Midterm Review

Review Lecture 2

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- Most Important: High-Frequency Problems
- 3 Other Useful Knowledge in Exam
- 4 Proof: Rank Relations

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We Have Learnt...

Chapter 3:

- 3.1 Orthogonal Vectors and Subspaces
- 3.2 Cosines and Projection onto Lines
- 3.3 Projections and Least Squares

A Summary: You are encouraged to understand the concept of orthogonal complement and least-square algorithm. Important section: 3.3. It will take approximately 10 marks in your exam.

Midterm in Previous Years: 2019

Midterm in 2019 has 7 problems, with a total of 100 marks.

- \bigcirc (12 marks) True or False. (2'×6)
- (15 marks) Fill the blanks. $(3' \times 5)$
- (24 marks) Ax = 0, Ax = b and the dimensions, bases of four fundamental subspaces.
- ♠ (14 marks) LDL^T factorization and find the inverse of given matrix.
- (15 marks) Least squares and split vector into 2 orthogonal subspaces.
- 6 (10 marks) Linear Transformations. (Transposing)
- (10 marks) Proof about linear independence and subspaces.

Problem 3, 4, 5 are not that difficult, and they take up 53 marks!

Midterm in Previous Years: 2020

Midterm in 2020 has 8 problems, with a total of 110 marks.

- \bigcirc (15 marks) Multiple Choices. (3' \times 5)
- 2 (25 marks) Fill the blanks. $(5' \times 5)$
- (10 marks) LU factorization.
- 4 (16 marks) Dimensions, bases of four fundamental subspaces.
- 6 (10 marks) Linear Transformations. (given basis)
- 6 (6 marks) Proof or counter example.
- (10 marks) Linear Transformations. (natural basis)
- (12 marks) Proof about linear independence.

Again, Problem 3, 4, 5 are not that difficult, and they take up 36 marks!

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Linear Transformation: Given Basis

Example

Let $L: \mathbb{R}^2 \to \mathbb{R}^3$ be the linear transformation defined by

$$L(\mathbf{x}) = \begin{bmatrix} x_2 \\ x_1 + x_2 \\ x_1 - x_2 \end{bmatrix}$$

Find the matrix representations of L with respect to the ordered bases $\{\mathbf{u}_1, \mathbf{u}_2\}$ and $\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$, where

$$\mathbf{u_1} = \left[\begin{array}{c} 1 \\ 2 \end{array} \right], \mathbf{u_2} = \left[\begin{array}{c} 3 \\ 1 \end{array} \right]$$

and

$$\mathbf{b_1} = \left[\begin{array}{c} 1 \\ 0 \\ 0 \end{array} \right], \mathbf{b_2} = \left[\begin{array}{c} 1 \\ 1 \\ 0 \end{array} \right], \mathbf{b_3} = \left[\begin{array}{c} 1 \\ 1 \\ 1 \end{array} \right]$$

Linear Transformation: Given Basis

Solution:

Input coordinates $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$, which is $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ in natural basis.

The output in natural basis will be

$$L\left(\left[\begin{array}{c}1\\2\end{array}\right]\right) = \left[\begin{array}{c}2\\1+2\\1-2\end{array}\right] = \left[\begin{array}{c}2\\3\\-1\end{array}\right]$$

Transform to the coordinates under new basis \mathbf{b} .

$$\begin{bmatrix} 2 \\ 3 \\ -1 \end{bmatrix} = a_{11} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + a_{21} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + a_{31} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

Output coordinates $\begin{bmatrix} -1\\4\\-1 \end{bmatrix}$, which is $\begin{bmatrix} 2\\3\\-1 \end{bmatrix}$ in natural basis.

Linear Transformation: Given Basis

Solution:

Input coordinates $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$, which is $\begin{bmatrix} 3 \\ 1 \end{bmatrix}$ in natural basis.

The output in natural basis will be

$$L\left(\left[\begin{array}{c}3\\1\end{array}\right]\right) = \left[\begin{array}{c}1\\3+1\\3-1\end{array}\right] = \left[\begin{array}{c}1\\4\\2\end{array}\right]$$

Output coordinates $\begin{bmatrix} -3 \\ 2 \\ 2 \end{bmatrix}$, which is $\begin{bmatrix} 1 \\ 4 \\ 2 \end{bmatrix}$ in natural basis.

The transformation matrix:

$$A = \begin{bmatrix} -1 & -3 \\ 4 & 2 \\ -1 & 2 \end{bmatrix}$$

Least-Squares

Example

Let

$$A = \begin{bmatrix} -1 & 1 & 0 \\ 1 & -2 & 1 \\ 1 & -1 & 2 \\ 1 & -1 & -2 \end{bmatrix}, b = \begin{bmatrix} 2 \\ -2 \\ -1 \\ 3 \end{bmatrix}$$

Find the least squares solution to Ax = b.

The system must be inconsistent or we cannot find the least squares solution, because we can find the exact solution!

The method to find least squares solution is to solve

$$A^T A \hat{x} = A^T b$$



Least-Squares

Solve this system to get the least squares solution:

$$\begin{bmatrix} -1 & 1 & 1 & 1 \\ 1 & -2 & -1 & -1 \\ 0 & 1 & 2 & -2 \end{bmatrix} \begin{bmatrix} -1 & 1 & 0 & 2 \\ 1 & -2 & 1 & -2 \\ 1 & -1 & 2 & -1 \\ 1 & -1 & -2 & 3 \end{bmatrix}$$

Multiply that:

$$\begin{bmatrix} 4 & -5 & 1 & -2 \\ -5 & 7 & -2 & 4 \\ 1 & -2 & 9 & -10 \end{bmatrix} \rightarrow \begin{bmatrix} 4 & -5 & 1 & -2 \\ 0 & 3 & -3 & 6 \\ 0 & 0 & 32 & -32 \end{bmatrix}$$

The least squares solution is

$$\hat{x} = \left[\begin{array}{c} 1 \\ 1 \\ -1 \end{array} \right]$$

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Relations of Four Fundamental Subspaces

Always keep the 4 fundamental subspaces in mind!

- Row space is the orthogonal complement of nullspace, column space is the orthogonal complement of left nullspace.
- If C(A) is the same with $C(A^T)$, then N(A) is the same with $N(A^T)$.
- Dimension: $dim(C(A)) = dim(C(A^T)) = r$, dim(N(A)) = n r, $dim(N(A^T)) = m r$.
- If AB = 0, then $C(B) \in N(A)$, $C(A^T) \in N(B^T)$. $C(A^T)$ is orthogonal to C(B), that means every vector in $C(A^T)$ is orthogonal to every vector in C(B).

Properties of Linear Transformations

- T(0)=0.
- Ax is in the column space of A.
- Rank of A is the dimension of the output space C(A).
- Some special linear transformations: rotation, reflection, projection.

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Proof of Rank Inequalities

The Basics: $rank(AB) \leq max\{rank(A), rank(B)\}$.

1. rank(PAQ) = rank(A) if P, Q are invertible.

Proof:

$$rank(PAQ) \leqslant rank(A)$$
 $rank(A) = rank(P^{-1}PAQQ^{-1}) \leqslant rank(PAQ)$
 $rank(PAQ) = rank(A)$

It shows: Invertible (elementary) row and column operations will not change the rank.

So, we can prove a series of rank inequalities by constructing matrix and do invertible row and column operations.

Proof of Rank Inequalities

2.
$$rank(A + B) \leq rank(A) + rank(B)$$
.

Proof:

$$rank(A) + rank(B) = r \begin{pmatrix} \begin{bmatrix} A & 0 \\ 0 & B \end{bmatrix} \end{pmatrix}$$
$$= r \begin{pmatrix} \begin{bmatrix} A & A \\ 0 & B \end{bmatrix} \end{pmatrix} = r \begin{pmatrix} \begin{bmatrix} A & A \\ A & A + B \end{bmatrix} \end{pmatrix} \geqslant rank(A + B)$$

3.
$$r\left(\begin{bmatrix} A & 0 \\ 0 & B \end{bmatrix}\right) \leqslant r\left(\begin{bmatrix} A & 0 \\ C & B \end{bmatrix}\right)$$
.

Proof:

Rank equals number of pivots. Matrix C can provide additional pivots.

Proof of Rank Inequalities

4. $rank(A) + rank(B) \leqslant rank(AB) + n$.

Proof:

$$rank(AB) + n = r \begin{pmatrix} I_{n \times n} & 0 \\ 0 & AB \end{pmatrix}$$
$$= r \begin{pmatrix} I_{n \times n} & 0 \\ A & AB \end{pmatrix} = r \begin{pmatrix} I_{n \times n} & -B \\ A & 0 \end{pmatrix} \geqslant rank(A) + rank(B)$$

5. $rank(AB) + rank(BC) - rank(B) \leqslant rank(ABC)$.

Proof:

$$rank(ABC) + rank(B) = r \begin{pmatrix} \begin{bmatrix} ABC & 0 \\ 0 & B \end{bmatrix} \end{pmatrix}$$

$$= r \begin{pmatrix} \begin{bmatrix} ABC & 0 \\ AB & B \end{bmatrix} \end{pmatrix} = r \begin{pmatrix} \begin{bmatrix} ABC & 0 \\ AB & B \end{bmatrix} \end{pmatrix} = r \begin{pmatrix} \begin{bmatrix} 0 & -BC \\ AB & B \end{bmatrix} \end{pmatrix}$$

$$= r \begin{pmatrix} \begin{bmatrix} AB & B \\ 0 & -BC \end{bmatrix} \end{pmatrix} \geqslant rank(AB) + rank(BC)$$

A Conclusion: $rank(A^TA) = rank(A)$

Proof:

Firstly, we prove $A^TAx = 0 \rightarrow Ax = 0$.

Suppose x is in the nullspace of A^TA :

$$A^T A x = 0$$

Multiply x^T on both sides:

$$x^{T}A^{T}Ax = 0 \rightarrow (Ax)^{T}Ax = 0 \rightarrow ||Ax|| = 0 \rightarrow Ax = 0$$

Secondly, we prove $Ax = 0 \rightarrow A^T Ax = 0$.

If Ax = 0, $A^TAx = 0$ must hold because no matrix can bring zero vector out of the origin.

So, $N(A^TA) = N(A)$, so does the row space (orthogonal complement) and rank $R(A^TA) = R(A)$, $rank(A^TA) = rank(A)$.

Good Luck!

Hope you all can do well in the midterm exam!!!