## Series 9

- 1. In this exercise we are investigating the ozone dataset which you have already seen in the lecture. The dataset ozone is available in numerous R-packages, e.g. in the package gss. You can load it with data(ozone, package = "gss"). If you do not have access to the package, you can get the data from http://stat.ethz.ch/Teaching/Datasets/ozone.dat. A short description of the variables is available at help(ozone, package = "gss") or at http://stat.ethz.ch/Teaching/Datasets/ozone.txt.
  - a) Get an overview of the data with pairs(). You should take the log of the response (upo3) and remove the outlier in the predictor wdsp.

## **R-Hints:**

The transformation can be done efficiently using

- d.ozone <- subset(transform(ozone,logupo3=log(upo3)),select=-upo3).</pre>
- b) We want to use MARS as regression method. To decide on the "best" interaction degree we use a leave-one-out cross validation (see 4.3.1). You should use the sum of squared errors as loss function. Try interaction degrees 1, 2, 3 and 4. Which one performs best?
  R-Hints:

earth() (a function performing MARS regression method) is available in the package earth. Usage: earth(formula, data, degree), where formula is the model formula (syntax as in lm()), data is a dataframe containing the measurements for the different variables, and degree is the maximum interaction degree. For more information see help(earth).

- c) Independently from b) we work now with a model with maximal interaction degree equal 2. Fit the model (with summary() of the earth() output you can take a look at the fitted model) and check the model assumptions with a Tukey-Anscombe plot. The residuals and the fitted values can be found in the output object of earth(). Use the help to find out how to extract them. To visualize the effects of the predictors we vary one variable from the main effects or two from an interaction while leaving the others constant. This can be done with the command plotmo(), which is also in the earth package. Read the help (help(plotmo)) and try different options to create nice plots.
- d) Now we want to compare MARS with an additive model (7.2 in the lecture notes). Therefore you have to use a MARS regression model with maximal interaction degree 1. R-Hints:

gam() (to fit an additive model) can be found in the package mgcv. Usage: gam(formula, data). formula must be of the form

```
logupo3 ~ s(vdht) + s(wdsp) + ...
```

Make use of summary() to get an overview of your gam() output. Remove the insignificant explanatory variable from the model and fit this reduced model once again.

To create nice plots use the function p.gam() from Dr. Mächler. You can get it with source("ftp://stat.ethz.ch/Teaching/maechler/CompStat/plotGAM.R").

Preliminary discussion: Friday, May 13.

**Deadline:** Friday, May 20.