helge Mevik. baseline: Baseline Correction of Spectra, 2015. URL http://CRAN.R-project.org/package=baseline. R package version 1.1-4.
- [2] Ron Wehrens and Bjørn-Helge Mevik. pls: Partial Least Squares Regression (PLSR) and Principal Component Regression (PCR), 2007. URL http://mevik.net/work/software/pls.html. R package version 2.1-0.
- [3] Dirk Eddelbuettel and Romain François. Rcpp: Seamless R and C++ integration. *Journal of Statistical Software*, 40(8):1-18, 2011. URL http://www.jstatsoft.org/v40/i08/.
- [4] Dirk Eddelbuettel. Seamless R and C++ Integration with Rcpp. Springer, New York, 2013. ISBN 978-1-4614-6867-7.
- [5] Oleg Sklyar, Duncan Murdoch, Mike Smith, Dirk Eddelbuettel, and Romain Francois. *inline: Inline C, C++, Fortran function calls from R*, 2013. URL http://CRAN.R-project.org/package=inline. R package version 0.3.13.
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#### A. Overview of the functions provided by hyperSpec

Function	Explanation	
Access parts of the object		
[	Select / extract / delete spectra, wavelength ranges or extra data	
[<-	Set parts of spectra or extra data	
	Select / extract / delete spectra, wavelength ranges or extra data, get the result as matrix or data.frame $$	
[[<-	Set parts of spectra matrix	
\$	extract a data column (including \$spc)	
\$< <b>-</b>	replace a data column (including \$spc)	
i2wl	convert spectra matrix column indices to wavelengths	
isample	get a random sample of the spectra as index vector	
labels	get column labels	
labels<-	set column labels	
logbook	logging the data treatment	
logentry	make a logbook entry	
rownames<-		
sample	generate random sample of the spectra	
seq.hyperSpec	sequence along the spectra, either as $hyperSpec$ object or index vector	
wl	extract the wavelengths	
wl<-	replace the wavelengths	
wl2i	convert wavelengths to spectra matrix column indices	

Function	Explanation
Maths	
% <b>*</b> %	matrix multiplication
Vectorization	
aggregate	
apply	
sweep	
Comparison	
all.equal	
Plotting	
alois.palette	another palette
levelplot	
map.identify	identify spectra in map plot
map.sel.poly	identify spectra in map plot: select polygon
mark.dendrogram	mark samples in helust dendrogram
matlab.dark.palette	darker version of matlab.palette
matlab.palette	palette resembling Matlab's jet colors
plot	main switchyard for plotting
plotc	intensity over one other dimension: calibration plots, time series, depth series, etc.
plotmap	false-colour intensity over two other dimensions: spectral images, maps, etc. (rectangular tesselation)
plotspc	spectra plots: intensity over wavelength
plotvoronoi	false-colour intensity over two other dimensions: spectral images, maps, etc. (Voronoi tesselation)
sel.poly	polygon selection in lattice plot
spc.identify	identify spectra and wavelengths in spectra plot
spc.label.default	helper for spc.identify
spc.label.wlonly	helper for spc.identify
spc.point.default	helper for spc.identify
spc.point.max	helper for spc.identify
spc.point.min	helper for spc.identify
spc.point.sqr	helper for spc.identify
stacked.offsets	calculate intensity axis offsets for stacked spectral plots
trellis.factor.key	modify list of levelplot arguments according to factor levels
Type conversion	
as.data.frame	
as.long.df	convert to a long-format data.frame.

Function	Explanation
as.matrix	
as.t.df	convert to a transposed data.frame (spectra in columns)
as.wide.df	convert to a wide-format data.frame with each wavelength one column
decomposition	re-import results of spectral matrix decomposition (or the like) into $\ensuremath{\textit{hyperSpec}}$ object
Combine/split	
bind	commom interface for rbind and cbind
cbind.hyperSpec	
collapse	combine objects by adding columns if necessary. See plyr::rbind.fill.
merge	combines spectral ranges. works if spectra are in only one of the data sets $% \left\{ 1,2,\ldots ,2,\ldots ,2,\ldots \right\}$
rbind.hyperSpec	bind objects by row, i.e. add wavelength ranges or extra data
split	
$Basic\ information$	
chk.hy	checks whether the object is a hyperSpec object
colnames	
colnames<-	
ncol	number of data columns (extra data plus spectra matrix)
nrow	number of spectra
nwl	number of data points per spectrum
print	summary information
rownames	
summary	summary information including the log
Create and initialize an object	
empty	creates an hyper Spec object with 0 rows, but the same wavelengths as another object
Options	
hy.getOption	get an option
hy.getOptions	get more options
hy.setOptions	set options
Tests	
hy.unittest	run all unit tests
Utility functions	
mean	mean spectrum
mean_pm_sd	mean $\pm$ one standard deviation of a vector
mean_sd	mean and standard deviation of a vector

Function	Explanation
pearson.dist	distance measure based on Pearson's $\mathbb{R}^2$
quantile	quantile spectra
rbind.fill.matrix	transitional until plyr::rbind.fill.matrix is out
WC	word count using wc if available on the system
$Spectra-specific\ transformations$	
orderwl	sort columns of spectra matrix according to the wavelengths
spc.bin	spectral binning
spc.fit.poly	least squres fit of a polynomial
spc.fit.poly.below	least squres fit of a polynomial with automatic support point determination
spc.loess	loess smoothing interpolation
File import/export	
read.ENVI	import ENVI file
read.ENVI.Nicolet	import ENVI files writen by Nicolet spectrometers
read.spc	import .spc file
read.spc.KaiserMap	import a Raman map saved by Kaiser Optical Systems' Hologram software as multiple .spc files $$
read.txt.long	import long-type ASCII file
read.txt.wide	imort wide-type ASCII file
scan.txt.Renishaw	import ASCII files produced by Renishaw (InVia) spectrometers
scan.txt.Witec	import ASCII files produced by Witec Raman spectrometers
scan.zip.Renishaw	directly read zip packed ASCII files produced by Renishaw spectrometers
write.txt.long	export as long-type ASCII file
write.txt.wide	export as wide-type ASCII file

#### **Session Info**

[,1] "Linux" sysname release "3.13.0-46-generic" "#75-Ubuntu SMP Tue Feb 10 15:24:04 UTC 2015" version "cbdesktop" nodename"x86\_64" machine "cb" login "cb" user effective\_user "cb" R version 3.1.2 (2014-10-31) Platform: x86\_64-pc-linux-gnu (64-bit) [1] LC\_CTYPE=de\_DE.UTF-8 LC\_NUMERIC=C LC\_TIME=de\_DE.UTF-8 [4] LC\_COLLATE=de\_DE.UTF-8 LC\_MONETARY=de\_DE.UTF-8 LC\_MESSAGES=de\_DE.UTF-8

LC\_NAME=C LC\_ADDRESS=C [7] LC\_PAPER=de\_DE.UTF-8

[10] LC\_TELEPHONE=C LC\_MEASUREMENT=de\_DE.UTF-8 LC\_IDENTIFICATION=C

attached base packages:

methods stats graphics grDevices utils datasets base [1] grid

other attached packages:

plotrix\_3.5-11 MASS\_7.3-39 ggplot2\_1.0.0 lattice\_0.20-[1] baseline\_1.1-4 hyperSpec\_0.98-20150303

[5] mvtnorm\_1.0-2 lattice\_0.20-30

loaded via a namespace (and not attached):

[1] colorspace\_1.2-4 digest\_0.6.8 gtable\_0.1.2 labeling\_0.3 munsell\_0.4.2 [6] plyr\_1.8.1 proto\_0.3-10 Rcpp\_0.11.4 reshape2\_1.4.1 scales\_0.2.4 [6] plyr\_1.8.1 proto\_0.3-10 Rcpp\_0.11.4
[11] SparseM\_1.6 stringr\_0.6.2 tools\_3.1.2