MA 106 D1-T3 Recap-1

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■ Multiply the i^{th} row by $c \neq 0$.



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- No. of pivots \leq min(no. of rows, no. of columns)
- Not unique.
- Can be reduced.

RREF

Theorem

Let A be a $n \times n$ square matrix. There exist ERM's E_i such that

$$\Pi_i E_i A$$

is either the $n \times n$ identity matrix or has the last row zero.

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Definition

Let $A \in \mathbb{R}^{n \times n}$. It is said to be invertible if there exists $B \in \mathbb{R}^{n \times n}$ such that

$$AB = BA = I$$

where I is the $n \times n$ identity matrix. Such a B is unique and is denoted as A^{-1}

Gauss Jordan

Theorem

A square matrix A is invertible **iff** it is a product of ERM's.

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Questions:

- Is \mathbb{Q}^2 a subspace of \mathbb{R}^2 ?
- Is R^2 a subspace of \mathbb{R}^3 ?

<u>To do</u>: Categorize the subspaces of \mathbb{R}^2 and \mathbb{R}^3

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- Linear dependence: There exist a_i not all zero such that

$$\sum_i a_i v_i = 0$$

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- Redundancy?
- Linear Independence



Fundamental Lemma

Lemma

Let V be a vector subspace of \mathbb{R}^n generated by k vectors. Any set of I vectors such that $I \geq k+1$ is linearly dependent.

Basis and Dimension

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Let V be a vector subspace of \mathbb{R}^n . A subset S of V is called a basis of V if

- *S* is linearly independent.
- \blacksquare Span(S) = V

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- V has a basis.
- **2** Each basis of *V* has the same number of elements.

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(Revisiting the) Question: Show that $v_1, v_2, v_3 \in \mathbb{R}^3$ being linearly independent is equivalent to $det(\begin{bmatrix} v_1 & v_2 & v_3 \end{bmatrix}) \neq 0$

Rank and all that

Definitions

Let $A \in \mathbb{R}^{m \times n}$. In terms of its columns and rows, A can be written as

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Kronecker-Capelli Theorem

Theorem

A system of linear equations

$$Ax = b$$

has a solution iff

$$Rank(A) = Rank([A|b])$$

Rank & Nullity

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Rank Nullity Theorem

Given $A \in \mathbb{R}^{m \times n}$.

$$R(A) + N(A) = n$$

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Fact:

$$det \ rank(A) = rank(A)$$