

Midterm 1 Review

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

Linear System : $a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$
 $\vdots \quad \vdots \quad \vdots \quad \vdots$
 $a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m$

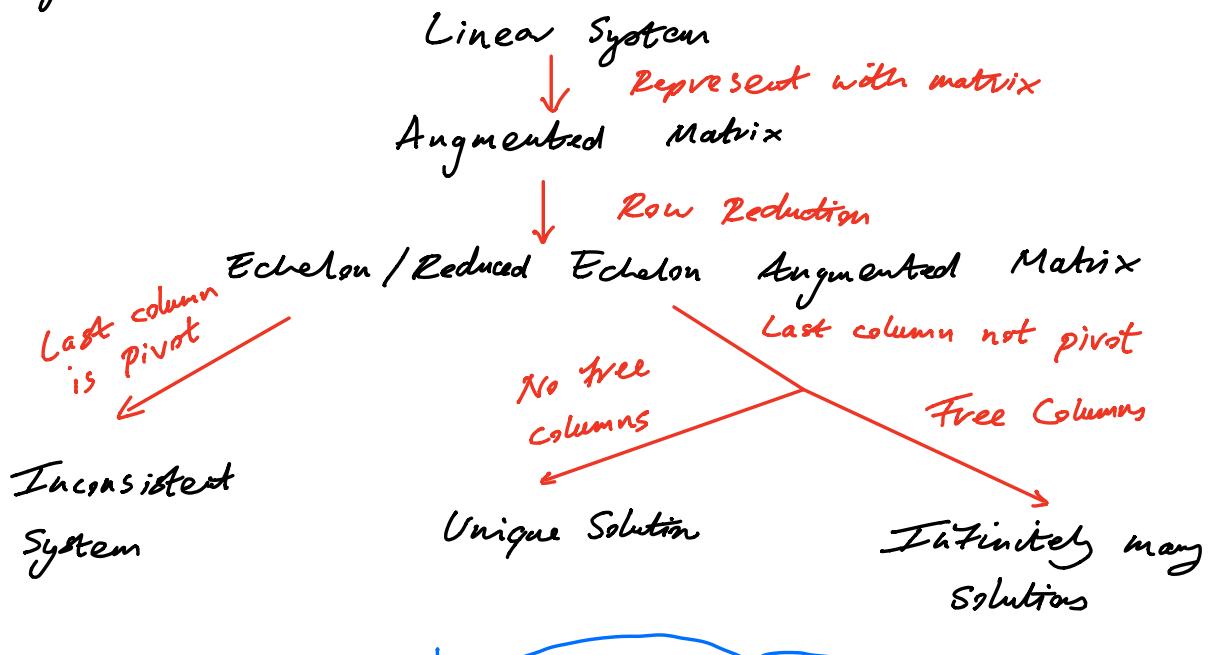
x_1, x_2, \dots, x_n = Solution to linear system

Augmented Matrix :
$$\left(\begin{array}{ccc|c} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_m \end{array} \right)$$

Coefficient Matrix A

Key Fact: Row operations on augmented matrix alter linear system but not the set of solutions.

Algorithm to find general solution :



Found by backwards substitution.

Example

$$\left(\begin{array}{ccc|c} 1 & 0 & 2 & 0 \\ 0 & 1 & 3 & 1 \end{array} \right) \Rightarrow \begin{array}{l} x_1 + 2x_3 = 0 \\ x_2 + 3x_3 = 1 \end{array}$$

↑ ↑ ↑
Pivot Free
Columns Column

$$\Rightarrow \text{General Solution} = \begin{cases} -2x_3 \\ 1 - 3x_3 \\ x_3 \text{ free} \end{cases}$$

called vector in \mathbb{R}^m

Vector/Matrix Perspective

$$\mathbb{R}^m = \left\{ \begin{pmatrix} v_1 \\ \vdots \\ v_m \end{pmatrix} \text{ such that } v_i \in \mathbb{R} \right\} \text{ we write } \underline{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_m \end{pmatrix}$$

Key Fact: We can add and scale vectors in \mathbb{R}^m .

$$\underline{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_m \end{pmatrix}, \underline{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_m \end{pmatrix} \Rightarrow \underline{u} + \underline{v} := \begin{pmatrix} u_1 + v_1 \\ \vdots \\ u_m + v_m \end{pmatrix}$$

$$\lambda \in \mathbb{R} \quad \lambda \underline{u} = \begin{pmatrix} \lambda u_1 \\ \vdots \\ \lambda u_m \end{pmatrix}$$

$$A - m \times n \text{ matrix} \Rightarrow A = (\underline{a}_1 \dots \underline{a}_n) \quad \text{in } \mathbb{R}^m$$

Notation : $\underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \text{ in } \mathbb{R}^n$ linear combination

$$\Rightarrow A \underline{x} := x_1 \underline{a}_1 + x_2 \underline{a}_2 + \dots + x_n \underline{a}_n$$

$$\text{Let } \underline{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix} \text{ in } \mathbb{R}^m$$

$$\underline{x} \text{ is solution to } (A \mid \underline{b}) \Leftrightarrow A \underline{x} = \underline{b}$$

$$\Rightarrow (A \mid \underline{b}) \text{ consistent} \Leftrightarrow \underline{b} \text{ a linear combination of } \{\underline{a}_1, \dots, \underline{a}_n\}$$

admits a solution $\Leftrightarrow \underline{b} \text{ in } \text{Span}(\underline{a}_1, \dots, \underline{a}_n)$

" all linear combinations

Conclusion

\underline{b} in $\text{Span}(\underline{a}_1, \dots, \underline{a}_n) \Leftrightarrow (A| \underline{b})$ consistent \Leftrightarrow

Last column
at
reduced $(A| \underline{b})$
not a pivot

E.g.
$$\left(\begin{array}{ccc|c} 1 & 1 & 1 & 3 \\ 0 & 1 & 2 & 4 \\ 0 & 0 & 0 & 0 \end{array} \right) \Rightarrow \left(\begin{array}{c} 3 \\ 4 \\ 0 \end{array} \right) \text{ in } \text{Span} \left(\left(\begin{array}{c} 1 \\ 0 \end{array} \right), \left(\begin{array}{c} 1 \\ 1 \end{array} \right), \left(\begin{array}{c} 1 \\ 2 \end{array} \right) \right)$$

*↑
Not pivot*

$$\left(\begin{array}{ccc|c} 1 & 1 & 1 & 3 \\ 0 & 1 & 2 & 4 \\ 0 & 0 & 0 & 1 \end{array} \right) \Rightarrow \left(\begin{array}{c} 3 \\ 4 \\ 1 \end{array} \right) \text{ not in } \text{Span} \left(\left(\begin{array}{c} 1 \\ 0 \end{array} \right), \left(\begin{array}{c} 1 \\ 1 \end{array} \right), \left(\begin{array}{c} 1 \\ 2 \end{array} \right) \right)$$

Pivot Position in every row of reduced A \Leftrightarrow Every row at reduced A is non-zero \Leftrightarrow Last column at reduced $(A| \underline{b})$ is not pivot for any \underline{b} in \mathbb{R}^m

$\text{Span}(\underline{a}_1, \dots, \underline{a}_n) = \mathbb{R}^m \Leftrightarrow$ Reduced A has pivot position in every row

$\{\underline{a}_1, \dots, \underline{a}_n\}$ linearly independent :
(L.I.)

$$x_1 \underline{a}_1 + x_2 \underline{a}_2 + \dots + x_n \underline{a}_n = \underline{0} \Rightarrow x_1 = x_2 = \dots = x_n = 0$$

"

$$A\underline{x}$$

homogeneous linear system

$\Rightarrow \{\underline{a}_1, \dots, \underline{a}_n\}$ L.I. $\Leftrightarrow A\underline{x} = \underline{0}$ has unique solution $\underline{x} = \underline{0}$

\Leftrightarrow Reduced $(A| \underline{0})$ has no free columns

\Leftrightarrow Reduced A has pivot position in every column

$$\begin{array}{l} \text{Fact} \quad \text{General solution} \\ \text{to} \quad A\bar{x} = \underline{b} \end{array} = \underline{v_p} + \begin{array}{l} \text{General Solution} \\ \text{to} \\ \text{Fixed particular} \\ \text{solution} \end{array} \quad A\bar{x} = \underline{0}$$

$$\Rightarrow \begin{array}{l} \text{Consistent } A\underline{x} = \underline{b} \\ \text{has unique solution} \end{array} \Leftrightarrow \begin{array}{l} A\underline{x} = \underline{0} \\ \text{has unique solution} \end{array} \Leftrightarrow \{ \underline{a}_1, \dots, \underline{a}_n \} \text{ L.I.}$$

Conclusion

$\{\underline{a}_1, \dots, \underline{a}_n\}$ L.I. \Leftrightarrow Reduced A has pivot position in every column

Linear Transformations from \mathbb{R}^n to \mathbb{R}^m

$$T : \mathbb{R}^n \rightarrow \mathbb{R}^m \quad \underline{\text{linear}} \quad \Leftrightarrow \quad \begin{aligned} T(x+y) &= T(x) + T(y) \\ T(\lambda x) &= \lambda T(x) \end{aligned}$$

$$A = (\underline{a}_1 \dots \underline{a}_n) - m \times n \text{ matrix} \Rightarrow$$

$$T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$x \mapsto Ax = x_1 \underline{a}_1 + \cdots + x_n \underline{a}_n$$

Linear

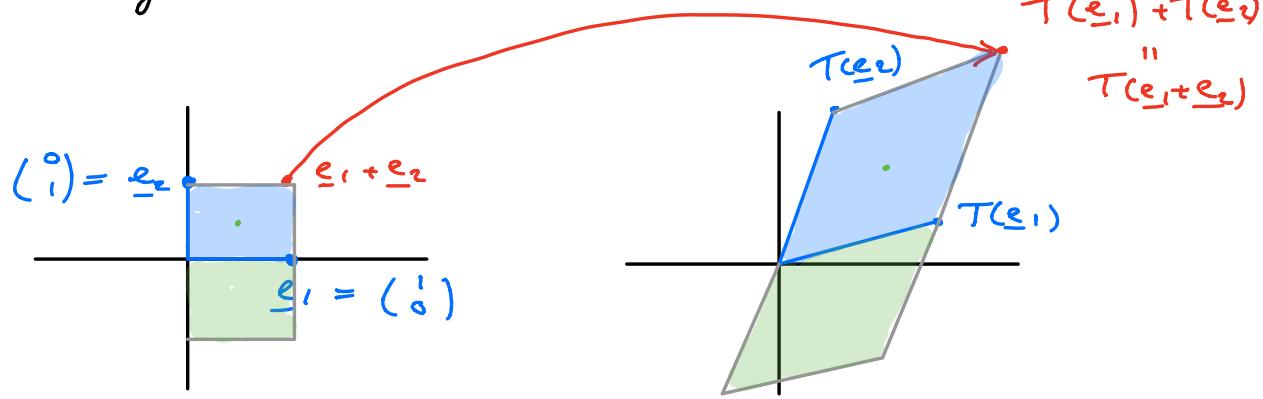
Fact : T linear $\Rightarrow T(\underline{x}) = A\underline{x}$ where

$$A = (T(e_1) \ T(e_2) \ \dots \ T(e_n)) = \text{Standard Matrix of } T$$

$$\underline{e}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \underline{e}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \dots, \underline{e}_n = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}$$

If we know $T(\underline{e}_1), \dots, T(\underline{e}_n)$ we know $T(\underline{x})$ for all \underline{x}

Visualising $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ linear:



$$\text{Range of } T_A = \{A\bar{x}\} = \text{Span}(a_1, \dots, a_n)$$

$$\Rightarrow T_A \text{ onto} \Leftrightarrow \text{Span}(a_1, \dots, a_n) = \mathbb{R}^m$$

T_A one-to-one $\Leftrightarrow T_A(\bar{x}) = \underline{b}$ has at most one solution $\Leftrightarrow A\bar{x} = \underline{b}$ has at most one solution $\Leftrightarrow \{\bar{x}_1, \dots, \bar{x}_n\}$ L.I.

Overview

Linear Transformations

\underline{b} in range of T_A

Vectors

Linear Systems

Reduced Echelon Matrices

$\Leftrightarrow \frac{\underline{b}}{\text{in } \text{Span}(a_1, \dots, a_n)} \Leftrightarrow A\bar{x} = \underline{b}$ has solution \Leftrightarrow Reduced ($A|\underline{b}$) does not have pivot position in last column

T_A onto

$\Leftrightarrow \frac{\text{Span}(a_1, \dots, a_n)}{\mathbb{R}^m} \Leftrightarrow A\bar{x} = \underline{b}$ has solution for all \underline{b}

Pivot position in every row of reduced A

T_A one-to-one

$\Leftrightarrow \frac{\{\bar{x}_1, \dots, \bar{x}_n\}}{\text{L.I.}} \Leftrightarrow A\bar{x} = \underline{b}$ has at most one solution

Pivot position in every column of reduced A .

Matrix Algebra

A, B - $m \times n$ matrices, $\lambda \in \mathbb{R}$

$$(A + B)_{ij} := (A)_{ij} + (B)_{ij} \quad \text{← } ij^{\text{th entry}}$$

$$(\lambda A)_{ij} := \lambda (A)_{ij}$$

A - $m \times n$, B - $n \times p$

$$(AB)_{ij} = (A)_{i1}(B)_{1j} + (A)_{i2}(B)_{2j} + \dots + (A)_{in}(B)_{nj}$$

Key Facts : $T_{(A+B)} = T_A + T_B$

$$T_{(\lambda A)} = \lambda T_A$$

$$T_{AB} = T_A \circ T_B$$

Invertible Matrices

A - $n \times n$ matrix

A invertible \Leftrightarrow There exists B , an $n \times n$ matrix, such that $AB = BA = I_n = n \times n$ identity matrix denoted A^{-1} , the inverse of A

Fact : A invertible $\Leftrightarrow T_A : \mathbb{R}^n \rightarrow \mathbb{R}^n$ one-to-one and onto

\Leftrightarrow Reduced A has pivot position in every column and every row

Crazy coincidence $\Leftrightarrow A$ row equivalent to $\begin{pmatrix} * & * & * \\ 0 & \ddots & 0 \end{pmatrix}$

in $n \times n$ case

: These are equivalent \Leftrightarrow Reduced A equals I_n

Algorithm to calculate A^{-1}

- 1/ Write $(A | I_n)$
- 2/ Put in reduced echelon form $(I_n | B)$ using row operations
- 3/ $B = A^{-1}$.

Determinant

Given $A - n \times n$ matrix, $\det(A)$ is a number which helps to determine invertibility.

Facts / If upper triangular $\Rightarrow \det(U) = \text{product of diagonal entries}$

$$\text{E.g. } \det \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & 3 \\ 0 & 0 & 3 \end{pmatrix} = 1 \cdot 2 \cdot 3 = 6$$

- 1/ If switching two rows multiplies determinant by -1
- 2/ Adding scalar multiple of one row to another does not change determinant
- 3/ Square matrix in echelon form \Rightarrow upper triangular

Algorithm to calculate $\det(A)$

- 1/ Put A in echelon form using only A_j and B_j
- 2/ Compute product of diagonal entries (*This is the determinant*)
- 3/ $\det(A) = (-1)^r \times \text{product of diagonal entries in echelon form}$
 $r = \text{Number of row switches}$

Fact A^{-1} exists \Leftrightarrow $\det(A) \neq 0$