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## Lab - Eigenvalues and eigenvectors

### Introduction

The objectives of this lab are to show different contexts where eigenvalues are of interest and to introduce you to the simplest method for computing eigenvalues and eigenvectors, the power method.

### The power method

For a given matrix  $A$ , the power method can be used for computing the eigenvalue with the largest magnitude and the corresponding eigenvector.

### The theory behind the power method

Assume that the  $n \times n$  matrix  $A$  has the eigenvalues  $\lambda_j$ ,  $j = 1, \dots, n$  and that

$$|\lambda_1| > |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_n|. \quad (1)$$

Also assume that the eigenvectors  $x_j$ ,  $j = 1, \dots, n$  are linearly independent. This means that any vector  $z$  can be written as

$$z = \sum_{j=1}^n c_j x_j,$$

and

$$Az = \sum_{j=1}^n c_j Ax_j = \sum_{j=1}^n c_j \lambda_j x_j.$$

Repeated multiplication with  $A$  leads to

$$A^k z = \sum_{j=1}^n c_j \lambda_j^k x_j = \lambda_1^k \left( c_1 x_1 + c_2 \left( \frac{\lambda_2}{\lambda_1} \right)^k x_2 + \dots + c_n \left( \frac{\lambda_n}{\lambda_1} \right)^k x_n \right).$$

Because of (1) this behaves asymptotically as  $\lambda_1^k c_1 x_1$  (as  $k \rightarrow \infty$ ), i.e., a multiple of the eigenvector  $x_1$ .



## The power method algorithm

Choose an arbitrary initial guess  $z_0$  for the eigenvector.

For  $k = 0, 1, \dots$  repeat

$$\begin{cases} y_k &= \frac{z_k}{\|z_k\|_2}, \\ z_{k+1} &= Ay_k, \\ \mu_k &= y_k^H z_{k+1}, \end{cases}$$

until  $|\mu_k - \mu_{k-1}| < \text{tol.}$

The vectors  $y_k$  converge to the normalized eigenvector  $x_1$ . This means that  $\mu_k = y_k^H z_{k+1} = y_k^H Ay_k \approx y_k^H \lambda_1 y_k = \lambda_1 \|y_k\|_2 = \lambda_1$ .

Note that all vectors in the algorithm ( $z_k$  and  $y_k$ ) must be column vectors.

### Laboration

- Write a MATLAB function `[lambda,x]=mypower(A)` that applies the power method to a matrix  $A$  and returns the eigenvalue  $\lambda_1$  and the eigenvector  $x_1$ . Apply your algorithm to random matrices of different sizes. You can use the MATLAB function `eig` to check your results.
- Try some different ways of choosing the initial guess  $z_0$ . How much does it affect the number of iterations?
- Create a matrix  $A$  with  $A = \text{inv}(C) * \text{diag}(\text{lambda}) * C$  where `lambda` is a vector with the eigenvalues you want to give matrix  $A$  and  $C$  is a random matrix. Experiment with different eigenvalues and study how the convergence of the power method depends on the eigenvalues.

### Pay attention to:

- Convergence of the power method depends on the eigenvalues.