

**22S:166**  
**Homework due Mon. 10/09/06**

NOTE: Save your program for question 2, as you will be adding to it in a future homework assignment.

1. The function

$$x^3 + 4x^2 - 10$$

has a real root in  $[0,3]$ .

Use the built-in R function called `uniroot` to locate this root.

2. EM algorithm for grouped exponential data.

Suppose that 50 machines were simultaneously put into operation in an adverse environment. Once a week for four weeks, an engineer was sent in to observe how many machines had failed since the last check. Any time a machine failed, it was removed and not replaced. At the end of the 4-week test, the following report was issued:

Time interval in days	Number failing in interval
$a_j - b_j$	$n_j$
0 - < 8	15
8 - < 15	9
15 - < 22	7
22 - < 28	6
$\geq 28$	13

We believe that the exact failure times are drawn from an exponential distribution; that is

$$f(t) = \lambda \exp(-\lambda t), \quad t > 0$$

Obtain the maximum likelihood estimate of  $\lambda$ , based on the grouped data. Proceed as follows.

- Determine the complete-data sufficient statistic for  $\lambda$ , that is, the sufficient statistic if all the exact failure times were known.
  - Find the complete-data maximum likelihood estimator of  $\lambda$  as a function of the sufficient statistic (the expression, not the number).
  - Find a closed form expression for the conditional expectation of the exact failure time of an observation known to lie in the interval  $a_j - b_j$ , conditional on  $\lambda = \lambda^{(k)}$ .
  - Use the results of the previous parts to code an EM algorithm to compute the mle of  $\lambda$  (the numeric estimate).
  - Choose a stopping criterion and run your algorithm. Report your maximum likelihood estimate for  $\lambda$ . Turn in your code and output.
3. Verify that

$$\log[p(\theta|Y)] = \log[p(\theta|Z, Y)] - \log[p(Z|\theta, Y)] + \log[p(Z|Y)]$$

(This is the first step in proving that every EM algorithm increases the posterior (or likelihood)  $p(\theta|Y)$  at each iteration.