Lab 2: Flow Control and the Urban Economy

36-350, Statistical Computing

Friday, 9 September 2011

Agenda: Manipulating data objects; understanding flow control; iterative approximation; computing for data analysis.

Instructions: Save all your answers in a single plain text file (Word files will not be graded), and upload it to Blackboard, using the page for this assignment. (There is no general digital dropbox any more.) When a question asks you to do something, give the command you use to do it. For questions which ask you to explain, write a short explanation in coherent, complete sentences. (You will be graded on your written explanation, not what you might say to the TA.)

In your intro. stats. class, you saw the linear regression model

\[ Y = b_0 + b_1 X + \text{noise} \]  

and learned how to fit it by least squares. In the linear model, there is an exact formula for the best estimate of the parameters \( b_0 \) and \( b_1 \). This is not usually the case with non-linear regression models (they “have no closed-form solution”), but we can estimate them by iteration.

Recently, it has been proposed that there is a non-linear relation between the number of people \( N \) in a city, and its economic output per person \( Y \) (per capita “gross metropolitan product”):

\[ Y = y_0 N^a + \text{noise} \]  

We will use data from the 366 metropolitan area of the US (as of 2006) to estimate \( a \), by minimizing the mean squared error,

\[ \frac{1}{366} \sum_{i=1}^{366} (Y_i - y_0 N_i^a)^2 \]  

over all \( a \). We will see later in the class how we could estimate both parameters at once, but for now we’ll treat \( y_0 \) as known, and in particular equal to $6611.

\(^1\text{by Geoffrey West and collaborators; there’s a reasonably entertaining on-line video of him giving a talk about the work to TED conference, if you’re into that sort of thing.}\)
1. Using the `read.table` command, load the `gmp.dat` file into a data frame called `gmp`. (5)

2. The last two columns give, for each city, the total gross metropolitan product (YN) and the per-capita metropolitan product (Y'). Use this to make a vector which has the population (N) for each city. (5)

3. Add the vector of populations to the `gmp` data frame as a new column, named `pop`. (5)

4. Calculate the mean squared error (MSE) of the power-law model, when \( y_0 = 6611 \) (in dollars) and \( a = 0.15 \). (10)

5. Use `seq()` to make a sequence of values of \( a \), from 0.10 to 0.15, in steps of 0.005. (5)

6. Calculate the MSE for each value of \( a \) in that sequence, and store all the MSEs in a vector. (Hint: There are many ways to do this, including a `for` loop.) (15)

7. Plot the square roots of the MSEs vs. \( a \). What are the units of the root mean squared error? What value of \( a \), among those you looked at, leads to the least squared error? (15)

8. Explain what the following code does. (20)

   ```r
   maximum.iterations <- 100
   deriv.step <- 1/1000
   step.scale <- 1e-12
   stopping.deriv <- 1/100
   iteration <- 0
   deriv <- Inf
   a <- 0.15
   while ((iteration < maximum.iterations) && (deriv > stopping.deriv)) {
     iteration <- iteration + 1
     mse.1 <- mean((gmp$pcgmp - 6611*gmp$pop^a)^2)
     mse.2 <- mean((gmp$pcgmp - 6611*gmp$pop^(a+deriv.step))^2)
     deriv <- (mse.2 - mse.1)/deriv.step
     a <- a - step.scale*deriv
   }
   list(a=a,iterations=iteration,converged=(iteration < maximum.iterations))
   ```

9. What value of \( a \) does the code estimate? What is the root mean squared error at this value of \( a \)? How well does this agree with what you expect from the plot you made for question 7? (10)

10. Re-run this code, but change the initial value of \( a \) from 0.15, to your estimate from question 9. What happens? Why? (10)