

Optimization and Computational Linear Algebra for Data Science

Homework 5: Orthogonal matrices, eigenvalues and eigenvectors

Due on October 8, 2019

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- Unless otherwise stated, all answers must be mathematically justified.
 - Partial answers will be graded.
 - Your submission has to be uploaded to Gradescope. Indicate Gradescope the page on which each problem is written.
 - You can work in groups but each student must write his/her own solution based on his/her own understanding of the problem. Please list on your submission the students you work with for the homework (this will not affect your grade).
 - Problems with a (\star) are extra credit, they will not (directly) contribute to your score of this homework. However, for every 4 extra credit questions successfully answered your lowest homework score get replaced by a perfect score.
 - If you have any questions, feel free to contact me (lm4271@nyu.edu) or to stop at the office hours.
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Problem 5.1 (2 points). Let $A \in \mathbb{R}^{n \times n}$ be a symmetric matrix.

Show that if $v_1, v_2 \in \mathbb{R}^n$ are two eigenvectors of A associated to some eigenvalues $\lambda_1 \neq \lambda_2$ ($Av_1 = \lambda_1 v_1$ and $Av_2 = \lambda_2 v_2$), then $v_1 \perp v_2$.

Problem 5.2 (3 points). Let S be a subspace of \mathbb{R}^n and let P_S be the matrix of the orthogonal projection onto S . Let $M = \text{Id}_n - 2P_S$.

- (a) Show that the matrix M is orthogonal.
- (b) Show that if $\lambda \in \mathbb{R}$ is an eigenvalue of M , then $\lambda = 1$ or $\lambda = -1$.

Problem 5.3 (5 points). Let $A \in \mathbb{R}^{n \times n}$. We assume that there exists a basis (v_1, \dots, v_n) of \mathbb{R}^n consisting of eigenvectors of A :

$$Av_i = \lambda_i v_i$$

for all $i \in \{1, \dots, n\}$. We assume that

$$\lambda_1 > |\lambda_i| \quad \text{for all } i \in \{2, \dots, n\}.$$

We consider the following algorithm:

- Initialize $x_0 \in \mathbb{R}^n$.
- Perform the updates: $x_{t+1} = \frac{Ax_t}{\|Ax_t\|}$.

- (a) Show that for all $t \geq 1$,

$$x_t = \frac{A^t x_0}{\|A^t x_0\|}.$$

- (b) Assume that x_0 is a unit vector ($\|x_0\| = 1$) whose direction is chosen uniformly at random (this basically means that all the possible directions for x_0 are equally likely to be chosen). Let $(\alpha_1, \dots, \alpha_n)$ be the coordinates of x_0 in the basis (v_1, \dots, v_n) . Explain why we can be sure that $\alpha_1 \neq 0$. You do not have to do a rigorous proof of that, just give an intuitive argument.

(c) Show that

$$x_t \xrightarrow{t \rightarrow \infty} \frac{\alpha_1 v_1}{\|\alpha_1 v_1\|} \quad \text{and} \quad \|Ax_t\| \xrightarrow{t \rightarrow \infty} \lambda_1.$$

(d) Code the algorithm in *Python*: write a function **power_iteration** that takes a matrix A as input and outputs x_t and $\|Ax_t\|$ for the first t such that $\|x_t - x_{t-1}\| \leq 10^{-6}$. Here we assume that the matrix A given as input verifies the assumption of this problem.

Problem 5.4 (★). Let $A \in \mathbb{R}^{n \times n}$ be a symmetric matrix. Define the function

$$\begin{aligned} f: \mathbb{R}^n \setminus \{0\} &\rightarrow \mathbb{R} \\ x &\mapsto \frac{x^\top Ax}{x^\top x}. \end{aligned}$$

Show that f has a maximum at some $x_\star \in \mathbb{R}^n \setminus \{0\}$ and that x_\star verifies

$$Ax_\star = \lambda x_\star, \quad \text{where} \quad \lambda = f(x_\star).$$

