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

hyperSpec is a R package that allows convenient handling of (hyper)spectral data sets, i. e. data sets comprising spectra together with further data on a per-spectrum basis. The spectra can be anything that is recorded over a common discretized axis, the so-called wavelength axis. Throughout the documentation of the package, the terms intensity and wavelength refer to the spectral ordinate and abscissa, respectively.

However, *hyperSpec* works perfectly fine with any data that fits in that general scheme, so that the three terms may also be used for:

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

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

extra data: spatial information (spectral images, maps, or profiles), temporal information (kinetics,

time series), concentrations (calibration series), class membership information, etc.

Note that there is no restriction on the number of extra data columns.

This vignette gives an introduction on basic working techniques using the R package *hyperSpec*. It comes with five data sets,

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 (acetaminophen) ranging from 100 to 3200 cm⁻¹

with some overlapping wavelength ranges. It is used mainly in the *plotting* vignette.

barbituates 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 vignetted that show exemplary work flows for the respective data type.

This document describes how to accomplish spectroscopic tasks. 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.

After some remarks on the notation used in the document and on the general behaviour of R, sections 3 shows how to load the package. Section 4 describes how *hyperSpec* objects are organized internally.

A list of all functions available in hyperSpec is given in appendix A (23)

1.1. Notation

This vignette demonstrates working techniques mostly from a spectroscopic point of view: rather than going through the functions provided by *hyperSpec*, it is organized more closely on spectroscopic tasks. However, the functions discussed are printed on the margin for a fast overview.

In R, slots of a S4 class can be accessed directly by the @ operator. In this vignette, the notation @xxx will thus mean "slot xxx of an object" see figure 1 on page 5).

Likewise, named elements of a *list*, like the 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" see figure 1 on page 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 table ?? on page ??.

2.2. Functionality Can be Extended at Runtime

The concept of functions in R 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.

This offers the opportunity of easily writing specialized functions that are adapted to specific tasks.

For examples, see the setup of the lattice plotting functions in 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.

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

```
> .is.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.

3. Loading the package

```
To load hyperSpec, use
```

> library (hyperSpec)

4. The structure of hyperSpec objects

hyperSpec is a S4 (or new-style) class. It has four so-called slots that hold the data:

Owavelength 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)

@log a data.frame keeping track of what is done with the object

However, it is good practice to use the functions provided by *hyperSpec* to handle the objects rather than accessing the slots directly (tab). This also ensures that proper (valid) objects are retained.

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

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

Table 1: Get and set functions for the slots of hyperSpec objects

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

illustrated in figure 1. Even if there are no spectra, \$spc must still be present but it can contain 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. 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 figure 2 on page 12). To get familiar with expressions for axis annotation, see? plotmath and demo (plotmath).

5. Functions provided by hyperSpec

Table A (p. 23) in the appendix gives an overview of the functions implemented by hyperSpec.

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:

```
> 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/(mu * m) [numeric] rng -4.77 -3.77 ... 19.23
   2. x: x/(mu * m) [numeric] rng -11.55 -10.55 ... 22.45
   3. spc: I / a.u. [matrix300] rng 80.04420 81.75761 ... 1858.881
   4. clusters: clusters [factor] rng cell lacuna matrix + NA
> 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/(mu * m) [numeric] rng -4.77 -3.77 ... 19.23
2. x: x/(mu * m) [numeric] rng -11.55 -10.55 ... 22.45
   3. spc: I / a.u. [matrix300] rng 80.04420 81.75761 ... 1858.881
   4. clusters: clusters [factor] rng cell lacuna matrix + NA
log:
                      short
                                   long
                                                            date
                                                                      user
   1
        scan.txt.Renishaw
                              list(...
                                           2010-02-24 10:04:06
                                                                    cb@cb
   2
                   orderwl
                              list(...
                                           2010-02-24 10:04:06
                                                                    cb@cb
                                           2010-02-24 10:04:29
   3
                 spc.loess
                              list(...
                                                                    cb@cb
   4
                        $<-
                              list(...
                                           2010-02-24 10:04:47
                                                                    cb@cb
   5
                        $<-
                              list(...
                                           2010-02-24 10:04:47
                                                                     cb@cb
```

The data set chondro consists of 875 spectra with 300 data points each, and 4 data columns: two for the spatial information plus \$spc. These informations 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

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).

7. Creating a hyperSpec Object, Data Import and Export

hyperSpec comes with filters for a variety of file formats. ASCII files are supported for import and export, Matlab files can be read and written with the help of package R.matlab. ENVI files can be read (for writing, see e.g. package caTools). Also reading of Thermo Galactic's .spc file format is supported.

E.g. to read a spectrum saved in Renishaw's ASCII file format, use:

```
> paracetamol <- scan.txt.Renishaw ("rawdata/paracetamol.txt", "spc")
> paracetamol
```

```
hyperSpec object
  1 spectra
  1 data columns
  4064 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 96.7865 98.1432 ... 3200.07
data: (1 rows x 1 columns)
  1. spc: I / a.u. [AsIs matrix x 4064] rng 299.229 317.041 ... 49052.2
> save (paracetamol, file = "paracetamol.rda")
```

Vignette FileIO discusses the available filters and also how to derive own filters for importing manufacturer specific data.

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)

You will usually give the following arguments:

spc the spectra matrix

wavelength the wavelength axis vector

data the extra data (possibly already including the spectra matrix in column spc)

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

8. Combining and Decomposing hyperspec Objects

8.1. Binding Objects together

cbind rbind

hyperspec Objects can be bound together, either by rows to append a new spectral range or by columns to append new spectra

```
> cbind (chondro [, , 600 \tilde{} 800], chondro [, , 1600 \tilde{} 1800])
hyperSpec object
   875 spectra
   4 data columns
   101 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (875 rows x 4 columns)
   1. y: y/(mu * m) [numeric] rng -4.77 -3.77 \dots 19.23
   2. x: x/(mu * m) [numeric] rng -11.55 -10.55 ... 22.45
   3. spc: I / a.u. [matrix101] rng 80.04420 81.75761 ... 1541.625
   4. clusters: clusters [factor] rng cell lacuna matrix + NA
> rbind (chondro [, , 600 ~ 800], chondro [, , 600 ~ 800])
hyperSpec object
   1750 spectra
   4 data columns
   50 data points / spectrum
```

```
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 798
data: (1750 rows x 4 columns)
   1. y: y/(mu * m) [numeric] rng -4.77 -3.77 ... 19.23
   2. x: x/(mu * m) [numeric] rng -11.55 -10.55 ... 22.45
   3. spc: I / a.u. [matrix50] rng 195.5281 212.0432 ... 729.5765
   4. clusters: clusters [factor] rng cell lacuna matrix + NA
```

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

collapse

8.2. Binding Objects that do not share the same extra data and/or wavelength axis

%*%

8.3. Matrix Multiplication

Two hyperSpec objects can be matrix multiplied by **. For an example, see the principal component analysis below (section 12.1 on page 19).

decomposition

8.4. 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. 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 12.1) on page 19.

9. Access to the data

9.1. Selecting and Deleting Spectra

The extraction function [] (or [[]], if the spectra matrix or the data.frame is needed rather than a hyperSpec object) takes the spectra as first argument (For detailed help: ? "["). 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

```
> flu [1 : 3]
hyperSpec object
   3 spectra
   3 data columns
   181 data points / spectrum
wavelength: lambda/nm [numeric] 405.0 405.5 ... 495
data: (3 rows x 3 columns)
   1. file: [factor] rng rawdata/flu1.txt rawdata/flu2.txt rawdata/flu3.txt
   2. spc: I[fl] / a.u. [AsIs matrix x 181] rng 27.15000 32.34467 ... 336.5057
   3. c: c / (mg / 1) [numeric] rng 0.05 0.10 0.15
```

```
> flu [-3]
hyperSpec object
   5 spectra
   3 data columns
   181 data points / spectrum
wavelength: lambda/nm [numeric] 405.0 405.5 ... 495
data: (5 rows x 3 columns)
   1. file: [factor] rng rawdata/flu1.txt rawdata/flu2.txt rawdata/flu4.txt rawdata/flu5.txt rawdata/flu6.tx
   2. spc: I[f1] / a.u. [AsIs matrix x 181] rng 27.15000 32.34467 ... 677.4947
   3. c: c / (mg / 1) [numeric] rng 0.05 0.10 0.20 0.25 0.30
> chondro [chondro$y > 10]
hyperSpec object
   350 spectra
   4 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (350 rows x 4 columns)
   1. y: y/(mu * m) [numeric] rng 10.23 11.23 ... 19.23
   2. x: x/(mu * m) [numeric] rng -11.55 -10.55 ... 22.45
   3. spc: I / a.u. [matrix300] rng 88.98556 89.99474 ... 1745.724
   4. clusters: clusters [factor] rng cell lacuna matrix
```

9.2. Accessing the Extra Data

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"
                          "spc"
                                     "clusters"
> chondro [[1 : 3, 1]]
1 -4.77
2 -4.77
3 -4.77
> chondro [[1 : 3, -3]]
            x clusters
1 -4.77 -11.55
                matrix
2 -4.77 -10.55
3 -4.77 -9.55
> chondro [[1 : 3, "x"]]
1 -11.55
2 -10.55
3 -9.55
> chondro [[1 : 3, c (TRUE, FALSE, FALSE)]]
```

```
y clusters
1 -4.77 matrix
2 -4.77 matrix
3 -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
```

The extra data may also be set this way:

```
> flu$n <- list (1 : 6, label = "sample no.")
```

This function will append new columns, if necessary.

9.3. Wavelengths and Spectral Axis

9.3.1. Wavelength Indices

w12i i2w1

Spectra in hyperSpec have always discretized wavelength axes, they are stored in a matrix with column corresponding to one wavelength. hyperSpec provides two conversion functions: i2wl returns the wavelength corresponding to the given indices and wl2i calculates index vectors from wavelengths.

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

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^{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 doesn't work. In that (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 vectors¹:

```
> wl2i (flu, 450 - 2i ~ 450 + 2i)
[1] 89 90 91 92 93
> wl2i (flu, c (min ~ 406.5, max - 2i ~ max))
[1] 1 2 3 4 179 180 181
```

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

9.3.2. Wavelength Axis Conversion

9.4. Fast Access to Parts of the hyperSpec Object

[[]] \$. \$..

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

- x [[]] returns the spectra matrix (x\$spc).
- x [[i, , 1]] the cut spectra matrix is returned if wavelengths are specified in l.
- x [[i, j, 1]] If data columns are selected (second index), the result is a data frame.
- x [[i, , 1]] <- Also, parts of the spectra matrix can be set (only indices for spectra and wavelength are allowed for this function).
- x [i, j] <- 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). However, the columns must match exactly in this case.

¹Formulas are combined to a list by c.

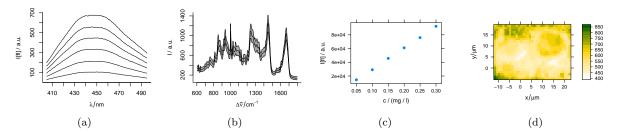


Figure 2: Some example plots. (a)plotspc (flu), (b)plot (chondro, "spcmeansd"), (c)plotc (flu), and (d)plotmap (chondro, col.regions = YG (20)) (using a yellow-green palette).

10. Plotting

hyperSpec comes with three predefined plotting functions.

plotspc plots the spectra, i.e. the intensities \$spc over the wavelengths @wavelength.

plotmap plots a false colour map: a single value (e.g. average intensity or cluster membership) over two data columns (default \$x and \$y).

plots a time series or calibration plot: e.g. an intensity over a single other data column (like concentration, depth, or time).

All three plus some more handy abbreviations are also accessible via plot:

plot

plot (flu, "spc") is equivalent to plotspc (flu)

plot (chondro, "spcmeansd") plots mean spectrum ± 1 standard deviation

plot (chondro, "spcprctl") plots median, 16th and 84th percentile. This is similar to "spcmeansd". Spectroscopic data frequently are not Gaussian distributed. The percentiles give a better idea of the true distribution. They are also less sensitive to outliers.

plot (chondro, "spcprct15") like "spcprct1" plus $5^{\rm th}$ and $95^{\rm th}$ percentile.

plot (chondro, "map") is equivalent to plotmap (chondro)

plot (flu, "c") is equivalent to plotc (flu)

plot (laser, "ts") plots a time series plot, equivalent to plotc (laser, use.c = "t")

plot (x, "depth") plots a depth profile plot, equivalent to plotc (laser, use.c = "z")

Figure 2 shows some example plots.

plot uses its second argument to determine which of the three specialized plot functions to call. All further arguments are handed over to this function.

10.1. Plotting Spectra

plotspc

plotspc offers a variety of parameters for customized plots. To plot ...

with reversed abscissa use wl.reverse = TRUE

in different colours colours use col = vector.of.colours

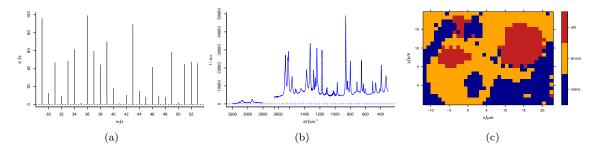


Figure 3: Arguments to plotspc. (a) plot (fake.mass.spec, lines.args = list (type = "h")) (b) plotspc (paracetamol, wl.range = c (300 ~ 1800, 2800 ~ max), xoffset = 850, wl.reverse = TRUE) (c) plotmap with a factor, see section 12.2.

dots instead of lines use lines.args = list (pch = 20, type = "p")

```
mass spectra use lines.args = list (type = "h")

particular wavelength ranges use wl.range = list (600 ~ 1800, 2800 ~ 3100)

If wl.range already contains indices: use wl.index = TRUE

Cut the wavelength axis appropriately with xoffset = 800

stacked spectra use stacked

more spectra into an existing plot use add = TRUE

with different line at I=0 use zeroline = list.of.arguments.to.abline. NULL suppresses the
```

10.2. Plotting the Spectra Matrix

It is often useful to plot the spectra against an additional coordinate, e.g. the time for time series, the depth for depth profiles, etc.

This is done by levelplot, or by using plot (object, "mat", model = spc .wavelength * other.data.column). The actual plotting is done by levelplot, so the plots can be grouped or conditioned. See figure 4

```
> levelplot (spc ~ .wavelength * t, laser, col.regions = YG (20))
> print (plot (flu, "mat", contour = TRUE, labels = TRUE, col = "#00000080"))
```

10.3. Calibration Plots, (Depth) Profiles, and Time Series Plots

plotc

plots plots an intensity over one of the extra data columns. The abscissa uses column c by default, another column can be specified using use.c = name. The ordinate can be calculated as a sum characteristic (with parameter func = function, defaulting to sum). If parameter z is given, these values are used instead. z may be the name of an extra data column, or a numeric that should be used directly.

To customize the plot, give any arguments that you would usually supply to plot as a list using argument plot.args.

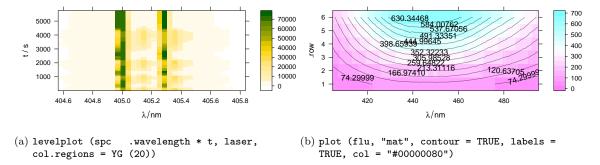


Figure 4: plotting the spectra matrix.

10.4. Plotting False-Colour Maps

plotmap

plotmap uses levelplot, a *lattice* function. Therefore, in loops, functions, Sweave chunks, etc. the lattice object needs to be printed explicitly by print (plotmap (object)) (R FAQ: Why do lattice/trellis graphics not work?).

plotmap produces a 3d plot, with the z axis colour-coded. plotmap's arguments x and y take the name of extra data columns.

The colour-coded axis. Also z can be used to select one column of the extra data by name. Alternatively, it may be a numeric or factor directly giving the values to be used. Each level of a factor will have one colour. It is also possible to plot a sum characteristic of the spectra: supply the function in argument func. The default setting is to plot the average intensity (no z and func=mean).

To plot with a different palette, use trellis.args= list (col.regions = palette).

Conditioning. Lattice graphics have a concept of conditioning a plot. Instead of plotting all data in one diagram, a diagram is produced for each of the groups specified by the condition. plotmap's argument *cond* takes he name of the extra data column used for conditioning. This could e.g. be a column containing the sample number of a *hyperSpec* object that contains several samples.

11. Spectral (Pre)processing

11.1. 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 data columns were specified, too) only.

> flu [,, min ~ 408.5]

```
hyperSpec object
   6 spectra
   4 data columns
   8 data points / spectrum
wavelength: lambda/nm [numeric] 405.0 405.5 ... 408.5
data: (6 rows x 4 columns)
   1. file: [factor] rng rawdata/flu1.txt rawdata/flu2.txt ... rawdata/flu6.txt
   2. spc: I[f1] / a.u. [AsIs matrix x 8] rng 27.15000 32.34467 ... 256.8913
   3. c: c / (mg / 1) [numeric] rng 0.05 0.10 ... 0.3
   4. n: sample no. [integer] rng 1 2 ... 6
> flu [[,, c (min ~ min + 2i, max - 2i ~ max)]]
           405
                   405.5
                               406
                                         494
                                                 494.5
                                                            495
                                   47.16267
[1,] 27.15000 32.34467 33.37867
                                             46.41233
                                                       45.25633
[2,] 66.80133 63.71533 66.71200 96.60167
                                             96.20600
                                                       94.61033
[3,] 93.14433 103.06767 106.19367 149.53900 148.52667 145.79333
[4,] 130.66367 139.99833 143.79767 201.48433 198.86733 195.86733
[5,] 167.26667 171.89833 177.47067 252.06567 248.06700 246.95200
[6,] 198.43033 209.45800 215.78500 307.51850 302.32550 294.64950
```

11.2. Spectral Interpolation and Smoothing

spc.bin spc.loess

Frequently, a hyperSpec object needs to be interpolated onto a new wavelength axis. e.g. because measurements resulted in slightly shifted wavelength axes. Or data from a grating spectrometer with unequal data point spacing should be interpolated onto an evenly spaced wavelength axis. 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 change the wavelength axis of hyperSpec objects: 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)
```

spc.loess applies R's loess function for spectral interpolation. Figure 5 shows the result of interpolating from 300 to 1800 and 2850 to 3150 cm⁻¹ with 2 cm⁻¹ data point distance. This corresponds to a spectral resolution of about 4 cm⁻¹, 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⁻¹). The interpolated spectrum has 902 data points.

11.3. Background Correction

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

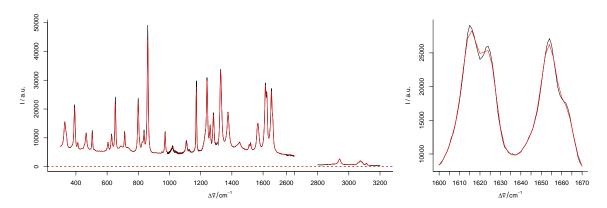


Figure 5: Smoothing interpolation by spc.loess with new data point spacing of 2 cm⁻¹. The magnification on the right shows how interpolation may cause a loss in signal.

11.4. Offset Correction

Calculate the offsets and sweep them off the spectra:

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

11.5. Baseline Correction

hyperSpec comes with two functions to fit polynomial baselines.

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).

11.6. Intensity Calibration

11.6.1. 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
```

11.6.2. Correcting Wavelength Dependence

This means that 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, "/")

11.6.3. Spectra Dependent Correction

If the correction depends on the spectra (e.g. due to inhomogeneous illumination while collecting imaging data²), the MARGIN of the sweep function needs to be 1:

```
    Pixel dependent offsets are subtracted:
sweep (spectra, 2, pixel.offsets, "-")
```

2. A multiplicative dependency: sweep (spectra, 2, illumination.factors, "*")

11.7. Normalization

1. Calculate appropriate normalization factors:

factors <- 1 / apply (spectra, 1, sum) for area normalization. mean gives equal results, just that the Intensities are on the same scale as before.

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

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, "*")</p>
```

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

11.8. Centering the Data

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.

However, from a spectroscopic point of view it depends on the particular data set whether centering does make sense or not.

It is perfectly fine to centre the flu data set: the interpretation is that centering the data cancels the offset (background spectrum etc.) of the calibration:

 $^{^2}$ imaging (as opposed to mapping) refers to simultaneously collecting spatially resolved spectra, either 2d images or line imaging.

```
> flu.centered <- sweep (flu, 2, apply (flu, 2, mean), "-")
> 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 is 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 lots of the negative noise. In order to avoid that, using e. g. the 5th percentile spectrum is more suitable:

```
> chondro <- sweep (chondro, 2, apply (chondro, 2, quantile, 0.05), "-")
> plot (chondro, "spcprctl5")
```

11.9. Variance Scaling

Variance scaling is often used in multivariate analysis to adjust the influence and scaling of the variates (that are typically different physical values). However, it is hardly appropriate for spectra that do have the same scale of the same physical value.

11.10. Multiplicative Scatter Correction (MSC)

MSC can be done using msc from package pls[1]. It operates on the spectra matrix:

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

11.11. Spectral Arithmetic

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"
```

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.

12. Data Analysis

12.1. 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)
```

Results of such a decomposition can be put again into *hyperSpec* objects. This allows to plot e.g. the loading like spectra, or score maps, see figure 6.

The loadings can be similarly re-imported:

There is, however, one important difference. The loadings are thought of as values computed from all spectra togehter. 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:

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

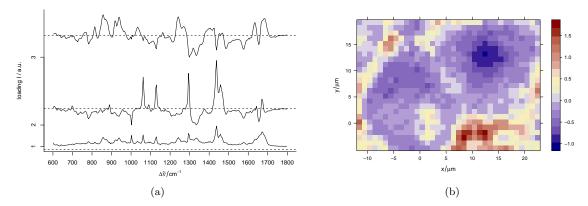


Figure 6: (a) The first three loadings: plot (loadings [1 : 3], stacked = TRUE). (b) The second score map: plotmap (scores [, , 2])

y x clusters 1 NA NA <NA>

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

12.1.1. 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.

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 ϵ .

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:

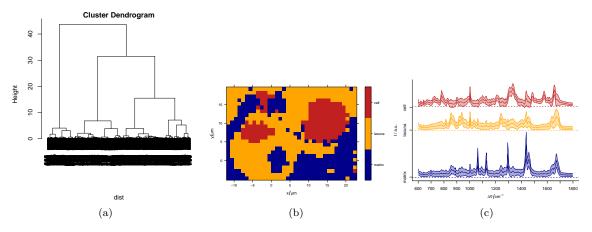


Figure 7: The results of the cluster analysis: (a) the dendrogram (b) the map of the 3 clusters (c) the mean spectra.

```
> 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 875×300 spectra matrix was reduced to 10.

12.2. Data Analysis Methods using a matrix e.g. Hierarchical Cluster Analysis

```
> 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 dendrogam 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 7b):

12.3. 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.

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")
```

12.4. Splitting an Object

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

```
> clusters <- split (chondro, chondro$clusters)
> clusters
$matrix
hyperSpec object
   292 spectra
   5 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (292 rows x 5 columns)
   1. y: y/(mu * m) [numeric] rng -4.77 -3.77 \dots 19.23
   2. x: x/(mu * m) [numeric] rng -11.55 -10.55 ... 22.45
   3. spc: I / a.u. [matrix300] rng -0.1372477 -0.1306328 ... 0.9382884
   4. clusters: clusters [factor] rng matrix
   5. measurement: measurement [numeric] rng 1
$lacuna
hyperSpec object
   417 spectra
   5 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (417 rows x 5 columns)
   1. y: y/(mu * m) [numeric] rng -4.77 -3.77 ... 19.23
   2. x: x/(mu * m) [numeric] rng -11.55 -10.55 ... 22.45
   3. spc: I / a.u. [matrix300] rng -0.2540939 -0.2373205 ... 1.043128
   4. clusters: clusters [factor] rng lacuna
   5. measurement: measurement [numeric] rng 1
$cell
hyperSpec object
   166 spectra
   5 data columns
   300 data points / spectrum
wavelength: Delta * tilde(nu)/cm^-1 [numeric] 602 606 ... 1798
data: (166 rows x 5 columns)
   1. y: y/(mu * m) [numeric] rng -2.77 5.23 ... 18.23
   2. x: x/(mu * m) [numeric] rng -7.55 -6.55 ... 22.45
   3. spc: I / a.u. [matrix300] rng -0.2713962 -0.2156510 ... 0.4380194
   4. clusters: clusters [factor] rng cell
   5. measurement: measurement [numeric] rng
```

Splitting can be reversed by rbind (see section 8.1, page 7).

References

[1] 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.

A. Overview of the functions provided by hyperSpec

Function	Explanation
Create and initialize an object	
initialize	
Basic information	
colnames	
colnames<-	
dim	
dimnames	
length	
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	
show	
summary	summary information including the log
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)
\$< -	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
Type conversion	
as.character	
as.data.frame	
as.long.df	convert to a long-format data.frame.
as.matrix	
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
File import/export	
\emph{R.matlab::readMat}	import matlab file
\emph{R.matlab::writeMat}	export as matlab file
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
write.txt.long	export as long-type ASCII file
write.txt.wide	export as wide-type ASCII file
Combine/split	
bind	commom interface for rbind and cbind
cbind2	bind two hyperSpec objects by column
cbind.hyperSpec	
collapse	combine objects by adding columns if necessary. See plyr::rbind.fill.
rbind2	bind two $hyperSpec$ objects by row, i. e. add wavelength ranges or extra data
rbind.hyperSpec	bind objects by row, i.e. add wavelength ranges or extra data
split	
Vectorization	
aggregate	
apply	
sweep	
Maths	
% * %	matrix multiplication

Function	Explanation
Arith	
log	
Math	mathematical functions. See (help (■Math■))
Math2	rounding
Summary	summary measures such as min, max, etc.
Comparison	
all.equal	
Compare	> < == >= <= return a logical matrix
is.na	
Plotting	
levelplot	
map.identify	identify spectra in map plot
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. $$
plotmap	false-colour intensity over two other dimensions: spectral images, maps, etc. (rectangular tesselation)
plotspc	spectra plots: intensity over wavelength
plotvoronoi	false-colour intensity over two other dimensions: spectral images, maps, etc. (Voronoi tesselation) $$
spc.identify	identify spectra and wavelengths in spectra plot
stacked.offsets	calculate intensity axis offsets for stacked spectral plots
trellis.factor.key	modify list of levelplot arguments according to factor levels
$Spectra-specific\ transformations$	
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 sq ares fit of a polynomial with automatic support point determination $% \left(1\right) =\left(1\right) \left(1\right) \left($
spc.loess	loess smoothing interpolation
Utility functions	
array2df	convert array into a matrix or data.frame
array2vec	convert array indices (n element vector) into vector indices
factor2num	convert a factor with numeric levels into the numeric
mean_pm_sd	mean \pm one standard deviation of a vector

Function	Explanation
mean_sd	mean and standard deviation of a vector
pearson.dist	distance measure based on Pearson's \mathbb{R}^2
rbind.fill	transitional patch of plyr::rbind.fill working with matrices
rbind.fill.matrix	transitional until plyr::rbind.fill.matrix is out
vec2array	convert vector (one element) index into an array into an \boldsymbol{n} element array index
WC	word count using wc if available on the system