

# Optimization and Computational Linear Algebra for Data Science

## OUTLINE

Léo MIOLANE · [leo.miolane@gmail.com](mailto:leo.miolane@gmail.com)

July 8, 2019

### 1. VECTOR SPACES

1. General definitions
2. Linear dependency
3. Proof of Theorem [2.1](#)

### 2. LINEAR TRANSFORMATIONS

1. Linear transformations
2. Matrix representation
3. Kernel and image

### 3. RANK

1. More on basis
2. Definition of the rank
3. Properties of the rank
4. Transpose of a matrix, symmetric matrices

### 4. NORM AND DOT PRODUCT

1. Norm
2. Dot product
3. Orthogonality
4. Orthogonal projection and distance to a subspace

### 5. MATRICES AND ORTHOGONALITY

1. Gram-Schmidt orthogonalisation method
2. Orthogonal matrices

### 6. EIGENVALUES, EIGENVECTORS AND MARKOV CHAINS

1. Eigenvalues and eigenvectors
2. Diagonalizable matrices
3. Application to Markov chains
4. Example: Google's PageRank algorithm

### 7. SINGULAR VALUE DECOMPOSITION

1. The Spectral Theorem
2. Singular value decomposition
3. Interpretation of the SVD

### 8. GRAPHS AND LINEAR ALGEBRA

1. Graphs
2. Graph Laplacian
3. Spectral clustering with the graph Laplacian
4. Spectral clustering as a relaxation