## **ASSIGNMENT 3**

## Due October 21, 2004, before 11:00 am

## Problem 3

This problem is essentially the same as Computer Problem 6.10 on p. 304 of Heath's book.

Let **A** be an  $n \times n$  real symmetric matrix with eigenvalues  $\lambda_1 \leq \lambda_2 \leq \cdots \leq \lambda_n$ . (Note that  $\lambda_n$  is the dominant eigenvalue and  $\lambda_1$  is the least dominant eigenvalue.) It can be shown that the critical points of the Rayleigh quotient (see Section 4.5.3) are eigenvectors of **A**, in particular

$$\lambda_1 = \min_{\mathbf{x} \neq \mathbf{0}} \frac{\mathbf{x}^T \mathbf{A} \mathbf{x}}{\mathbf{x}^T \mathbf{x}}$$
$$\lambda_n = \max_{\mathbf{x} \neq \mathbf{0}} \frac{\mathbf{x}^T \mathbf{A} \mathbf{x}}{\mathbf{x}^T \mathbf{x}}$$

with the minimum and maximum occurring at the corresponding eigenvectors. Thus we can in principle compute the extreme eigenvalues and corresponding eigenvectors of  $\mathbf{A}$  using any suitable method for optimization.

(a) Use the steepest descent method for unconstraint optimization to compute the extreme eigenvalues and corresponding eigenvectors of the matrix

$$\mathbf{A}_1 = \left[ \begin{array}{rrr} 6 & 2 & 1 \\ 2 & 3 & 1 \\ 1 & 1 & 1 \end{array} \right].$$

Is the solution unique in each case? Why or why not?

(b) The foregoing characterization of  $\lambda_1$  and  $\lambda_n$  remains valid if we restrict the vector  $\mathbf{x}$  to be normalized by taking  $\mathbf{x}^T \mathbf{x} = 1$ . Repeat part *a*, but apply the steepest descent method for constrained optimization to impose this normalization constraint.