# DM559 Linear and Integer Programming

# Systems of Linear Equations

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Solving Linear Systems

# Outline

1. Solving Linear Systems

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### **Problem Statement**

### Given the system of linear equations:

R1: 
$$\begin{vmatrix} x_1 + x_2 + x_3 = 3 \\ 2x_1 + x_2 + x_3 = 4 \\ R3: \begin{vmatrix} x_1 - x_2 + 2x_3 = 5 \end{vmatrix}$$

Find whether it has any solution and in case characterize the solutions.

# **Augmented Matrix**

### Definition (Augmented Matrix and Elementary row operations)

For a system of linear equations  $A\mathbf{x} = \mathbf{b}$  with

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

the augmented matrix of the system and the row operations are:

$$[A | \mathbf{b}] = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} & b_1 \\ a_{21} & a_{22} & \cdots & a_{2n} & b_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} & b_m \end{bmatrix}$$

RO1: multiply a row by a non-zero constant

RO2: interchange two rows

RO3: add a multiple of one row to another

They modify the linear system into an equivalent system (same solutions)

# Gaussian Elimination: Example

Let's consider the system Ax = b with:

$$[A | \mathbf{b}] = \begin{bmatrix} 1 & 1 & 1 & 3 \\ 2 & 1 & 1 & 4 \\ 1 & -1 & 2 & 5 \end{bmatrix}$$

- 1. Left most column that is not all zeros (it is column 1)
- 2. A non-zero entry at the top of this column (it is the one on the top)
- 3. Make the entry 1 (it is already)

$$\begin{bmatrix}
1 & 1 & 1 & 3 \\
2 & 1 & 1 & 4 \\
1 & -1 & 2 & 5
\end{bmatrix}$$

4. make all entries below the leading one zero:

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# Example, cntd. Row Echelon Form

- 5. Cover up the top row and apply steps (1) and (4) again
- 1. Left most column that is not all zeros is column 2
- 2. Non-zero entry at the top of the column
- 3. Make this entry the leading 1 by elementary row operations RO1 or RO2.
- 4. Make all entries below the leading 1 zero by RO3

$$\begin{bmatrix} \frac{1}{0} & \frac{1}{1} & \frac{1}{3} \\ 0 & -1 & -1 & -2 \\ 0 & -2 & 1 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 1 & 3 \\ 0 & -1 & -1 & -2 \\ 0 & 0 & 3 & 6 \end{bmatrix} \equiv \begin{cases} x_1 + x_2 + x_3 = 3 \\ x_2 + x_3 = 2 \\ x_3 = 2 \end{cases}$$

### Definition (Row echelon form)

A matrix is said to be in row echelon form (or echelon form) if it has the following three properties:

- 1. the first nonzero entry in each nonzero row is 1
- 2. a leading 1 in a lower row is further to the right
- 3. zero rows are at the bottom of the matrix

# Back substitution

$$x_1 + x_2 + x_3 = 3$$
  
 $x_2 + x_3 = 2$   
 $x_3 = 2$ 

From the row echelon form we solve the system by back substitution:

- from the last equation: set  $x_3 = 2$
- substitute  $x_3$  in the second equation  $\rightsquigarrow x_2$
- substitute  $x_2$  and  $x_3$  in the first equation  $\rightsquigarrow x_1$

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### Reduced Row Echelon Form

In the augmented matrix representation:

6. Begin with the last row and add suitable multiples to each row above to get zero above the leading 1.

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

#### Definition (Reduced row echelon form)

A matrix is said to be in reduced (row) echelon form if it has the following properties:

- 1. The matrix is in row echelon form
- 2. Every column with a leading 1 has zeros elsewhere

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From a Reduced Row Echelon Form (RREF) we can read the solution:

$$\begin{bmatrix} A \mid \mathbf{b} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} \rightarrow \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$$

The system has a unique solution. Is it a correct solution? Let's check:

# Gaussian Elimination: Algorithm

Gaussian Elimination algorithm for solving a linear system: (puts the augmented matrix in a form from which the solution can be read)

- 1. Find left most column that is not all zeros
- 2. Get a non-zero entry at the top of this column (pivot element)
- 3. Make this entry 1 by elementary row operations RO1 or RO2. This entry is called leading one
- 4. Add suitable multiples of the top row to rows below so that all entries below the leading one become zero
- 5. Cover up the top row and apply steps (1) and (4) again The matrix left is in (row) echelon form
- 6. Back substitution

# **Gauss-Jordan Reduction**

Gauss Jordan Reduction algorithm for solving a linear system: (puts the augmented matrix in a form from which the solution can be read)

- 1. Find left most column that is not all zeros
- 2. Get a non-zero entry at the top of this column (pivot element)
- 3. Make this entry 1 by elementary row operations RO1 or RO2. This entry is called leading one
- **4**. Add suitable multiples of the top row to rows below so that all entries below the leading one become zero
- 5. Cover up the top row and apply steps (1) and (4) again The matrix left is in (row) echelon form
- 6. Begin with the last row and add suitable multiples to each row above to get zero above the leading 1.

The matrix left is in reduced (row) echelon form

# Will there always be exactly one solution?

R1: 
$$\begin{vmatrix} 2x_3 = 3 \\ R2: & 2x_2 + 3x_3 = 4 \\ x_3 = 5 \end{vmatrix} \rightarrow \begin{bmatrix} A & b \end{bmatrix} = \begin{bmatrix} 0 & 0 & 2 & 3 \\ 0 & 2 & 3 & 4 \\ 0 & 0 & 1 & 5 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & 2 & 3 \\ 0 & 2 & 3 & 4 \\ 0 & 0 & 1 & 5 \end{bmatrix} \rightarrow \begin{array}{c} R2 \\ R1 \\ 0 \\ 0 & 0 & 1 \\ \end{array} \begin{bmatrix} 0 & 2 & 3 & 4 \\ 0 & 0 & 2 & 3 \\ 0 & 0 & 1 & 5 \\ \end{bmatrix} \rightarrow \begin{array}{c} R1/2 \\ 0 & 0 & 2 & 3 \\ 0 & 0 & 1 & 5 \\ \end{bmatrix} \rightarrow$$

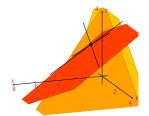
$$\rightarrow -R3/7 \begin{bmatrix} 0 & 1 & \frac{3}{2} & 2\\ 0 & 0 & 1 & 5\\ 0 & 0 & 0 & 1 \end{bmatrix} \qquad \begin{bmatrix} 0 & 1 & \frac{3}{2}\\ 0 & 0 & 1\\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1\\ x_2\\ x_3 \end{bmatrix} = \begin{bmatrix} 2\\ 5\\ 1 \end{bmatrix}$$
 No Solution!

### Definition (Consistent)

A system of linear equations is said to be consistent if it has at least one solution. It is inconsistent if there are no solutions.

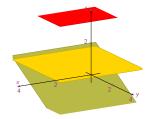
$$\begin{cases} x_1 + x_2 + x_3 = 3 \\ 2x_1 + x_2 + x_3 = 4 \\ x_1 - x_2 + 2x_3 = 5 \end{cases}$$

$$[A | \mathbf{b}] = \begin{bmatrix} 1 & 1 & 1 & 3 \\ 2 & 1 & 1 & 4 \\ 1 & -1 & 2 & 5 \end{bmatrix}$$



$$\begin{cases} 2x_3 = 3\\ 2x_2 + 3x_3 = 4\\ x_3 = 5 \end{cases}$$

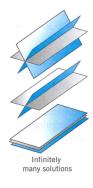
$$[A | \mathbf{b}] = \begin{vmatrix} 0 & 0 & 2 & 3 \\ 0 & 2 & 3 & 4 \\ 0 & 0 & 1 & 5 \end{vmatrix}$$

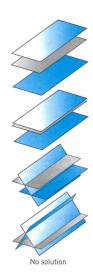


# Geometric Interpretation

### Three equations in three unknowns interpreted as planes in space







#### Definition (Overdetermined)

A linear system is said to be over-determined if there are more equations than unknowns. Over-determined systems are usually (but not always) inconsistent.

#### Definition (Underdetermined)

A linear system of m equations and n unknowns is said to be under-determined if there are fewer equations than unknowns (m < n). They have usually infinitely many solutions (never just one).

# Linear systems with free variables

$$x_1 + x_2 + x_3 + x_4 + x_5 = 3$$
  
 $2x_1 + x_2 + x_3 + x_4 + 2x_5 = 4$   
 $x_1 - x_2 - x_3 + x_4 + x_5 = 5$   
 $x_1 + x_4 + x_5 = 4$ 

$$[A | \mathbf{b}] = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 3 \\ 2 & 1 & 1 & 1 & 2 & 4 \\ 1 & -1 & -1 & 1 & 1 & 5 \\ 1 & 0 & 0 & 1 & 1 & 4 \end{bmatrix}$$

$$\rightarrow (1/2) R3 \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 3 \\ 0 & 1 & 1 & 1 & 0 & 2 \\ 0 & 0 & 0 & 1 & 0 & 3 \\ 0 & 0 & 0 & 1 & 0 & 3 \end{bmatrix}$$

R4-R3

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

$$\xrightarrow{R1-R3} \begin{bmatrix}
1 & 1 & 1 & 0 & 1 & 0 \\
0 & 1 & 1 & 0 & 0 & -1 \\
0 & 0 & 0 & 1 & 0 & 3 \\
0 & 0 & 0 & 0 & 0 & 0
\end{bmatrix}$$

$$\xrightarrow{R1-R2} \begin{bmatrix}
1 & 0 & 0 & 0 & 1 & 1 \\
0 & 0 & 0 & 1 & 1 & 1
\end{bmatrix}$$

$$x_1 + 0 + 0 + 0 + x_5 = 1$$
  
+  $x_2 + x_3 + 0 + 0 = -1$   
+  $x_4 + 0 = 3$ 

### Definition (Leading variables)

The variables corresponding with leading ones in the reduced row echelon form of an augmented matrix are called leading variables. The other variables are called non-leading variables

- $x_1, x_2$  and  $x_4$  are leading variables.
- $x_3, x_5$  are non-leading variables.
- we assign  $x_3, x_5$  the arbitrary values  $s, t \in \mathbb{R}$  and solve for the leading variables.
- there are infinitely many solutions, represented by the general solution:

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} 1 - t \\ -1 - s \\ s \\ 3 \\ t \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 0 \\ 3 \\ 0 \end{bmatrix} + s \begin{bmatrix} 0 \\ -1 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} -1 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

### Solution Sets

#### **Theorem**

A system of linear equations either has no solutions, a unique solution or infinitely many solutions.

#### Proof.

Let's assume the system  $A\mathbf{x} = \mathbf{b}$  has two solutions  $\mathbf{p}$  and  $\mathbf{q}$ . Then all points on the line connecting these two points are also solutions and so there are infinitely many solutions.

$$A\mathbf{p} = \mathbf{b}$$
  $A\mathbf{q} = \mathbf{b}$   $\mathbf{p} \neq \mathbf{q}$ 

$$\mathbf{v} = \mathbf{p} + t(\mathbf{q} - \mathbf{p}), t \in \mathbb{R}$$

$$A\mathbf{v} = A(\mathbf{p} + t(\mathbf{q} - \mathbf{p})) = A\mathbf{p} + t(A\mathbf{q} - A\mathbf{p}) = \mathbf{b} + t(\mathbf{b} - \mathbf{b}) = \mathbf{b}$$

# Homogeneous systems

### Definition (Homogenous system)

An homogeneous system of linear equations is a linear system of the form  $A\mathbf{x} = \mathbf{0}$ .

- A homogeneous system Ax = 0 is always consistent A0 = 0.
- If Ax = 0 has a unique solution, then it must be the trivial solution x = 0.

In the augmented matrix the last column stays always zero  $\leadsto$  we can omit it.

# Example

$$A = \left[ \begin{array}{rrrr} 1 & 1 & 3 & 1 \\ 1 & -1 & 1 & 1 \\ 0 & 1 & 2 & 2 \end{array} \right]$$

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

$$\begin{bmatrix} 1 & 1 & 0 & -5 \\ 0 & 1 & 0 & -2 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1 & 1 & 3 & 1 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 0 & -5 \\ 0 & 1 & 0 & -2 \\ 0 & 0 & 1 & 2 \end{bmatrix}$$

$$\rightarrow \quad \begin{bmatrix} 1 & 0 & 0 & -3 \\ 0 & 1 & 0 & -2 \\ 0 & 0 & 1 & 2 \end{bmatrix} \qquad \mathbf{x} = \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} = t \begin{bmatrix} 3 \\ 2 \\ -2 \\ 1 \end{bmatrix}, t \in \mathbb{R}$$

#### **Theorem**

If A is an  $m \times n$  matrix with m < n, then  $A\mathbf{x} = \mathbf{0}$  has infinitely many solutions.

#### Proof.

- The system is always consistent since homogeneous.
- Matrix A brought in reduced echelon form contains at most m leading ones (variables).
- $n-m \ge 1$  non-leading variables

How about  $A\mathbf{x} = \mathbf{b}$  with A  $m \times n$  and m < n? If the system is consistent, then there are infinitely many solutions.

# Example

$$A\mathbf{x} = \mathbf{0}$$

$$RREF(A)$$

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

$$A\mathbf{x} = \mathbf{b}$$

$$RREF([A|\mathbf{b}])$$

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

$$\mathbf{x} = \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} = t \begin{bmatrix} 3 \\ 2 \\ -2 \\ 1 \end{bmatrix}, t \in \mathbb{R} \qquad \mathbf{x} = \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} 3 \\ 2 \\ -2 \\ 1 \end{bmatrix}, t \in \mathbb{R}$$

### Definition (Associated homogenous system)

Given a system of linear equations,  $A\mathbf{x} = \mathbf{b}$ , the linear system  $A\mathbf{x} = \mathbf{0}$  is called the associated homogeneous system

Eg:

$$RREF(A) = \begin{bmatrix} 1 & 0 & 0 & -3 \\ 0 & 1 & 0 & -2 \\ 0 & 0 & 1 & 2 \end{bmatrix}$$

How can you tell from here that  $A\mathbf{x} = \mathbf{b}$  is consistent with infinitely many solutions?

#### Definition (Null space)

For an  $m \times n$  matrix A, the null space of A is the subset of  $\mathbb{R}^n$  given by

$$N(A) = \{ \mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0} \}$$

where  $\mathbf{0} = (0, 0, \dots, 0)^T$  is the zero vector of  $\mathbb{R}^n$ 

### Theorem (Principle of Linearity)

Suppose that A is an  $m \times n$  matrix, that  $\mathbf{b} \in \mathbb{R}^m$  and that the system  $A\mathbf{x} = \mathbf{b}$  is consistent. Suppose that  $\mathbf{p}$  is any solution of  $A\mathbf{x} = \mathbf{b}$ . Then the set of all solutions of  $A\mathbf{x} = \mathbf{b}$  consists precisely of the vectors  $\mathbf{p} + \mathbf{z}$  for  $\mathbf{z} \in \mathcal{N}(A)$ ; ie,

$$\{x \mid Ax = b\} = \{p + z \mid z \in N(A)\}.$$

#### Proof: We show that

- 1.  $\mathbf{p} + \mathbf{z}$  is a solution for any  $\mathbf{z}$  in the null space of A  $(\{\mathbf{p} + \mathbf{z} \mid \mathbf{z} \in N(A)\} \subseteq \{\mathbf{x} \mid A\mathbf{x} = \mathbf{b}\})$
- 2. all solutions, x, of Ax = b are of the form p + z for some  $z \in N(A)$   $(\{x \mid Ax = b\} \subseteq \{p + z \mid z \in N(A)\})$
- 1. A(p+z) = Ap + Az = b + 0 = b so  $p + z \in \{x \mid Ax = b\}$
- 2. Let x be a solution. Because  ${\bf p}$  is also we have  $A{\bf p}={\bf b}$  and  $A({\bf x}-{\bf p})=A{\bf x}-A{\bf p}={\bf b}-{\bf b}={\bf 0}$  so  ${\bf z}={\bf x}-{\bf p}$  is a solution of  $A{\bf z}={\bf 0}$  and  ${\bf x}={\bf p}+{\bf z}$

(Check validity of the theorem on the previous examples.)

# Summary

• If Ax = b is consistent, the solutions are of the form:

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{solutions of A\mathbf{x} = \mathbf{b}} = \mathbf{p} + {solutions of A\mathbf{x} = \mathbf{0}}
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- if Ax = b has a unique solution, then Ax = 0 has only the trivial solution
- if Ax = b has a infinitely many solutions, then Ax = 0 has infinitely many solutions
- Ax = b may be inconsistent, but Ax = 0 is always consistent.