

Fitting Polynomial Baselines to Spectra

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Introduction

This document discusses baseline correction methods that can be used with *hyperSpec*. *hyperSpec* provides two fitting functions for polynomial baselines, `spc.fit.poly` and `spc.fit.poly.below`.

In contrast to many other programs that provide baseline correction methods, these functions do least squares fits. However, the baselines can be forced through particular points, if this behaviour is needed.

The main difference between the two functions is that `spc.fit.poly` returns a least squares fit through the complete spectrum that is given in *fit.to* whereas `spc.fit.poly.below` tries to find appropriate spectral regions to fit the baseline to.

Syntax & parameters

```
spc.fit.poly (fit.to, apply.to = fit.to, poly.order = 1  
             short = NULL, user = NULL, date = NULL)
```

```
spc.fit.poly.below (fit.to, apply.to = fit.to, poly.order = 1, npts.min = NULL,  
                   noise = 0, short = NULL, user = NULL, date = NULL)
```

fit.to: *hyperSpec* object with the spectra whose baselines are to be fitted.

apply.to: *hyperSpec* object giving the spectral range, on which the baselines should be evaluated.
If **apply** is `NULL`, a *hyperSpec* object with the polynomial coefficients is returned instead of evaluated baselines.

poly.order: polynomial order of the baselines

npts.min: minimal number of data points per spectrum to be used for the fit.
npts.min defaults to the larger of 3 times (**poly.order** + 1) or $\frac{1}{20th}$ of the number of data points per spectrum.
If **npts.min** \leq **poly.order**, a warning is issued and **npts.min** \leftarrow **poly.order** + 1 is used.

noise: a vector giving the amount of noise, see below.

short, user, date: are handed to `logentry`

General Use

Both functions fit the polynomial to the spectral range given in *hyperSpec* object *fit.to*. If *apply.to* is not NULL, the polynomials are evaluated on the spectral range of *apply.to*. Otherwise, the polynomial coefficients are returned.

Subtracting the baseline is up to the user, it is easily done as *hyperSpec* provides the - (minus) operator.

Fitting polynomial baselines using least squares

Commonly, baselines are fit using (single) support points that are specified by the user. Also, usually $n + 1$ support point is used for a polynomial of order n . This approach is appropriate for spectra with high signal to noise ratio.

Such a baseline can be obtained by restricting the spectra in *fit.to* to the respective points (see figure 1):

```
> bl <- spc.fit.poly (chondro [1,, c(630, 1775)], chondro [1])
> plot (chondro [1], plot.args = list (ylim = c(250, 750)))
> plot (chondro [1,, c(630, 1775)], add = TRUE, lines.args = list (type = "p", pch = 20))
> plot (bl, add = TRUE)
```

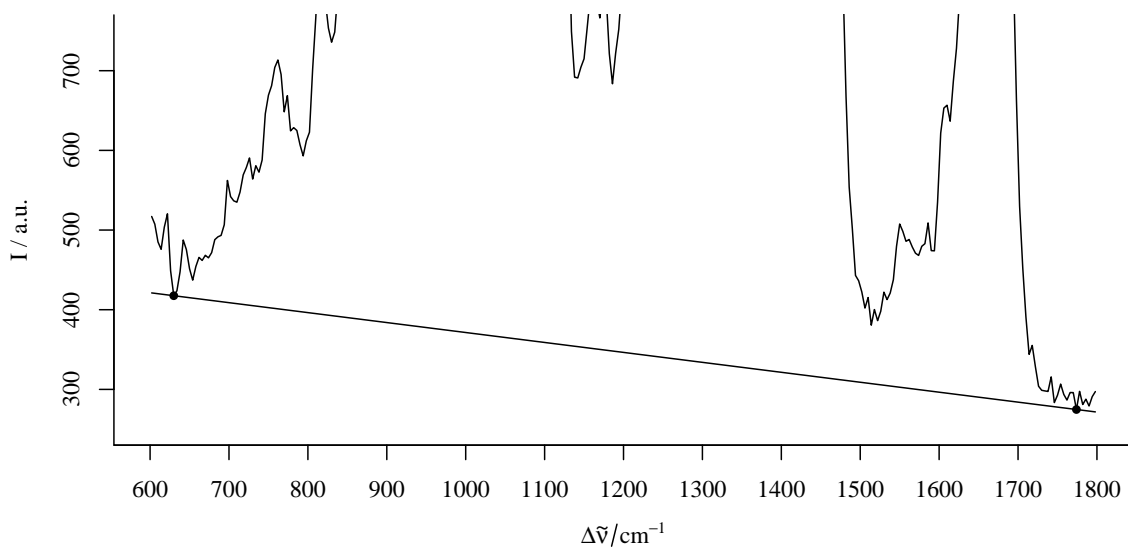


Figure 1: Fitting a linear baseline through two points.

However, if the signal to noise ratio is not ideal, a polynomial with $n + 1$ supporting points (i.e. with zero degrees of freedom) is subject to a considerable amount of noise. If on the other hand, more data points consisting of baseline only are available, the uncertainty on the polynomial can be reduced by a least squares fit.

Both *spc.fit.poly* and *spc.fit.poly.below* therefore provide least squares fitting for the polynomial.

As `spc.fit.poly` fits to the whole spectral region of `fit.to`. Thus, the spectra need to be cut to appropriate wavelength ranges that do not contain any signal.

In order to speed up calculations, the least squares fit is done by using the Vandermonde matrix and solving the equation system by `qr.solve`.

This fit is not weighted. A spectral region with many data points therefore has greater influence on the resulting baseline than a region with just a few data points. It is up to the user to decide whether this should be corrected for by selecting appropriate numbers of data points (e.g. by using replicates of the shorter spectral region).

The mechanism of automatically fitting the baseline in `spc.fit.poly.below`

`spc.fit.poly.below` tries to automatically find appropriate spectral regions for baseline fitting. This is done by excluding spectral regions that contain signals from the baseline fitting. The idea is that all data points that lie above a fitted polynomial (through the whole spectrum) will be treated as signal and thus be excluded from the baseline fitting.

The supporting points for the baseline polynomials are therefore calculated iteratively:

1. A polynomial of the requested order is fit to the considered spectral range, initially to the whole spectrum given in `fit.to`
2. Only the parts of the spectrum that lie below this polynomial plus the `noise` are retained as supporting points for the next iteration.

These two steps are repeated until either

- no further points are excluded, or
- the next polynomial would have less than `npts.min` supporting points.

The baselines and respective supporting points for each iteration of `spc.fit.poly.below` (`chondro [1]`, `poly.order = 1`) are shown in figure 2.

Specifying the spectral range

It is possible to exclude spectral regions that do not contribute to the baseline from the fitting, while the baseline is used for the whole spectrum. The choice of the spectral range in `fit.to` influences the resulting baselines to a certain extent, as becomes clear from figure 3. Narrowing the spectral range in `fit.to` speeds up the fitting of the polynomials:

```
> system.time (spc.fit.poly.below (chondro[], NULL, npts.min = 20))
   user  system elapsed 
 2.060   0.010   2.071 

> system.time (spc.fit.poly.below (chondro [, c (min ~ 700, 1700 ~ max)], NULL, npts.min = 20))
   user  system elapsed 
 1.060   0.000   1.058
```

Fitting polynomials of different orders

Figure 4 shows the resulting baseline polynomial of `spc.fit.poly.below` (`chondro [1]`, `poly.order = order`) with `order = 0` to 3 for the first spectrum of the `chondro` data set.

The noise level

Besides defining a minimal number of supporting points, a “noise level” may be given. Consider a spectral range consisting only of noise. The upper part of figure 5 illustrates the problem. As the baseline fitting algorithm cannot distinguish between noise and real bands appearing above the fitted polynomial, the resulting baseline (black) is too low if the **noise** parameter is not given.

Setting the noise level to 10 (2 standard deviations), the fitting converges immediately with a much better result. The resulting baselines for `spc.fit.poly.below (chondro [1], poly.order = 1, noise = 12)` of the whole spectrum are shown in the middle and lower part of figure 5

noise may be a single value for all spectra, or a vector with the noise level for each of the spectra separately.

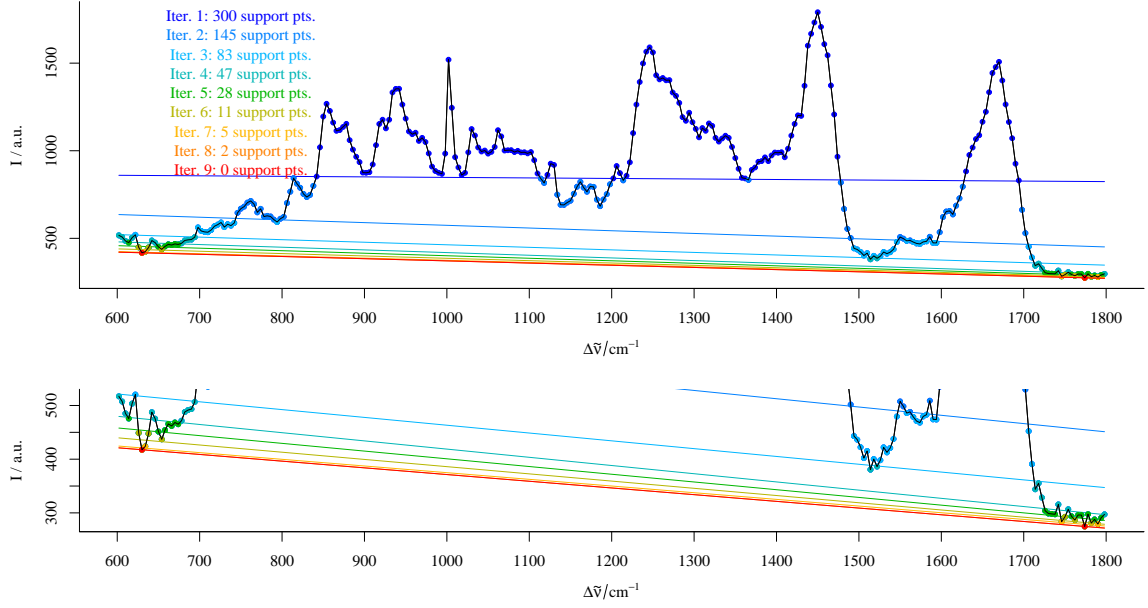


Figure 2: Iterative fitting of the baseline. The dots give the supporting points for the baselines in the same colour. The lower part is a magnification of the intensity axis.

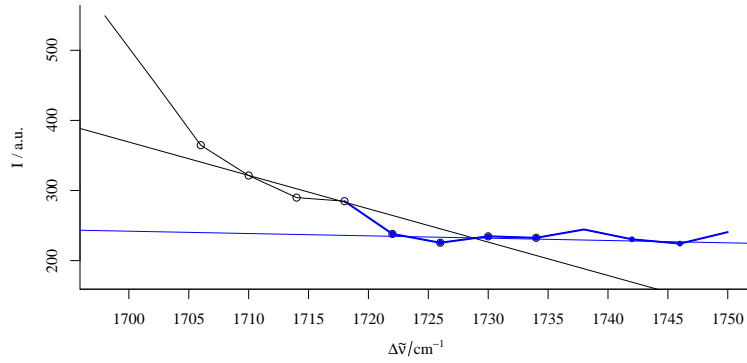


Figure 3: Influence of `fit.to` on the baseline polynomial. The black baseline is fit to the spectral range 1700 – 1800 cm^{-1} , the blue to 1715 – 1800 cm^{-1} only (dots & circles: supporting points).

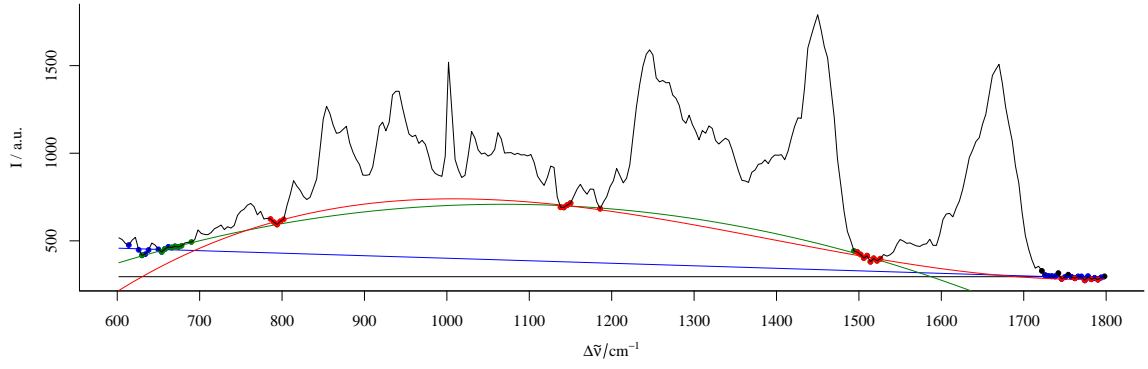


Figure 4: Baseline polynomial fit to the first spectrum of the chondro data set of order 0 (black), 1 (blue), 2 (green), and 3 (red). The dots indicate the points used for the fitting of the polynomial.

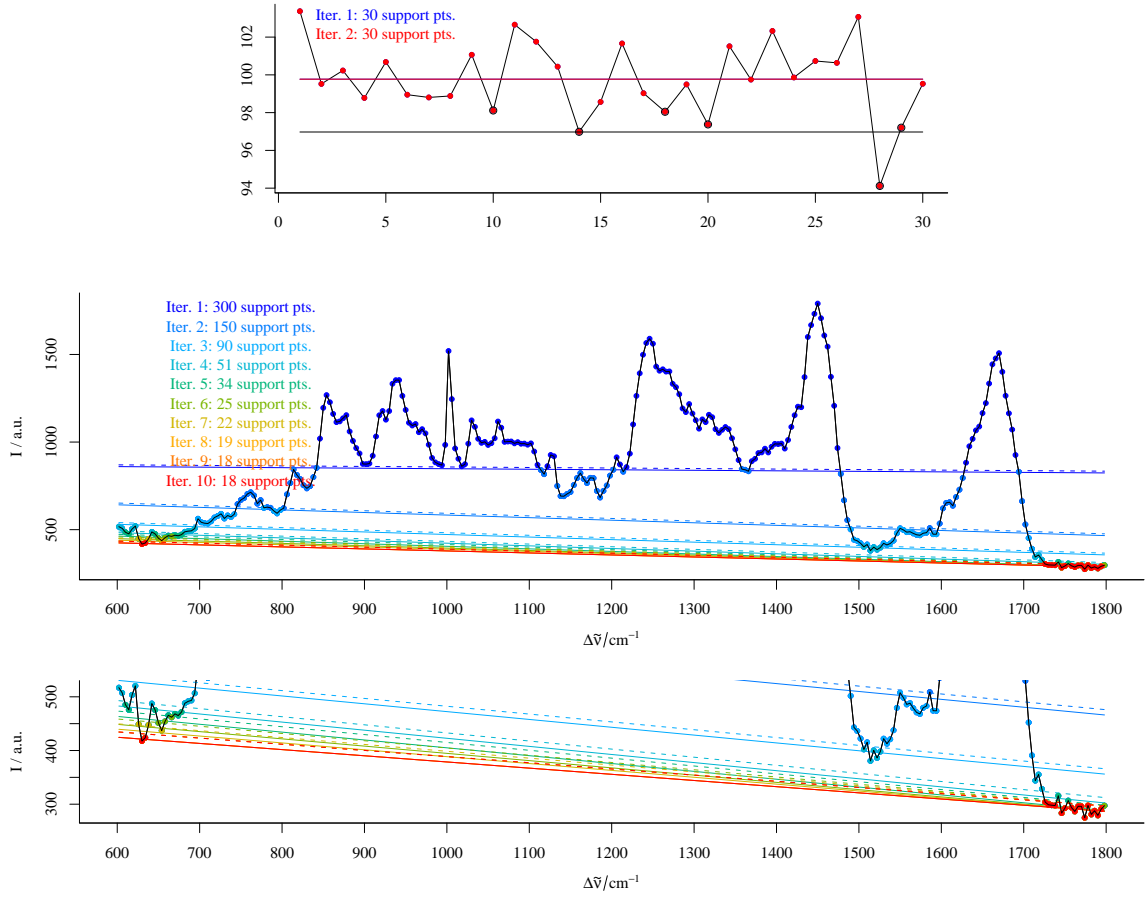


Figure 5: Iterative fitting of the baseline with noise level. Upper part: effects of the noise parameter on the baseline of a spectrum consisting only of noise and offset: without giving **noise** the resulting baseline (black) is clearly too low. A noise level of 10 results in the red baseline. The middle and lower part show the baseline fitting with noise level on the complete spectrum. Colour: iterations, dots/circles: supporting points for the respective baselines. Dashed: baseline plus noise. All points above this line are excluded from the next iteration.