

Math 216 Differential Equations

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Systems of equations as matrix equations

A general first-order linear equation in normal form

$$\begin{aligned}x_1'(t) &= a_{11}(t)x_1 + a_{12}(t)x_2 + \dots + a_{1n}(t)x_n + f_1(t), \\x_2'(t) &= a_{21}(t)x_1 + a_{22}(t)x_2 + \dots + a_{2n}(t)x_n + f_2(t), \\&\vdots \\x_n'(t) &= a_{n1}(t)x_1 + a_{n2}(t)x_2 + \dots + a_{nn}(t)x_n + f_n(t),\end{aligned}$$

can be written as a matrix equation $\mathbf{x}' = \mathbf{A}(t)\mathbf{x} + \mathbf{f}(t)$

$$\begin{bmatrix}x_1' \\x_2' \\ \vdots \\x_n'\end{bmatrix} = \begin{bmatrix}a_{11}(t) & a_{12}(t) & \dots & a_{1n}(t) \\a_{21}(t) & a_{22}(t) & \dots & a_{2n}(t) \\ \vdots & \vdots & & \vdots \\a_{n1}(t) & a_{n2}(t) & \dots & a_{nn}(t)\end{bmatrix} \begin{bmatrix}x_1 \\x_2 \\ \vdots \\x_n\end{bmatrix} + \begin{bmatrix}f_1(t) \\f_2(t) \\ \vdots \\f_n(t)\end{bmatrix}$$

\mathbf{A} is called the coefficient matrix.

Solutions for matrix equations

Let $\mathbf{A} = [a_{ij}(t)]$ be an $n \times n$ matrix and $\mathbf{f}(t) = [f_i(t)]$ be an $n \times 1$ vector.

Def. A solution on interval I to the vector equation

$$\mathbf{x}' = \mathbf{A}(t)\mathbf{x} + \mathbf{f}(t)$$

is an $n \times 1$ vector $\mathbf{b}(t) = [b_i(t)]$ satisfying for all t in I

$$\begin{bmatrix}b_1'(t) \\b_2'(t) \\ \vdots \\b_n'(t)\end{bmatrix} = \begin{bmatrix}a_{11}(t) & a_{12}(t) & \dots & a_{1n}(t) \\a_{21}(t) & a_{22}(t) & \dots & a_{2n}(t) \\ \vdots & \vdots & & \vdots \\a_{n1}(t) & a_{n2}(t) & \dots & a_{nn}(t)\end{bmatrix} \begin{bmatrix}b_1(t) \\b_2(t) \\ \vdots \\b_n(t)\end{bmatrix} + \begin{bmatrix}f_1(t) \\f_2(t) \\ \vdots \\f_n(t)\end{bmatrix}.$$

Solutions for matrix equations

Let $\mathbf{A} = [a_{ij}(t)]$ be an $n \times n$ matrix and $\mathbf{f}(t) = [f_i(t)]$ be an $n \times 1$ vector.

If $\mathbf{b}(t) = [b_i(t)]$ is a solution to the matrix equation

$$\mathbf{b}'(t) = \mathbf{A}(t)\mathbf{b}(t) + \mathbf{f}(t)$$

then $b_1(t), b_2(t), \dots, b_n(t)$ are solutions to the system of first-order equations

$$\begin{aligned}b_1'(t) &= a_{11}(t)b_1 + a_{12}(t)b_2 + \dots + a_{1n}(t)b_n + f_1(t), \\b_2'(t) &= a_{21}(t)b_1 + a_{22}(t)b_2 + \dots + a_{2n}(t)b_n + f_2(t), \\&\vdots \\b_n'(t) &= a_{n1}(t)b_1 + a_{n2}(t)b_2 + \dots + a_{nn}(t)b_n + f_n(t),\end{aligned}$$

Existence and uniqueness

Theorem

Let $\mathbf{A} = [a_{ij}(t)]$ be an $n \times n$ matrix and $\mathbf{f}(t) = [f_i(t)]$ be an $n \times 1$ vector.

Suppose the components $a_{ij}(t)$ and $f_i(t)$ are *continuous on an open interval I containing the point t_0* .

Then for any choice of an initial $n \times 1$ vector \mathbf{b} , there exists a *unique solution $\mathbf{x}(t)$ on the interval I to the initial value problem*

$$\mathbf{x}' = \mathbf{A}(t)\mathbf{x} + \mathbf{f}(t), \quad \mathbf{x}(t_0) = \mathbf{b}.$$

Matrix operations

Equality. Two $m \times n$ matrices $\mathbf{A} = [a_{ij}]$ and $\mathbf{B} = [b_{ij}]$ are equal when they are equal componentwise:

$$\mathbf{A} = \mathbf{B} \quad \text{when } a_{ij} = b_{ij} \text{ for each } 1 \leq i \leq m, 1 \leq j \leq n.$$

Addition. Add matrices componentwise: when \mathbf{A} and \mathbf{B} are $m \times n$ matrices

$$\mathbf{A} + \mathbf{B} = [a_{ij}] + [b_{ij}] = [a_{ij} + b_{ij}]$$

Scalar multiplication. Multiplication of a matrix by a scalar is componentwise.

$$c\mathbf{A} = c[a_{ij}] = [ca_{ij}]$$

Examples

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \mathbf{B} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \quad \mathbf{Z} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

Addition.

$$\mathbf{A} + \mathbf{B} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} + \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 3 \\ 4 & 5 \end{bmatrix}$$

$$\mathbf{A} + \mathbf{Z} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \mathbf{A}$$

Scalar Multiplication.

$$27\mathbf{A} = 27 \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 27 & 54 \\ 81 & 108 \end{bmatrix}$$

Basic properties of addition and scalar multiplication

$\mathbf{0}$ is an $m \times n$ matrix whose entries are all 0.

Let $\mathbf{A}, \mathbf{B}, \mathbf{C}$ be $m \times n$ matrices and c, d are real numbers.

$$\begin{aligned} \mathbf{A} + (\mathbf{B} + \mathbf{C}) &= (\mathbf{A} + \mathbf{B}) + \mathbf{C} & \mathbf{A} + \mathbf{B} &= \mathbf{B} + \mathbf{A} \\ \mathbf{A} + \mathbf{0} &= \mathbf{A} & \mathbf{A} + (-\mathbf{A}) &= \mathbf{0} \\ c(\mathbf{A} + \mathbf{B}) &= c\mathbf{A} + c\mathbf{B} & (c + d)\mathbf{A} &= c\mathbf{A} + d\mathbf{A} \\ c(d\mathbf{A}) &= (cd)\mathbf{A} = d(c\mathbf{A}) \end{aligned}$$

Matrix multiplication


Def. Let $\mathbf{A} = [a_{ik}]$ be an $m \times p$ matrix and $\mathbf{B} = [b_{kj}]$ be an $p \times n$ matrix.

The **product** \mathbf{AB} is the $m \times n$ matrix $\mathbf{C} = [c_{ij}]$ where

$$c_{ij} = \mathbf{a}_i \cdot \mathbf{b}_j \quad (\textit{ith row of } \mathbf{A} \textit{ times } \textit{jth column of } \mathbf{B})$$

$$= [a_{i1} \ a_{i2} \ \dots \ a_{ip}] \begin{bmatrix} b_{1j} \\ b_{2j} \\ \dots \\ b_{pj} \end{bmatrix}$$

$$= a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{ip}b_{pj}$$

 **Warning** . The dimensions of \mathbf{A} , \mathbf{B} and \mathbf{C} are

$$\begin{array}{ccc} \mathbf{A} & \mathbf{B} & \mathbf{C} \\ (m \times p) & (p \times n) & = m \times n \end{array}$$

Matrix multiplication

$$\mathbf{AB} = [a_{ik}] [b_{kj}] = [c_{ij}]$$

where c_{ij} is the product of the i th row of \mathbf{A} and j th column of \mathbf{B} :

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1p} \\ a_{21} & a_{22} & \dots & a_{2p} \\ \vdots & \vdots & & \vdots \\ a_{i1} & a_{i2} & \dots & a_{ip} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mp} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1j} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2j} & \dots & b_{2n} \\ \vdots & \vdots & & \vdots & & \vdots \\ b_{p1} & b_{p2} & \dots & b_{pj} & \dots & b_{pn} \end{bmatrix}$$

Example

Example.

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 4 & 3 \end{bmatrix}$$

and

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 3 & 4 \\ 1 & 2 \end{bmatrix}$$

 **Warning** . Matrix multiplication is not generally **commutative**.


$\mathbf{AB} \neq \mathbf{BA}$ is usually true.

Example

$$[1 \ 2 \ 3 \ 4] \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix} = 5$$

and

$$\begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix} [1 \ 2 \ 3 \ 4] = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

 **Warning** . Order makes a difference.

Basic properties of matrix multiplication


\mathbf{A} , \mathbf{B} , and \mathbf{C} are matrices of the right dimensions, and c is a real number.

$$\begin{aligned}(\mathbf{AB})\mathbf{C} &= \mathbf{A}(\mathbf{BC}) && \text{Associativity} \\(\mathbf{A} + \mathbf{B})\mathbf{C} &= \mathbf{AC} + \mathbf{BC} && \text{Distributivity} \\ \mathbf{A}(\mathbf{B} + \mathbf{C}) &= \mathbf{AB} + \mathbf{AC} && \text{Distributivity} \\ c(\mathbf{A})\mathbf{B} &= c(\mathbf{AB}) = \mathbf{A}(c\mathbf{B}) && \text{Associativity} \\ \mathbf{A}\mathbf{0} &= \mathbf{0} = \mathbf{0A} && \text{Zero}\end{aligned}$$

Example

Example.

$$\begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix} \begin{bmatrix} -2 & 6 \\ 1 & -3 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

 **Warning** . It is possible that $\mathbf{AB} = \mathbf{0}$ with $\mathbf{A} \neq \mathbf{0}$ and $\mathbf{B} \neq \mathbf{0}$.

Matrix Identity

Def. The $n \times n$ **identity matrix** is the square matrix

$$\mathbf{I}_n = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

having ones down the **main diagonal** and zeros everywhere else.

Example.

$$\mathbf{I}_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

We write \mathbf{I} leaving the dimension to be implicitly determined.

Matrix identity is multiplicative identity

Example. \mathbf{I}_3 is a multiplicative identity:

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

If \mathbf{A} is any $n \times n$ matrix and $\mathbf{I} = \mathbf{I}_3$.

$$\mathbf{AI} = \mathbf{A} = \mathbf{IA}.$$

Matrix inverses

Def. A square matrices ($n \times n$) is said to have an **inverse** if there is a square matrix **B** with the property

$$\mathbf{AB} = \mathbf{I}.$$

When this happens it is also true that

- $\mathbf{BA} = \mathbf{I}$, and
- **B** is the **unique** matrix with $\mathbf{AB} = \mathbf{I}$.

We write \mathbf{A}^{-1} for the unique inverse to **A** (when it exists!!)

Def. A matrix is **invertible** (or **nonsingular**) when it has an inverse.

Example


Example.

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

Also, verify that

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

Example

 **Warning** . Not all square matrices are invertible.

Example. Recall,

$$\begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix} \begin{bmatrix} -2 & 6 \\ 1 & -3 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

So, none of these three matrices has inverse.

Having inverses is useful

Fact. If **A** is an invertible matrix, then the matrix equation

$$\mathbf{Ax} = \mathbf{b}$$

always has a **unique solution**, $\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$.

Reason.

$$\mathbf{x} = \mathbf{Ix} = \mathbf{A}^{-1}(\mathbf{Ax}) = \mathbf{A}^{-1}\mathbf{b};$$

the solution is unique since if **x** and **y** are solutions, then

$$\mathbf{Ax} = \mathbf{b} = \mathbf{Ay}$$

so multiply both sides by \mathbf{A}^{-1} to get $\mathbf{x} = \mathbf{y}$.

When do inverses exist?

Question. When does an $n \times n$ matrix have a solution?

Answer. When its **determinant** is nonzero.

(It is more useful to know a matrix is invertible than to compute its inverse.)

 2×2 determinants

♥ Determinants hold the key to the existence of inverses.

Def. For a 2×2 matrix \mathbf{A} , the **determinant of \mathbf{A}** , denoted by $\det(\mathbf{A})$ or $|\mathbf{A}|$ is defined by

$$\det(\mathbf{A}) := \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}.$$

Inverses from determinants

♥ If $\det(\mathbf{A}) \neq 0$ then we can compute \mathbf{A}^{-1} by

$$\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \quad \mathbf{A}^{-1} = \frac{1}{\det(\mathbf{A})} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

Example.

$$\mathbf{A} = \begin{bmatrix} 2 & 1 \\ -1 & 4 \end{bmatrix}, \quad \det(\mathbf{A}) = 9, \quad \mathbf{A}^{-1} = \frac{1}{9} \begin{bmatrix} 4 & 1 \\ -1 & 2 \end{bmatrix}$$

Verify by multiplying the matrices.

 3×3 determinants

Def. For a 3×3 matrix \mathbf{A} , the **determinant of \mathbf{A}** , denoted by $\det(\mathbf{A})$ or $|\mathbf{A}|$ is defined by

Expansion along the first row:

$$\det(\mathbf{A}) := \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

Equivalently, you can compute $\det(\mathbf{A})$ as follows:

Expansion along the first column:

$$\det(\mathbf{A}) := \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{21} \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} + a_{31} \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix}$$

Example of 3×3 determinants

Example.

Expansion along the first row:

$$\begin{vmatrix} 1 & 2 & 1 \\ 0 & 3 & 5 \\ 2 & 1 & -1 \end{vmatrix} = 1 \begin{vmatrix} 3 & 5 \\ 1 & -1 \end{vmatrix} - 2 \begin{vmatrix} 0 & 5 \\ 2 & -1 \end{vmatrix} + 1 \begin{vmatrix} 0 & 3 \\ 2 & 1 \end{vmatrix} \\ = 1(-3 - 5) - 2(0 - 10) + 1(0 - 6) = 6$$

Expansion along the first column:

$$\det(\mathbf{A}) := \begin{vmatrix} 1 & 2 & 1 \\ 0 & 3 & 5 \\ 2 & 1 & -1 \end{vmatrix} = 1 \begin{vmatrix} 3 & 5 \\ 1 & -1 \end{vmatrix} - 0 \begin{vmatrix} 2 & 1 \\ 1 & -1 \end{vmatrix} + 2 \begin{vmatrix} 2 & 1 \\ 3 & 5 \end{vmatrix} \\ = 1(-3 - 5) - 0(-2 - 1) + 2(10 - 3) = 6.$$

Example

Example. Find all solutions to the system of equations

$$\begin{aligned} 8 &= 2x_1 + 4x_2 + x_3 \\ 6 &= 2x_1 + 4x_2 \\ -10 &= -4x_1 - 8x_2 + x_3 \end{aligned}$$

As a matrix equations:

$$\begin{bmatrix} 8 \\ 6 \\ -10 \end{bmatrix} = \begin{bmatrix} 2 & 4 & 1 \\ 2 & 4 & 0 \\ -4 & -8 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

The key to determinants

The determinant for $n \times n$ matrices can be computed by [cofactor expansion](#). See Edwards and Penney, Chapter 5.1, p. 290 (version 3) for details.

♥ The key to determinants.

Theorem

Let \mathbf{A} be an $n \times n$ matrix. The following properties are equivalent.

- 1 $\det(\mathbf{A}) \neq 0$.
- 2 \mathbf{A} has an inverse.
- 3 $\mathbf{Ax} = \mathbf{0}$ has only the trivial solution $\mathbf{x} = \mathbf{0}$.
- 4 $\mathbf{Ax} = \mathbf{b}$ has a unique solution.

So, if $\det(\mathbf{A}) \neq 0$, then \mathbf{A}^{-1} exists and

$$\mathbf{Ax} = \mathbf{b} \quad \text{has the unique solution } \mathbf{x} = \mathbf{A}^{-1}\mathbf{b}.$$

Example continued

Compute the determinant

$$\begin{vmatrix} 2 & 4 & 1 \\ 2 & 4 & 0 \\ -4 & -8 & 1 \end{vmatrix} = 2 \begin{vmatrix} 4 & 0 \\ -8 & 1 \end{vmatrix} - 2 \begin{vmatrix} 4 & 1 \\ -8 & 1 \end{vmatrix} - 4 \begin{vmatrix} 4 & 0 \\ 4 & 1 \end{vmatrix} \\ = 2(4 - 0) - 2(4 + 8) - 4(0 - 4) \\ = 0.$$

So, we know the matrix equation does not have a [unique solution](#).

$$\begin{bmatrix} 8 \\ 6 \\ -10 \end{bmatrix} = \begin{bmatrix} 2 & 4 & 1 \\ 2 & 4 & 0 \\ -4 & -8 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

Example continued

Solve the system of equations.

$$\begin{aligned} 8 &= 2x_1 + 4x_2 + x_3 \\ 6 &= 2x_1 + 4x_2 \\ -10 &= -4x_1 - 8x_2 + x_3 \end{aligned}$$

Eliminate x_1 from row 2 (R_2) and row 3 R_3 , isolating x_1 to row 1 (R_1).

R_2 : by adding $-1 \times R_1$ to R_2 :

$$\begin{aligned} 8 &= 2x_1 + 4x_2 + x_3 \\ -2 &= 0x_1 + 0x_2 - x_3 \\ -10 &= -4x_1 - 8x_2 + x_3 \end{aligned}$$

R_3 : by adding $2 \times R_1$ to R_3 :

$$\begin{aligned} 8 &= 2x_1 + 4x_2 + x_3 \\ -2 &= 0x_1 + 0x_2 - x_3 \\ 6 &= 0x_1 + 0x_2 + 3x_3 \end{aligned}$$

Note that the solutions to this new system of equations is the same as the

Example continued

Solve the system of equations.

$$\begin{aligned} 8 &= 2x_1 + 4x_2 + x_3 \\ -2 &= -x_3 \\ 6 &= 3x_3 \end{aligned}$$

Eliminate x_3 from row 1 and row 3, isolating x_3 to row 2.

R_1 : by adding R_2 to R_1 :

$$\begin{aligned} 6 &= 2x_1 + 4x_2 + 0x_3 \\ -2 &= x_3 \\ 6 &= 3x_3 \end{aligned}$$

R_3 : by adding $3 \times R_2$ to R_3 :

$$\begin{aligned} 8 &= 2x_1 + 4x_2 + 0x_3 \\ -2 &= -x_3 \\ 0 &= 0x_3 \end{aligned}$$

Note that the solutions to this new system of equations is the same as the

Example continued

Solve the system of equations.

$$\begin{aligned} 8 &= 2x_1 + 4x_2 \\ -2 &= -x_3 \\ 0 &= 0 \end{aligned}$$

Read off the solutions:

$$\begin{aligned} 4 - 2x_2 &= x_1 \\ 2 &= x_3 \end{aligned}$$

There are **no constraints** placed on x_2 , so we are free to let it be **any value**.

The following are the solutions to the original equation:

$$x_1 = 4 - 2c, \quad x_2 = c, \quad x_3 = 2 \quad \text{for each real value } c$$

There are **infinitely many solution**.