

Numerische Mathematik II/ Numerical Analysis II

9. Assignment

Homework: HW9 (28-29.06.2011)

1. Let $U^T A V = \Sigma = \text{diag}(\sigma_1, \dots, \sigma_p)$, $p = \min\{m, n\}$ be the singular value decomposition of $A \in \mathbb{R}^{m \times n}$ with $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > \sigma_{r+1} = \dots = \sigma_p = 0$. Show that

- (a) $\text{rank}(A) = r$;
- (b) $\text{Ker}(A) = \text{span}\{v_{r+1}, \dots, v_n\}$;
- (c) $\mathcal{R}(A) (= \text{Image}(A)) = \text{span}\{u_1, \dots, u_r\}$.

(3 pts.)

2. Let the matrix $A = U + iV \in \mathbb{C}^{n \times n}$, $U, V \in \mathbb{R}^{n \times n}$, be Hermitian. Moreover, let

$$B = \begin{bmatrix} U & -V \\ V & U \end{bmatrix}.$$

Show that

- (a) U is symmetric, V is skew-symmetric and B is symmetric.
- (b) For $u, v \in \mathbb{R}^n$ let

$$z = \begin{bmatrix} u \\ v \end{bmatrix} \text{ and } z' = \begin{bmatrix} -v \\ u \end{bmatrix}$$

and $\lambda \in \mathbb{R}$. Then (λ, z) is an eigenpair of B if (λ, z') is an eigenpair of B .

- (c) Let $\lambda \in \mathbb{R}$ and $u, v \in \mathbb{R}^n$. Then $(\lambda, u + iv)$ is an eigenpair of A if (λ, z) and (λ, z') are eigenpairs of B .

(4 pts.)

3. Let $U = \begin{bmatrix} U_1 \\ U_2 \end{bmatrix} \in \mathbb{C}^{n \times k}$, $k \leq 2n$ be isometric, $U_1 \in \mathbb{C}^k$. Moreover let U_1 have the singular values $\alpha_1 \geq \dots \geq \alpha_k$ and U_2 the singular values $\beta_1 \leq \dots \leq \beta_k$. Then

$$\alpha_i^2 + \beta_i^2 = 1 \text{ for all } i.$$

(3 pts.)

4. Let $\mathcal{U}, \mathcal{V} \subseteq \mathbb{C}^n$ be k -dimensional subspaces and $x_1, \dots, x_k, y_1, \dots, y_k$ the corresponding canonical vectors. Show that

$$\langle x_i, y_j \rangle = 0, \text{ for } i \neq j.$$

(5 pts.)

- Using the singular value decomposition find the rank- k approximation of the image matrix associated with the image `gatlin`. Use

```
load gatlin
```

in Matlab to load the image data. For this image matrix

- show all the singular values (plot $\sigma_1, \sigma_2, \dots, \sigma_p$, $p = \min\{m, n\}$, use Matlab `semilogy`),
- show images for low-rank approximations with $k = 100, 25, 10, 1$,
- calculate the error $\|A - A_k\|_2$ and compare it to σ_{k+1} .

Use Matlab commands `image` and `colormap(map)` to display images.

- Write the program `powermethod.m` which computes the largest eigenvalue and the corresponding eigenvector for a given matrix using the *power method*.

- Test your program for matrix A and starting vector q , i.e.,

$$A = \begin{bmatrix} 1 & 1 & 1 \\ -1 & 9 & 2 \\ 0 & -1 & 2 \end{bmatrix}, \quad q = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}.$$

Perform at least 10 iterations.

- Compute the dominant eigenvector v using `[V,D] = eig(A)` and scale it such that you can compare it with q_k . Compare $|\frac{\lambda_2}{\lambda_1}|$ with

$$\frac{\|q_{k+1} - v\|}{\|q_k - v\|}, \quad k = 1, 2, \dots$$

- Repeat (a) and (b) for matrices

$$B = \begin{bmatrix} 1 & 1 & 1 \\ -1 & 9 & 2 \\ -4 & -1 & 2 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 1 & 1 \\ -1 & 3 & 2 \\ -4 & -1 & 2 \end{bmatrix}.$$

What difference do you observe and why? For the matrix B compare $|\frac{\lambda_2}{\lambda_1}|$ with

$$\sqrt[k]{\frac{\|q_{k+1} - v\|}{\|q_0 - v\|}}, \quad k = 1, 2, \dots$$