## Basic calibration functions for analytical chemistry

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May 12, 2006

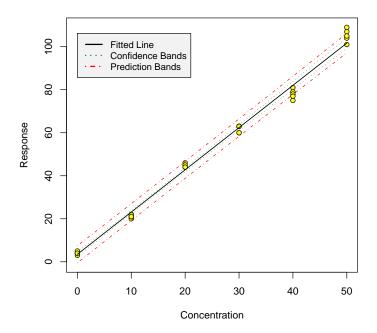
The chemCal package was first designed in the course of a lecture and lab course on "analytics of organic trace contaminants" at the University of Bremen from October to December 2004. In the fall 2005, an email exchange with Ron Wehrens led to the belief that it could be heavily improved if the inverse prediction method given in [1] would be implemented, since it also covers the case of weighted regression.

At the moment, the package only consists of two functions, working on univariate linear models of class lm or rlm.

When calibrating an analytical method, the first task is to generate a suitable model. If we want to use the chemCal functions, we will have to restrict ourselves to univariate, possibly weighted, linear regression so far.

Once such a model has been created, the calibration can be graphically shown by using the calplot function:

```
> library(chemCal)
> data(massart97ex3)
> attach(massart97ex3)
> yx <- split(y, factor(x))
> ybar <- sapply(yx, mean)
> s <- round(sapply(yx, sd), digits = 2)
> w <- round(1/(s^2), digits = 3)
> weights <- w[factor(x)]
> m <- lm(y ~ x, w = weights)
> calplot(m)
```



This is a reproduction of Example 8 in [1]. We can see the influence of the weighted regression on the confidence and prediction bands of the calibration.

If we now want to predict a new x value from measured y values, we use the inverse.predict function:

```
> inverse.predict(m, 15, ws = 1.67)
```

\$Prediction

[1] 5.865367

\$`Standard Error`

[1] 0.892611

\$Confidence

[1] 2.478285

\$`Confidence Limits`

[1] 3.387082 8.343652

The weight **ws** assigned to the measured y value has to be given by the user in the case of weighted regression. By default, the mean of the weights used in the linear regression is used.

## Theory

Equation 8.28 in [1] gives a general equation for predicting the standard error  $s_{\hat{x_s}}$  for an x value predicted from measurements of y according to the linear calibration function  $y = b_0 + b_1 \cdot x$ :

$$s_{\hat{x_s}} = \frac{s_e}{b_1} \sqrt{\frac{1}{w_s m} + \frac{1}{\sum w_i} + \frac{(\bar{y_s} - \bar{y_w})^2 \sum w_i}{b_1^2 \left(\sum w_i \sum w_i x_i^2 - (\sum w_i x_i)^2\right)}}$$
(1)

with

$$s_e = \sqrt{\frac{\sum w_i (y_i - \hat{y}_i)^2}{n - 2}} \tag{2}$$

where  $w_i$  is the weight for calibration standard  $i, y_i$  is the mean y value (!) observed for standard  $i, \hat{y_i}$  is the estimated value for standard i, n is the number calibration standards,  $w_s$  is the weight attributed to the sample s, m is the number of replicate measurements of sample  $s, \bar{y_s}$  is the mean response for the sample,  $\bar{y_w} = \frac{\sum_{w_i y_i} w_i y_i}{\sum_{w_i} w_i}$  is the weighted mean of responses  $y_i$ , and  $x_i$  is the given x value for standard i.

The weight  $w_s$  for the sample should be estimated or calculated in accordance to the weights used in the linear regression.

I adjusted the above equation in order to be able to take a different precisions in standards and samples into account. In analogy to Equation 8.26 from [1] we get

$$s_{\hat{x_s}} = \frac{1}{b_1} \sqrt{\frac{s_s^2}{w_s m} + s_e^2 \left( \frac{1}{\sum w_i} + \frac{(\bar{y_s} - \bar{y_w})^2 \sum w_i}{b_1^2 \left( \sum w_i \sum w_i x_i^2 - (\sum w_i x_i)^2 \right)} \right)}$$
(3)

## References

Massart, L.M, Vandenginste, B.G.M., Buydens, L.M.C., De Jong, S., Lewi,
 P.J., Smeyers-Verbeke, J. Handbook of Chemometrics and Qualimetrics:
 Part A, Elsevier, Amsterdam, 1997