

# Computational Statistics

## Exercise sheet 10

To be handed-in for marking until Tutorial of 05/07/2017

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[www.thphys.uni-heidelberg.de/~amendola/teaching.html](http://www.thphys.uni-heidelberg.de/~amendola/teaching.html)

### Problem 1 - Higher-order fitting [10 points]

Start by loading the dataset in file 'rmr\_ISwR.dat' provided in the Dropbox folder.

1. Print the data to understand its structure. Then, plot the metabolic rate as a function of body weight to first visualize the data of interest;
2. Following the previous tutorial, perform a first estimation of the linear fit coefficients by finding the slope and the intercept that minimize the residual standard deviation of the fit

$$\sigma = \sqrt{\frac{\sum_i^n (y_i - y_{\text{fit}})^2}{n - 2}} \quad (1)$$

3. Confirm your result by using the function 'lm' and printing its summary;
4. Plot the data and overplot with the linear fits obtained in points 2 and 3. Using the function 'confint', also overplot the 95% confidence limit of the linear fitting found in point 3;
5. Now, using the function 'lm', perform a quadratic fitting to the data. Also do a cubic fitting.
6. Again, plot the data and now overplot with the quadratic and cubic fitting;
7. Lastly, using the linear fit of point 3, predict the metabolic rates for body weights of 150 and 200 kgs, including the errors of the estimations.

### Problem 2 - Linear fitting with errors [5 points]

Start by loading the dataset in file 'rmr\_ISwR\_errors.dat' provided in the Dropbox folder. This should have the same structure of the file used in problem 1 with an added column containing the standard deviation of the measurements of the metabolic rates.

1. Print the data to understand its structure;

2. Compute the linear fit to the data by solving the system of Eqs. (4.16) and (4.17) as described in the most recent version of the lecture notes;
3. Check your results with those of function 'lm', using the errors as weights;
4. Plot the data, using the standard deviations as error bars of the data points, and overplot with the linear fit found before.

### Problem 3 - Generalized polynomial fit [5 points]

Imagine that you've got  $n$  data points  $y_i$  at positions  $x_i$ , each with a Gaussian error  $\sigma_i$ . The data points originate from a polynomial model of the form  $y(x) = \sum_{\alpha=0}^m a_{\alpha} x^{\alpha}$ , where obviously  $n \gg m$ . Please

1. formulate a  $\chi^2$ -functional for the fitting problem,
2. derive a linear system of equations by using the minimum conditions  $\partial_{\alpha} \chi^2 = 0$ ,
3. write this system in a matrix-vector-notation,
4. formulate the coefficients as *moments including a weighting* with the errors  $\sigma_i$ ,
5. invert the linear system for getting the model parameters  $a_{\alpha}$ ,
6. solve the special case  $m = 0$ : What's the solution for  $a_0$ ?
7. What happens if  $n = m$ ?