

## Homework 8

**Objective:** Eigenvalue decomposition based on Givens Rotation. SVD based on the pivoted Gram-Schmidt.

1. Given n-by-n symmetric matrix A, write a Matlab function `lambda=jacobi(A,eps0)` to implement the Jacobi Algorithm and find the eigenvalues `lambda = [lambda_1, lambda_2, ..., lambda_n]` of A.

As usual, `eps0` is a threshold to terminate the sweep iterations. If the Frobenius norm of `A-diag(A)` divided by `norm(A, 'fro')` is smaller than `eps0`, we terminate the iterations.

`lambda` is a vector of size n. Note that it is not required to compute the eigenvectors in this exercise.

It is always prudent to first verify, on a 2-by-2 symmetric matrix, the basic step of Jacobi Algorithm

$$J^T \begin{bmatrix} a & d \\ d & b \end{bmatrix} J = \begin{bmatrix} * & 0 \\ 0 & * \end{bmatrix}, \quad J = \begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix}, \quad \theta = \text{atan2}(2 * d, b - a)/2. \quad (1)$$

- (i) Provide Matlab script for the function
- (ii) Run it on the matrix `A=hilb(12)` with `eps0=1.0e-16`. Provide the number of sweep required for the prescribed precision `eps0`. For all the sweeps, print the ratios

$$\frac{\|A - \text{diag}(A)\|_F}{\|A\|_F} \quad (2)$$

- (iii) Show the eigenvalues with long e format. Compare them with the eigenvalues obtained via Matlab function `[q,s]=eig(A)` by computing the ratios of the corresponding pairs

$$\lambda_j/s_j, \quad j = 1 : 12 \quad (3)$$

2. Given m-by-n matrix A, write a Matlab function `[sigma,rank]=qralg(A,eps0)` to implement the QR Algorithm and find the singular values `sigma = [sigma_1, sigma_2, ..., sigma_rank]` of A. You are required to use the pivoted Gram-Schmidt function you constructed in the previous homework.

The basic idea behind the QR Algorithm is to successively produce a sequence of matrices `A0, A1, A2, ...` with `A0=A`. More precisely, a cycle in the QR Algorithm iteration consists of

- a. Compute the QR decomposition (based on the pivoted Gram-Schmidt algorithm) of `A0`, so that `A0=Q*R`
- b. Compute the QR decomposition of `transpose(R)`, so that `transpose(R)=U*T` is its QR factorization where `U` is like `Q`, and `T` is like `R`.
- c. At this point, we have `A=A0=Q*R = Q * T' * U'`. Thus, we call `T'` our next A matrix: `A1=T'`. Repeat to `A1` what we did to `A0` by following steps a, b, c.

As in Problem 1, `eps0` plays exactly the same role to terminate the cycle iterations. Note that it is not required to compute the singular vectors in this exercise.

- (i) Provide Matlab script for the function

- (ii) Run it on the matrix  $A=\text{hilb1}(m,n)$  with  $m=20$ ,  $n=10$ , and  $\text{eps0}=1.0\text{e-}16$ . Show the singular values with long e format.
- (iii) Compare them with the singular values obtained via Matlab function  $[u,s,v]=\text{svd}(A)$  by computing the ratio of the corresponding pairs

$$\sigma_j/s_j, \quad j = 1 : \text{rank} \quad (4)$$

3. Use the Matlab function  $[\text{sigma},\text{rank}]=\text{qralg}(A,\text{eps0})$  you constructed in Problem 2 to investigate the behavior of convergence of the diagonal entries of  $A_0, A_1, A_2, \dots$  to the singular values.

- (i) Construct a “random” matrix  $A$  with known singular values  $\text{sigma}=[0.1:0.1:1]$  by:  
 $[U,r]=\text{qr}(\text{rand}(15,10)); [V,r]=\text{qr}(\text{rand}(10,10)); A = U * \text{diag}(\text{sigma}) * V'$
- (ii) Run  $[\text{sigma},\text{rank}]=\text{qralg}(A,\text{eps0})$  with  $\text{eps0}=1.0\text{e-}16$ , and watch how each of the designated singular value is approached by the corresponding diagonal entry of  $A_0, A_1, A_2, \dots$ . Pay particular attention to the rate of convergence of each diagonal entry. Use no more than 30 words to summarize your observations.