MLFCS

11: Elementary Matrices + Invertible Matrix Theorem

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Eigenvectors and Eigenvalues; Diagonalization

- ▶ Def.: $v \neq 0$ and λ are eigenvalues of $A \iff Av = \lambda v$
- Find λ_i as roots from equating the characteristic polnomial to zero: $det(A \lambda I_n) = 0$
- Find v_i (for each λ_i) by Gaussian Elimination

Coordinate Transformation by the Eigenvectors

- For an invertible matrix A, define $T := (v_1 v_2 \dots v_n)$ as the matrix formed by the eigenvectors v_i as **columns**
- Define a diagonal matrix D with the eigenvalues λ_i as the diagonal elements

$$P = \begin{pmatrix} c & -s \\ s & c \end{pmatrix} \qquad \mathcal{D} = \begin{pmatrix} c + is \\ s & c \end{pmatrix} = \begin{pmatrix} c + is \\ s & c$$

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Coordinate Transformation by the Eigenvectors

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- ▶ Define a diagonal matrix D with the eigenvalues λ_i as the diagonal elements
- ▶ Then $TDT^{-1} = A$, as well as $T^{-1}AT = D$
- Hence, A has an equivalent diagonal matrix D, in a transformed space
 Therefore this procedure is often called Diagonalization

Finding the Inverse via Gaussian Elimination

Recall: We can use Gaussian Elimination (row operations) to solve $A\vec{x} = \vec{e}_i$ for any i). (A(i)) = (A(i))

- We do this by "augmenting" the $n \times n$ matrix A by a rhs vector.
- Row operations are solely based on matrix A, the rhs is just enjoying the same row operations.
- Why not add more rows, namely **all** n unit vectors on the rhs. Actually, we just add the whole $n \times n$ unity matrix there...
- on the other hand, we simultaneously solve n equations $A\vec{x} = \vec{e}_i$
- We start with a $n \times 2n$ matrix composed of A (left) and a unit matrix (right).
- If A is invertible, we obtain the unix matrix on the left and A^{-1} on the right. $\rightarrow (1)$

Elementary Matrices

Each Gaussian Operation is equivalent by multiplying both sides (from the left) with an Elementary Matrix! Swapping 2 rows:

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 2 & 1 & 1 \\ 3 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 3 & 1 & 0 \\ 2 & 1 & 1 \end{pmatrix}$$

Adding 2 times R2 to R1: (whereby R2 stays unchanged)

$$\left(\begin{array}{ccc} 1 & 2 \\ 0 & 1 \end{array}\right) \left(\begin{array}{ccc} 2 & 1 & 1 \\ 3 & 1 & 0 \end{array}\right) = \left(\begin{array}{ccc} 8 & 3 & 1 \\ 3 & 1 & 0 \end{array}\right)$$

Multiplying row 2 by a:

$$\left(\begin{array}{ccc}1&0\\0&a\end{array}\right)\left(\begin{array}{ccc}2&1&1\\3&1&0\end{array}\right)=\left(\begin{array}{ccc}2&1&1\\3a&a&0\end{array}\right)$$

Okay...? Let's go!

Elementary Matrices

$$\begin{pmatrix} 1/a & 0 \\ -1/a & 1/c \end{pmatrix} \begin{pmatrix} a & b & 1 & 0 \\ c & d & 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & b/a & 1/a & 0 \\ 0 & (ad - bc)/ac & -1/a & 1/c \end{pmatrix}$$

Next operations

$$\begin{pmatrix} 1 & 0 \\ 0 & ac/(ad-bc) \end{pmatrix} \begin{pmatrix} 1 & b/a & 1/a & 0 \\ 0 & (ad-bc)/ac & -1/a & 1/c \end{pmatrix} =?$$

Work out your steps! The inverse is composed by a product of several elementary matrices Arriving at:

$$\begin{pmatrix} d/(ad-bc) & -b/(ad-bc) \\ -c/(ad-bc) & a/(ad-bc) \end{pmatrix} \begin{pmatrix} 1 & b/a & 1/a & 0 \\ 0 & (ad-bc)/ac & -1/a & 1/c \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 & d/(ad-bc) & -b/(ad-bc) \\ 0 & 1 & -c/(ad-bc) & a/(ad-bc) \end{pmatrix}$$

We recognize that the right half provides the inverse

$$\left(\begin{array}{cc} a & b \\ c & d \end{array}\right)^{-1} = \frac{1}{ad - bc} \left(\begin{array}{cc} d & -b \\ -c & a \end{array}\right)$$

Invertible Matrix Theorem

THEOREM. For a $n \times n$ square matrix, equivalently:

- (a) A is invertible
- (b) Ax = 0 has only the trivial solution
- (c) The reduced row echelon form of A is the identity matrix I_n (c2) A is row-equivalent to the $n \times n$ identity matrix I_n .
- (c3) A has n pivot positions
- (d) A is expressible as a product of elementary matrices
 - (e) Ax = b is consistent for every $n \times 1$ matrix b
 - (f) Ax = b has exactly one solution for every $n \times 1$ matrix b
- (g) $\det(A) \neq 0$ (h) The column vectors of A are linearly independent
 - (i) The row vectors of A are linearly independent
- (i) The column vectors of A span \mathbb{R}^n
- (k) The row vectors of A span \mathbb{R}^n
- (I) The column vectors of A form a basis for \mathbb{R}^n (m) The row vectors of A form a basis for \mathbb{R}^n
- (n) A has rank n (rank := number of lin. indep. rows) 8/11

Invertible Matrix Theorem (cont'd)

THEOREM. For a $n \times n$ square matrix, equivalently:

- (a) A is invertible
- (b) Ax = 0 has only the trivial solution
- (d) A is expressible as a product of elementary matrices
- (f) Ax = b has exactly one solution for every $n \times 1$ matrix b(g) $\det(A) \neq 0$
 - (h) The column (or row) vectors of A are linearly independent
 - (n) A has rank n (rank := number of lin. indep. rows)
- (i) A has nullity 0 (nullity := n- rank)
 - (p) The orthogonal complement of the null space of A is \mathbb{R}^n
 - (q) The orthogonal complement of the row space of A is 0 (r) The kernel of T_A is 0
 - (s) The range of T_A is \mathbb{R}^n (t) T_A is one-to-one
- (u) 0 fails to be an eigenvalue of A
- (u2) A has n nonzero eigenvalues

(v) The transposed matrix A^{T} is invertible

Invertible Matrix Theorem (selection)

THEOREM. For a $n \times n$ square matrix, equivalently:

- (a) A is invertible
- (b) Ax = 0 has only the trivial solution
- (c) The reduced row echelon form of A is the identity matrix I_n
- (d) A is expressible as a product of elementary matrices
- (f) Ax = b has exactly one solution for every $n \times 1$ matrix b
- (g) $\det(A) \neq 0$
- (h) The column (or row) vectors of A are linearly independent
- (n) A has rank n
 - (t) The mapping $x \to Ax$ is one-to-one
- (u) 0 fails to be an eigenvalue of A
- (u2) A has n nonzero eigenvalues
 - (v) The transposed matrix A^T is invertible

Linear Algebra: wrap-up!

 We solved systems of linear equations via Gaussian Elimination.

These could be inconsistent (0=1, no solutions), have a unique solution, or a *k*-dimensional subspace of (infinitely many) solutions

- ► Vector space (linear space). Linear independence of vectors, span, basis, inner product, orthogonality...
- ► Inverse of a matrix. Determinant of a matrix.
- ► Eigenvalues and Eigenvectors. Symmetric matrix → real-valued eigenvalues Asymmetric matrix: expect complex eigenvalues! e.g.: rotation matrices!
- Coordinate transform (via eigenvectors) to coordinates where the matrix is diagonal
- Finding the Inverse via Gaussian Elimination: each step corresponds to a elementary matrix
- ► Invertible Matrix Theorem: many useful equivalent conditions!