

Linear Algebra for Computer Science

Homework 6

Read these first:

- i You may write your solutions on paper, under a word processing software (MS-word, Libre Office, etc.), or under \LaTeX .
- ii If writing on paper, you must use a scanner device or a Camera Scanner (CamScanner) software to scan the document and submit a *single* PDF file.
- iii Up to 15% extra score will be given to solutions written under \LaTeX , provided that you follow either of the following conventions:
 - (a) Represent scalars with normal (italic) letters (a, A), vectors with bold lower-case letters (\mathbf{a} , using `\mathbf{a}`), and matrices with bold upper-case letters (\mathbf{A} , using `\mathbf{A}`), or
 - (b) represent scalars with normal (italic) letters (a, A), vectors with bold letters (\mathbf{a}, \mathbf{A}), and matrices with typewriter upper-case letters (\mathbf{A} , using `\mathbf{A}`).
 - (c) Your LaTeX document must contain a *title*, a *date*, and your name as the author.
 - (d) In all cases, you must submit a *single* PDF file.
 - (e) If writing under \LaTeX , you must submit the *.tex* source (and other necessary source files if there are any) in addition to the PDF file.

Here is a short tutorial on \LaTeX : https://www.overleaf.com/learn/latex/Learn_LaTeX_in_30_minutes

Questions

Singular Value Decomposition

1. Let A be a nonsingular square matrix and $A = U\Sigma V^T$ be its (full) SVD. Prove that $\det(U)\det(V) = \text{sign}(\det(A))$, that is $\det(U)\det(V) = 1$ if $\det(A) > 0$ and $\det(U)\det(V) = -1$ if $\det(A) < 0$.



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2. Show that for a symmetric positive definite matrix the eigenvalue decomposition $\mathbf{A} = \mathbf{V}\Lambda\mathbf{V}^{-1} = \mathbf{V}\Lambda\mathbf{V}^T$ is the same as its singular value decomposition.
 3. Find a way to obtain the SVD of a symmetric matrix from its eigenvalue decomposition $\mathbf{A} = \mathbf{V}\Lambda\mathbf{V}^T$. Notice that the diagonal elements of Λ might be negative.
 4. Consider a matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$ and two orthogonal matrices $\mathbf{P} \in \mathbb{R}^{m \times m}$ and $\mathbf{Q} \in \mathbb{R}^{n \times n}$. Show that the singular values of \mathbf{PAQ} is the same as the singular values of \mathbf{A} .

Multivariate Calculus

5. Show that for a matrix $\mathbf{A} \in \mathbb{R}^{n \times n}$ the gradient of the expression $\mathbf{x}^T \mathbf{A} \mathbf{x}$ is equal to $(\mathbf{A} + \mathbf{A}^T) \mathbf{x}$. What is the gradient when \mathbf{A} is symmetric?
6. Show that for a symmetric matrix \mathbf{B} the gradient of $1/(\mathbf{x}^T \mathbf{B} \mathbf{x})$ with respect to \mathbf{x} is $-2\mathbf{Bx}/(\mathbf{x}^T \mathbf{Bx})^2$ (if the gradient exists at \mathbf{x}).
7. Show that for symmetric matrices \mathbf{A} and \mathbf{B} the gradient of $f(\mathbf{x}) = (\mathbf{x}^T \mathbf{A} \mathbf{x}) / (\mathbf{x}^T \mathbf{B} \mathbf{x})$ with respect to \mathbf{x} is equal to

$$2(\mathbf{Ax}(\mathbf{x}^T \mathbf{Bx}) - \mathbf{Bx}(\mathbf{x}^T \mathbf{Ax})) / (\mathbf{x}^T \mathbf{Bx})^2 = 2(\mathbf{Ax} - f(\mathbf{x}) \mathbf{Bx}) / (\mathbf{x}^T \mathbf{Bx}),$$

if the gradient exists at \mathbf{x} .

8. Let \mathbf{A} be symmetric. Calculate the gradient of $\exp(-\mathbf{x}^T \mathbf{Ax})$ with respect to \mathbf{x} .
9. Let \mathbf{A} be (symmetric) positive definite. Compute the gradient of $\log(1 + \mathbf{x}^T \mathbf{Ax})$ with respect to \mathbf{x} .
10. Consider the function $f(\mathbf{x}) = \mathbf{x}^T \mathbf{Ax} / \|\mathbf{x}\|^2 = \mathbf{x}^T \mathbf{Ax} / (\mathbf{x}^T \mathbf{x})$ defined for a symmetric matrix \mathbf{A} . Show that the critical points of f are exactly the eigenvectors of \mathbf{A} . The critical points of a function f are points \mathbf{x} at which the gradient is zero or nonexistent.
11. Consider the function $f(\mathbf{x}) = \mathbf{x}^T \mathbf{Ax} / (\mathbf{x}^T \mathbf{Bx})$ defined for symmetric matrices \mathbf{A} and \mathbf{B} . Show that if \mathbf{B} is invertible then the critical points of f are either the points for which $\mathbf{x}^T \mathbf{Bx} = 0$ or the eigenvectors of $\mathbf{B}^{-1} \mathbf{A}$.