Chair for Algorithms and Data Structures
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## Search Engines WS 09/10

http://ad.informatik.uni-freiburg.de/teaching

## Exercise Sheet 10

complete until Thursday, January 21st

## Exercise 1

Compute the singular value decomposition of the following $2 \times 3$ matrix by hand.

$$
\left(\begin{array}{lll}
3 & 5 & 5 \\
3 & 7 & 1
\end{array}\right)
$$

## Exercise 2

Let $A_{k}$ be the the best rank- $k$ approximation of $A$ in the Frobenius norm, as defined in the lecture. Prove that $\left\|A-A_{k}\right\|_{F}^{2}=\sigma_{k+1}^{2}+\cdots+\sigma_{m}^{2}$, where $\sigma_{i}$ is the $i$ th largest singular value of $A$, and $m=\operatorname{rank}(A)$. First try to prove this on your own for some time, and then feel free to ask for hints.

## Exercise 3

Let $A$ be a symmetric $m \times m$ matrix with $m$ different, positive eigenvalues. The so-called power method starts with a random vector $x$ and repeatedly applies $A$ to it, and then normalizes the result after each step. That is, after $k$ steps, it has computed $x_{k}=A^{k} \cdot x /\left\|A^{k} \cdot x\right\|$. Prove that the power method converges to the eigenvector $u_{1}$ pertaining to the largest eigenvalue $\lambda_{1}$, that is, $\lim _{k \rightarrow \infty} x_{k}=u_{1}$.

Hint: Write $x$ as a linear combination of the $m$ eigenvectors $u_{1}, \ldots, u_{m}$ of $A$. You may assume without further proof that, because $x$ is chosen at random, it is not orthogonal to $u_{1}$.

## Exercise 4

Implement the power method for an arbitrary given symmetric matrix (it's ok to specify the matrix in the code). Don't use a library for matrix-matrix and matrix-vector multiplication but implement it yourself, it's simple. Run the method for a reasonable number of iterations. Check the correctness of your code by applying it to the $2 \times 2$ matrix $\left(\begin{array}{ll}2 & 1 \\ 1 & 2\end{array}\right)$ from the lecture. For that matrix, $u_{1}=(1 / \sqrt{2}, 1 / \sqrt{2})$ with eigenvalue 3 .

