Applied Linear Algebra in Data Analysis Solution to Linear Equations

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Linear equations

Matrices present a compact way to represent a set of linear equations. Consider the following.

$$\begin{array}{l}
a_{11}x_{1} + a_{12}x_{2} \dots + a_{1m}x_{m} = b_{1} \\
a_{21}x_{1} + a_{22}x_{2} \dots + a_{2m}x_{m} = b_{2} \\
a_{31}x_{1} + a_{32}x_{2} \dots + a_{3m}x_{m} = b_{3} \\
\vdots \\
a_{n1}x_{1} + a_{n2}x_{2} \dots + a_{nm}x_{m} = b_{n}
\end{array}
\right\} \longrightarrow \mathbf{A}\mathbf{x} = \mathbf{b}, \ \mathbf{A} \in \mathbb{R}^{n \times m}, \ \mathbf{x} \in \mathbb{R}^{n}, \ \mathbf{b} \in \mathbb{R}^{m}$$

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1m} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nm} \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

Linear equations in control problems

 \mathbf{x} : Input \mathbf{b} : Output \mathbf{A} : System dynamics

$$\begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1m} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nm} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix}$$

Linear equations in estimation problems

 $\mathbf{x}: \text{Parameter } \mathbf{b}: \text{Measurements } \mathbf{A}: \text{System characteristics}$

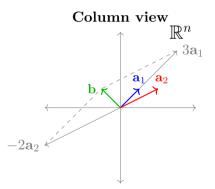
$$\begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1m} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nm} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix}$$

Geometry of linear equations

$$\begin{vmatrix} x_1 + 2x_2 = -1 \\ x_1 + x_2 = 1 \end{vmatrix} \longrightarrow \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

Two ways to view this: **row view** and the **column view**.

Traditional (row) view $x_1 + x_2 = 1$



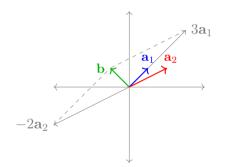
Solutions of linear equations

$$\mathbf{A}\mathbf{x} = \mathbf{b}, \ \mathbf{A} \in \mathbb{R}^{m \times n}, \ \mathbf{x} \in \mathbb{R}^n, \ \mathbf{b} \in \mathbb{R}^m$$

- ► Three possible situations: No solution, Infinitely many solutions, or Unique Solution.
- ▶ When do have infinitely many or no solutions? In \mathbb{R}^3 , we can visualize the different situations.

Understanding Ax = b: Unique solution

$$\begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

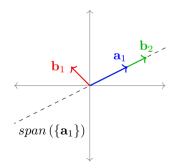


- ► Square matrix
- ▶ Linearly independent set of columns $\{a_1, a_2\}$
- ▶ $\mathbf{b} \in span(\{\mathbf{a}_1, \mathbf{a}_2\}).$
- ► Always solvable, and give an unique solution.

Understanding Ax = b: Unique solution or No solution

1.
$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} \begin{bmatrix} x_1 \end{bmatrix} = \mathbf{b}_1 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$
2.
$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} \begin{bmatrix} x_1 \end{bmatrix} = \mathbf{b}_2 = \begin{bmatrix} 3 \\ 1.5 \end{bmatrix}$$

2.
$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} \begin{bmatrix} x_1 \end{bmatrix} = \mathbf{b}_2 = \begin{bmatrix} 3 \\ 1.5 \end{bmatrix}$$



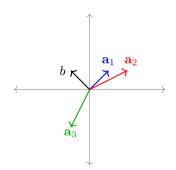
- Tall matrix
- ightharpoonup Linearly independent set of columns $\{a_1\}$

 $\mathbf{b}_1 \notin span(\{\mathbf{a}_1\}) \implies \text{Not solvable.}$

 $\mathbf{b}_2 \in span(\{\mathbf{a}_1\}) \implies$ Solvable with unque solution.

Understanding Ax = b: Infinitely many solution

$$\begin{bmatrix} 1 & 2 & -1 \\ 1 & 1 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$



- ► Fat matrix
- ▶ Linearly dependent set of columns $\{a_1, a_2, a_3\}$
- ightharpoonup **b** $\in span(\{a_1, a_2, a_3\}).$
- ► Always solvable, with infinitely many solutions.

Understanding Ax = b: Conditions for different types of solutions

$$\mathbf{A}\mathbf{x} = \mathbf{b}, \ \mathbf{A} \in \mathbb{R}^{n \times m}, \ \mathbf{x} = \mathbb{R}^m, \ \mathbf{b} \in \mathbb{R}^n$$

Full rank A:

- $ightharpoonup rank(\mathbf{A}) = n \implies \mathbf{Always} \text{ solvable}$
 - $\begin{cases} n = m & \Longrightarrow \text{ Unique solution} \\ n < m & \Longrightarrow \text{ Infinitely many solutions} \end{cases}$
- $ightharpoonup rank(\mathbf{A}) = m \implies \mathbf{No} \text{ infinite solutions}$

$$\begin{cases} m = n & \Longrightarrow \text{ Unique solution} \\ m < n & \to \begin{cases} \mathbf{b} \in span(\mathbf{a}_1, \dots \mathbf{a}_m) \Longrightarrow \text{ Unique solution} \\ \mathbf{b} \notin span(\mathbf{a}_1, \dots \mathbf{a}_m) \Longrightarrow \text{ No solution} \end{cases}$$



Understanding Ax = b: Conditions for different types of solutions

$$\mathbf{A}\mathbf{x} = \mathbf{b}, \ \mathbf{A} \in \mathbb{R}^{n \times m}, \ \mathbf{x} = \mathbb{R}^m, \ \mathbf{b} \in \mathbb{R}^n$$

Rank deficient A:

 $ightharpoonup rank(\mathbf{A}) < \min(n, m) \implies \mathbf{No} \text{ unique solution}$

$$\begin{cases} \mathbf{b} \in span\left(\mathbf{a}_{1}, \dots \mathbf{a}_{m}\right) \Longrightarrow \text{Infinitely many solutions} \\ \mathbf{b} \notin span\left(\mathbf{a}_{1}, \dots \mathbf{a}_{m}\right) \Longrightarrow \text{No solution} \end{cases}$$

Understanding Ax = b: Conditions for different types of solutions

$$\mathbf{A}\mathbf{x} = \mathbf{b}, \ \mathbf{A} \in \mathbb{R}^{n \times m}, \ \mathbf{x} = \mathbb{R}^m, \ \mathbf{b} \in \mathbb{R}^n$$

- ▶ $\mathbf{b} \notin span(\mathbf{a}_1, \dots \mathbf{a}_m) \Longrightarrow No solution$
- ▶ $\mathbf{b} \in span\left(\mathbf{a}_{1}, \dots \mathbf{a}_{m}\right) \Longrightarrow \begin{cases} rank\left(\mathbf{A}\right) = m \implies \text{Unique} \\ rank\left(\mathbf{A}\right) < m \implies \text{Infinitely many solutions} \end{cases}$

General solution of linear equations

$$\mathbf{A}\mathbf{x} = \mathbf{b}, \ \mathbf{A} \in \mathbb{R}^{n \times m}, \ \mathbf{x} = \mathbb{R}^m, \ \mathbf{b} \in \mathbb{R}^n$$

▶ Assuming that this system can be solved, the most general form of the solution is,

$$\mathbf{x} = \mathbf{x}_p + \mathbf{x}_h$$

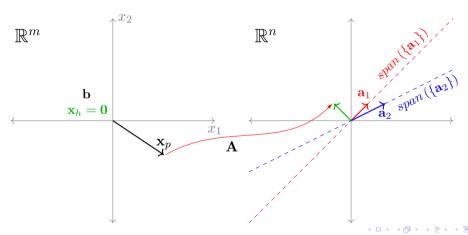
where, \mathbf{x}_p is called the particular solution, and \mathbf{x}_h is the homogenous solution.

- ▶ Homogenous solution: Solution of the equation Ax = 0.
- ▶ The set of all homogenous solutions of $\mathbf{A} \{\mathbf{x}_h \mid \mathbf{A}\mathbf{x}_h = \mathbf{0}\}$ form a subspace of \mathbb{R}^m .

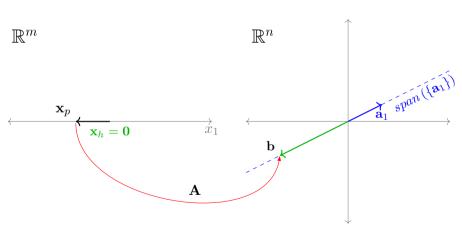
$$\mathbf{A}\mathbf{x} = \mathbf{b}, \ \mathbf{A} \in \mathbb{R}^{n \times m}, \ \mathbf{x} = \mathbb{R}^m, \ \mathbf{b} \in \mathbb{R}^n$$

$$\mathbf{A} = \begin{bmatrix} 1 & -1 \\ 2 & -2 \end{bmatrix}, \ \mathbf{b} = \begin{bmatrix} 1.5 \\ 3 \end{bmatrix}$$

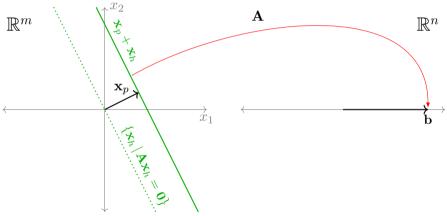
$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix}, \ \mathbf{b} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$



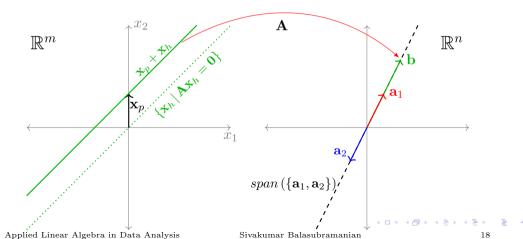
$$\mathbf{A} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \ \mathbf{b} = \begin{bmatrix} -4 \\ -2 \end{bmatrix}$$



$$\mathbf{A} = \begin{bmatrix} 2 & 1 \end{bmatrix}, \ \mathbf{b} = \begin{bmatrix} 5 \end{bmatrix}$$



$$\mathbf{A} = \begin{bmatrix} 1 & -1 \\ 2 & -2 \end{bmatrix}, \ \mathbf{b} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$



Four Fundamental Subspaces of $\mathbf{A} \in \mathbb{R}^{n \times m}$

ightharpoonup C(A): Column Space of A – the span of the columns of A.

$$\mathcal{C}\left(\mathbf{A}\right) = \left\{\mathbf{A}\mathbf{x} \,|\, \mathbf{x} \in \mathbb{R}^{m}\right\} \subseteq \mathbb{R}^{n}$$

▶ $\mathcal{N}(\mathbf{A})$: Nullspace of \mathbf{A} – the set of all $\mathbf{x} \in \mathbb{R}^m$ that are mapped to zero by \mathbf{A} .

$$\mathcal{N}(\mathbf{A}) = \{ \mathbf{x} \mid \mathbf{A}\mathbf{x} = \mathbf{0} \} \subseteq \mathbb{R}^m$$

 $ightharpoonup C(\mathbf{A}^{\top})$: Row Space of \mathbf{A} – the span of the rows of \mathbf{A} .

$$\mathcal{C}\left(\mathbf{A}^{\top}\right) = \left\{\mathbf{A}^{\top}\mathbf{y} \,|\, \mathbf{y} \in \mathbb{R}^{n}\right\} \subseteq \mathbb{R}^{m}$$

▶ $\mathcal{N}(\mathbf{A}^{\top})$: Nullspace of \mathbf{A}^{\top} – the set of all $\mathbf{y} \in \mathbb{R}^n$ that are mapped to zero by \mathbf{A}^{\top} .

$$\mathcal{N}\left(\mathbf{A}^{ op}
ight) = \left\{\mathbf{y} \,|\; \mathbf{A}^{ op}\mathbf{y} = \mathbf{0}
ight\} \subseteq \mathbb{R}^n$$

This is also called the **left nullspace** of **A**.



Linear Independence

- ▶ Given a set of vectors $\{\mathbf{v}_1, \mathbf{v}_2, \dots \mathbf{v}_m\}$, $\mathbf{v}_i \in \mathbb{R}^n$, how can we determine if this set is linear independent?
- We need to verify, $a_1\mathbf{v}_1 + a_2\mathbf{v}_2 + \cdots + a_m\mathbf{v}_m = 0$

$$\begin{bmatrix} \mathbf{v}_1 & \cdots & \mathbf{v}_m \end{bmatrix} \begin{bmatrix} a_1 \\ \vdots \\ a_m \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} = \mathbf{Va} = \mathbf{0} \right\} \mathcal{N}(\mathbf{V}) = \{\mathbf{0}\}, \ rank(\mathbf{V}) = n$$

- ▶ This is also equivalent to saying that when the $rank(\mathbf{A}) = n \implies$ the columns of **A** form an independent set of vectors.
- ▶ When do the rows of **A** form an independent set?
- ▶ What about both rows and columns? When does that happen?

Dimension of the four fundamental subspaces

- ightharpoonup Column space $C(\mathbf{A})$
 - $ightharpoonup dim C(\mathbf{A}) = rank(\mathbf{A}) = r$
- ightharpoonup Nullspace $N(\mathbf{A})$
 - $ightharpoonup \dim N(\mathbf{A}) = n r$
- ▶ Row space $C(\mathbf{A}^{\top})$
- ▶ Left Nullspace $N(\mathbf{A}^{\top})$
 - $ightharpoonup \dim N(\mathbf{A}^{\top}) = m r$