Optimization and Computational Linear Algebra for Data Science Outline

Léo MIOLANE · leo.miolane@gmail.com

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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