Homework 5

Due date: April 5, 2017, 11:55PM.

Bibliography: Course lecture notes Lessons 19-23.

Note: Save this file as Homework5Solution.tm in your /home/student/courses/MATH547/homework directory. Then carry out your work on solving the problems in TeXmacs. Submit your final version through Sakai. If you've understood the theoretical concepts reviewed in the Background section and lecture notes, each problem should require about 20 minutes to complete, hence the entire homework can be completed within 3 hours.

1 Background

This assignment concentrates on solving the eigenvalue problem $A x = \lambda x$, with $A \in \mathbb{R}^{m \times m}$, $\lambda \in \mathbb{C}$, $x \in \mathbb{C}^m$.

2 Theory problems

2.1 Eigenvalues, eigenvectors of related matrices

Often knowledge of the solution of $Ax = \lambda x$ allows easy solution of some related eigenproblems. Solve 7.1.1-7.1.4

Ex7.1.1 Solution.

Ex7.1.2 Solution.

Ex7.1.3 Solution.

Ex7.1.4 Solution.

2.2 Geometric significance of eigenbases

Sometimes, the action of matrices upon vectors has straightforward geometric interpretations. In this case an eigenbasis for the matrix can be determined from geometric considerations.

Ex7.1.58 Solution.
Ex7.1.59 Solution.
Ex7.1.60 Solution.
Ex7.1.62 Solution.

2.3 Eigenvalue computation

Determine the eigenvalues for these problems by hand computation. Show your work. Verify your result in Octave.

Ex7.2.9 Solution.

Ex7.2.10 Solution.

Ex7.2.11 Solution.

Ex7.2.12 Solution.

2.4 Eigenvector computation

Determine the eigenvalues eigenvectors by hand computation. Show your work. Use insight into the structure of the problem.

Ex7.3.17 Solution. Ex7.3.18 Solution. Ex7.3.19 Solution.

Ex7.3.20 Solution.

3 Computational problems

This assignment and the next illustrate the applications of eigenvalues and the singular value decomposition in the face recognition problem. The following instructions load a matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$, containing n = 99 columns, where adding the average $\mathbf{a} \in \mathbb{R}^m$ to columns of \mathbf{A} gives an image with $m = 2^{14} = 16384$ pixels. Also loaded are matrices $\mathbf{R}, \mathbf{S} \in \mathbb{R}^{n \times p}$ of n = 99 coefficients of linear combinations of columns of \mathbf{A} to give p = 2000 face images.

```
octave> cd /home/student/courses/MATH547/homework/
octave> load .../lessons/mitfaces/faces.mat
octave> who
  Variables in the current scope:
  А
            R
                      S
                                        ans
                                                  dispans prompt
                               а
                                                                     r
octave> [size(A) size(a) size(R) size(S)]
   ans =
                                                 2000
                                                           99
      16384
                 99
                       16384
                                   1
                                          99
                                                                  2000
octave> [m,n]=size(A); [n,p]=size(R);
octave> [m n p]
   ans =
      16384
                 99
                        2000
```

octave>

A vector of length *m*, say the average face *a*, is displayed through the instructions octave> imagesc(flip(reshape(a,128,128)',1)); colormap(gray(256)); print -dpng aface.png octave>

An image constructed from a linear combination of columns of A, say with coefficients r_1 the first column of R is computed and displayed through

octave> f = A*R(:,1)+a;

```
octave> imagesc(flip(reshape(f,128,128)',1)); colormap(gray(256)); print -dpng r1face.png
octave>
```

octave>



Figure 1. Left: average face among database subjects. Middle: face constructed from linear combination of database subjects. Right

Muliple images, e.g. from 1300 to 1350, can be read through the following instructions

```
octave> basedir = '/home/student/courses/MATH547/lessons/mitfaces/rawdata/';
octave> n1=1301; n2=1350; nim=n2-n1+1; m=16384; B=zeros([m,nim]);
octave> for i=n1:n2
        fid = fopen(strcat(basedir,num2str(i)));
        B(1:m,i-n1+1) = fread(fid); fclose(fid);
        end
octave> size(B)
        ans =
        16384 50
octave>
```

octave>

The matrix A from the data is constructed from a certain subset of the available faces. This assignment focuses on constructing a different matrix $B \in \mathbb{R}^{m \times q}$ from a different subset, and comparing the two representations.

- 1. Choose a subset of q=50 files from the available faces (e.g., choose some n1, n2=49+n1, do not choose n1=1301). Use the script above to the read the faces into a matrix $B \in \mathbb{R}^{m \times q}$. Compute a vector $b \in \mathbb{R}^m$ that is the average of the chosen vectors. Use the sum Octave function. Display the average image.
- 2. The correlation matrix is defined as $C = \frac{1}{q}MM^T$, with column m_i of $M \in \mathbb{R}^{m \times q}$ defined in terms of column b_i of B and the average as $m_i = b_i b$, for i = 1, ..., q. What is the size of C? Find the relationship between eigenvalues, eigenvectors of C and those of $D = \frac{1}{q}M^TM$. Which is easier to compute, the eigenvalues of C or those of D?
- 3. Compute the eigenvalues and eigenvectors of D. Use these and your result from 2 to compute the eigenvalues and eigenvectors of B. Place the normalized eigenvectors in a matrix $X \in \mathbb{R}^{m \times q}$. Are the columns of X orthogonal to one another?

4. Choose one of the faces within the dataset, i.e., choose j, $\boldsymbol{f} = \boldsymbol{b}_j - \boldsymbol{b}$. Find the projection of this vector onto $C(\boldsymbol{X})$. Denote the projection as $\boldsymbol{X}\boldsymbol{u}$. It solves the least squares problem $\min_{\boldsymbol{u}} \|\boldsymbol{f} - \boldsymbol{X}\boldsymbol{u}\|$. Display the images obtained as $\boldsymbol{v} = \sum_{j=1}^{r} u_j \boldsymbol{x}_j$ for r = 2, 4, 8, 16, 32. Comment on the result.

Note: As usual, the above questions can be answered using very few lines of Octave code, no more that 5 lines per question. The main effort is properly conceptualizing the linear algebra operations, and thinking in terms of vectors and matrix operations. Pay particular attention to the dimension of the result from every matrix multiplication.