

# MT-104

## Linear Algebra

National University of Computer and Emerging Sciences

Fall 2020

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# **Systems of Linear Differential Equations**

# First order linear homogenous differential equation

First order differential Equation

$$x' = kx, \quad k \text{ is a constant.}$$

Solution:  $x(t) = x_0 e^{kt}$ .

First order differential system

$$x' = 4x$$

$$y' = 9y$$

In Matrix form of above system can be written as

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 4 & 0 \\ 0 & 9 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}.$$

- ▶ Matrix is diagonal
- ▶ System is uncoupled

# First order linear system of differential equations

General first order linear system

$$x' = ax + by$$

$$y' = cx + dy$$

In Matrix form

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}.$$

We can write the above system as

$$X' = AX,$$

where  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$  and  $X = \begin{bmatrix} x \\ y \end{bmatrix}$ .

- ▶  $A$  is not a diagonal matrix.
- ▶ Can we diagonalize  $A$ ?

## Diagonalization

*Aim :*

To solve the system  $X' = AX$ .

*Challenges*

Matrix is not diagonal.

*Possible Solution*

Transform the matrix into a diagonal matrix i.e., diagonalize it.

# HOW?

We want to transform

$$X' = AX \xrightarrow{\text{to}} Y' = DY.$$

## Diagonalization

As  $PDP^{-1} = A$ , so we can write

$$X' = AX = PDP^{-1}X$$

Pre multiplying by  $P^{-1}$  we get

$$P^{-1}X' = DP^{-1}X$$

Since,  $P$  is a constant matrix, so

$$\left(P^{-1}X\right)' = D\left(P^{-1}X\right).$$

Put  $(P^{-1}X) = Y$  to get

$$\boxed{Y' = DY}.$$

# Uncoupling system of differential equations

## Summary

Coupled system of differential equation

$$X' = AX$$

can be transformed (uncoupled) to

$$Y' = DY$$

by using the transformation

$$X = PY.$$

## Example

Find a solution to the system

$$x' = x + 3y$$

$$y' = 2x + 2y$$

subject to initial conditions  $x(0) = 0$ ,  $y(0) = 5$ .

**Solution:** In Matrix form, we can write it as

$$X' = AX,$$

where  $A = \begin{bmatrix} 1 & 3 \\ 2 & 2 \end{bmatrix}$  and  $X = \begin{bmatrix} x \\ y \end{bmatrix}$ .

We can uncoupled the system by using the transformation

$$X = PY.$$



## Example

### Eigenvalues:

Characteristic Equation:

$$\lambda^2 - 3\lambda - 4 = 0.$$

Eigenvalues are:  $-1, 4$ .

Corresponding eigenvectors are

$$\begin{bmatrix} 3 \\ -2 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

Hence,

$$P = \begin{bmatrix} 3 & 1 \\ -2 & 1 \end{bmatrix}, D = \begin{bmatrix} -1 & 0 \\ 0 & 4 \end{bmatrix}.$$

By using the transformation  $X = PY$ , we get

$$Y' = DY,$$

where  $Y = \begin{bmatrix} u \\ v \end{bmatrix}$ .

## Example

$$\begin{bmatrix} u' \\ v' \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix}.$$

Above equations can be written as

$$u' = -u$$

$$v' = 4v.$$

Solving, we get

$$u = c_1 e^{-t}, \quad v = c_2 e^{4t}.$$

As  $X = PY$ , so, we can write

$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 3 & 1 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix}.$$

$$x = 3u + v$$

$$y = -2u + v.$$

Substituting values of  $u$  and  $v$ , we get

$$x = 3c_1 e^{-t} + c_2 e^{4t}$$

$$y = -2c_1 e^{-t} + c_2 e^{4t}.$$

## Example

Since,  $x(0) = 0$  and  $y(0) = 5$ , we get

$$0 = 3c_1 + c_2$$

$$5 = -2c_1 + c_2.$$

Solving, above system we get

$$c_1 = -1, c_2 = 3.$$

In matrix form we can the solution as

$$X = -x_1 e^{-t} + 3x_2 e^{4t}$$

where  $x_1 = \begin{bmatrix} 3 \\ -2 \end{bmatrix}$  and  $x_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  are the eigenvectors corresponding to eigenvalues  $-1$  and  $4$  respectively.

## Example

Find a solution to the system

$$r'(t) = w(t) - 12$$

$$w'(t) = -r(t) + 10$$

**Solution:**

*Issue :*

Presence of  $-12$  and  $10$ .

*How to resolve it :*

Put  $w(t) - 12 = y(t)$  and  $-r(t) + 10 = x(t)$ , we get

$$-x'(t) = y(t)$$

$$y'(t) = x(t).$$

In Matrix form, we can write it as

$$X' = AX,$$

where  $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$  and  $X = \begin{bmatrix} x \\ y \end{bmatrix}$ .

By using the substitution  $X = PY$  we get

$$Y' = DY$$

where  $P = \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix}$ ,  $D = \begin{bmatrix} -i & 0 \\ 0 & i \end{bmatrix}$ .

So, solution is

$$u = c_1 e^{-it}$$

$$v = c_2 e^{it}$$

By using the relation  $X = PY$ , we get

$$x = c_1 e^{-it} + c_2 e^{it}$$

$$y = c_1 i e^{-it} - i c_2 e^{it}.$$

Solution of the system is

$$r(t) = 10 - c_1 e^{-it} - c_2 e^{it}$$

$$w(t) = 12 + c_1 i e^{-it} - i c_2 e^{it}$$

- In case of single linear differential equation, we have

$$x' = kx, \text{ } k \text{ is a constant.}$$

Solution of the differential equation is

$$x = ce^{kt}.$$

- In case of system of coupled differential equations, we have

$$X' = AX, \text{ } A \text{ is a constant matrix.}$$

Solution of the linear differential system should be

$$X = c e^{At}.$$

## Exponential of a Matrix

Compute  $e^{Dt}$  where  $D = \begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix}$ .

Since,  $e^x = 1 + x + \frac{x^2}{2!} + \dots = \sum_{n=0}^{\infty} \frac{x^n}{n!}$ . so,

$$e^{Dt} = I + Dt + \dots = \sum_{n=0}^{\infty} \frac{D^n t^n}{n!}$$

$$e^{Dt} = \sum_{n=0}^{\infty} \frac{1}{n!} \begin{bmatrix} 4^n t^n & 0 \\ 0 & t^n \end{bmatrix}$$

$$e^{Dt} = \begin{bmatrix} \sum_{n=0}^{\infty} \frac{4^n}{n!} & 0 \\ 0 & \sum_{n=0}^{\infty} \frac{1^n}{n!} \end{bmatrix}$$

$$e^{Dt} = \begin{bmatrix} e^{4t} & 0 \\ 0 & e^t \end{bmatrix}$$

## Example

Compute  $e^{At}$  where  $A = \begin{bmatrix} 1 & 3 \\ 2 & 2 \end{bmatrix}$

For given matrix, we have

$$P = \begin{bmatrix} 3 & 1 \\ -1 & 1 \end{bmatrix}, D = \begin{bmatrix} -1 & 0 \\ 0 & 4 \end{bmatrix}.$$

$$e^{At} = \sum_{n=0}^{\infty} \frac{A^n t^n}{n!}$$

$$e^{At} = \sum_{n=0}^{\infty} \frac{PD^n P^{-1}}{n!}$$

$$e^{At} = P \sum_{n=0}^{\infty} \frac{D^n t^n}{n!} P^{-1}$$

$$e^{At} = P e^{Dt} P^{-1}$$

$$e^{At} = \begin{bmatrix} 3 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} e^{-t} & 0 \\ 0 & e^{4t} \end{bmatrix} \left( \begin{bmatrix} 3 & 1 \\ -1 & 1 \end{bmatrix} \right)^{-1}.$$



## Chapter # 6

## Motivation

**Recall:** This course is about learning to:

- ▶ Solve the matrix equation  $Ax = b$  (Echelon Form, Reduced Echelon Form, Column Space, Null Space)
- ▶ Solve the matrix equation  $Ax = \lambda x$  (Eigenvalues, eigenvectors and their applications)
- ▶ Almost solve the equation  $Ax = b$  (We are going to study)

**Idea 1:** In the real world, data is imperfect.

Suppose you measure a data point  $x$  which you know for theoretical reasons must lie on a plane spanned by two vectors  $u$  and  $v$ . Due to measurement error, though, the measured  $x$  is not actually in  $\text{Span}\{u, v\}$ .

In other words, the equation  $au + bv = x$  has no solution ( $x \in \text{Span of } u \text{ and } v$ ).

What do you do?

The real value is probably the *closest* point to  $x$  on  $\text{Span}\{u, v\}$ .

Which point is that?

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## Motivation

**Idea 2:** The cost of producing  $t$  books like this one is nearly linear,  $b = C + Dt$ , with editing and typesetting in  $C$  and then printing and binding in  $D$ .  $C$  is the set-up cost and  $D$  is the cost for each additional book. How to compute  $C$  and  $D$ ? If there is no experimental error, then two measurements of  $b$  will determine the line  $b = C + Dt$ .

But if there is **error**, we must be prepared to “average” the experiments and find an optimal line. Since there are two unknowns  $C$  and  $D$  to be determined, we now project onto a two-dimensional subspace. A perfect experiment would give a perfect  $C$  and  $D$ :

$$C + Dt_1 = b_1$$

$$C + Dt_2 = b_2$$

.....

$$C + Dt_m = b_m :$$

This is an overdetermined system, with  $m$  equations and only two unknowns. If errors are present, it will have no solution. The best solution  $(\hat{C}, \hat{D})$  is the  $\hat{x}$  that minimizes the squared error  $E^2$

## The Dot Product

**Purpose** To find *length* and *angle* between two vectors, and in particular, to understand the notion of *orthogonality* (i.e. perpendicularity).

### Definition

The **dot product** of two vectors  $x, y$  in  $\mathbb{R}^n$  is

$$x \cdot y = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \cdot \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} := x_1 y_1 + x_2 y_2 + \cdots + x_n y_n = x^T y.$$

### Example

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \cdot \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} = (1 \quad 2 \quad 3) \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} = 1 \cdot 4 + 2 \cdot 5 + 3 \cdot 6 = 32.$$

## Properties of the Dot Product

- ▶  $x \cdot x \geq 0$
- ▶  $x \cdot x = 0$  if and only if  $x = 0$ .
- ▶  $x \cdot y = y \cdot x$
- ▶  $(x + y) \cdot z = x \cdot z + y \cdot z$
- ▶  $(cx) \cdot y = c(x \cdot y)$

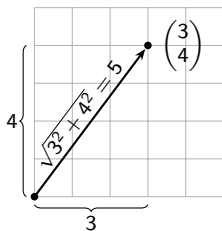
# The Dot Product and Length

## Definition

The **length** or **norm** of a vector  $x$  in  $\mathbb{R}^n$  is

$$\|x\| = \sqrt{x \cdot x} = \sqrt{x_1^2 + x_2^2 + \cdots + x_n^2}.$$

The Pythagorean theorem!



$$\left\| \begin{pmatrix} 3 \\ 4 \end{pmatrix} \right\| = \sqrt{3^2 + 4^2} = 5$$

## Fact

If  $x$  is a vector and  $c$  is a scalar, then  $\|cx\| = |c| \cdot \|x\|$ .



# The Dot Product and Distance

## Definition

The **distance** between two points  $x, y$  in  $\mathbb{R}^n$  is

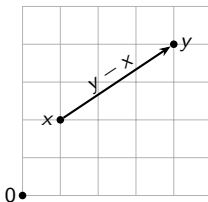
$$\text{dist}(x, y) = \|y - x\|.$$

This is just the length of the vector from  $x$  to  $y$ .

## Example

Let  $x = (1, 2)$  and  $y = (4, 4)$ . Then

$$\text{dist}(x, y) = \|y - x\| = \left\| \begin{pmatrix} 3 \\ 2 \end{pmatrix} \right\| = \sqrt{3^2 + 2^2} = \sqrt{13}.$$



# Unit Vectors

## Definition

A **unit vector** is a vector  $v$  with length  $\|v\| = 1$ .

## Example

The unit coordinate vectors are unit vectors:

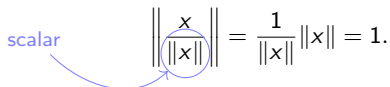
$$\|e_1\| = \left\| \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right\| = \sqrt{1^2 + 0^2 + 0^2} = 1$$

## Definition

Let  $x$  be a nonzero vector in  $\mathbb{R}^n$ . The **unit vector in the direction of  $x$**  is the vector  $\frac{x}{\|x\|}$ .

This is in fact a unit vector:

scalar

$$\left\| \frac{x}{\|x\|} \right\| = \frac{1}{\|x\|} \|x\| = 1.$$


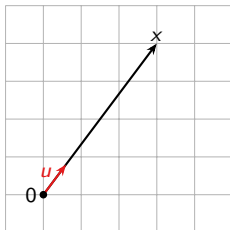
# Unit Vectors

## Example

### Example

What is the unit vector in the direction of  $x = \begin{pmatrix} 3 \\ 4 \end{pmatrix}$ ?

$$u = \frac{x}{\|x\|} = \frac{1}{\sqrt{3^2 + 4^2}} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 3 \\ 4 \end{pmatrix}.$$



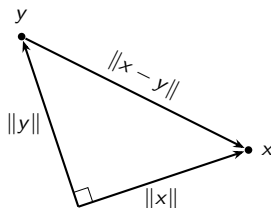
# Orthogonality

## Definition

Two vectors  $x, y$  are **orthogonal** or **perpendicular** if  $x \cdot y = 0$ .

[3] Why is this a good definition?

The Pythagorean theorem / law of cosines!



$x$  and  $y$  are  
perpendicular

$$\iff \|x\|^2 + \|y\|^2 = \|x - y\|^2$$

$$\iff x \cdot x + y \cdot y = (x - y) \cdot (x - y)$$

$$\iff x \cdot x + y \cdot y = x \cdot x + y \cdot y - 2x \cdot y$$

$$\iff x \cdot y = 0$$

**Fact:**  $x \perp y \iff \|x - y\|^2 = \|x\|^2 + \|y\|^2$

# Orthogonality

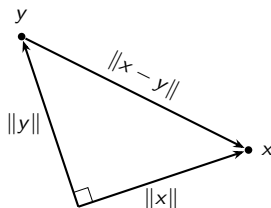
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*Notation:*  $x \perp y$  means  $x \cdot y = 0$ .

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$$\iff x \cdot y = 0$$

**Fact:**  $x \perp y \iff \|x - y\|^2 = \|x\|^2 + \|y\|^2$

## Orthogonality

### Definition

A set of vectors  $\{x_1, x_2, \dots, x_k\}$  in  $R^n$  is called an **orthogonal set** if  $x_i \cdot x_j = 0$  whenever  $i \neq j$  for  $i, j = 1, 2, \dots, k$ .

**Problem:** Show that  $\{x_1, x_2, x_3\}$  is an orthogonal set in  $R^3$  if

$$x_1 = \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix}, \quad x_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, \quad x_3 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}.$$

$$x_1 \cdot x_2 = 2(0) + 1(1) + (-1)(1) = 0$$

$$x_2 \cdot x_3 = 0(1) + 1(-1) + (1)(1) = 0$$

$$x_1 \cdot x_3 = 2(1) + 1(-1) + (-1)(1) = 0$$

# Orthogonality

## Example

**Problem:** Find *all* vectors orthogonal to  $v = \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix}$ .

We have to find all vectors  $x$  such that  $x \cdot v = 0$ . This means solving the equation

$$0 = x \cdot v = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix} = x_1 + x_2 - x_3.$$

The parametric form for the solution is  $x_1 = -x_2 + x_3$ , so the parametric vector form of the general solution is

$$x = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = x_2 \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} + x_3 \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}.$$

For instance,  $\begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} \perp \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix}$  because  $\begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix} = 0$ .

# Orthogonality

## Example

**Problem:** Find *all* vectors orthogonal to both  $v = \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix}$  and  $w = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ .

Now we have to solve the system of two homogeneous equations

$$0 = x \cdot v = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix} = x_1 + x_2 - x_3$$

$$0 = x \cdot w = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = x_1 + x_2 + x_3.$$

In matrix form:

The rows are  $v$  and  $w \longrightarrow \begin{pmatrix} 1 & 1 & -1 \\ 1 & 1 & 1 \end{pmatrix} \xrightarrow{\text{rref}} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$

The parametric vector form of the solution is

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = x_2 \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}.$$



## Orthogonal Basis

### Theorem

If  $\{x_1, x_2, \dots, x_k\}$  is an orthogonal set of nonzero vectors in  $R^n$ , then these vectors are linearly independent.

Suppose  $\{u_1, u_2, \dots, u_m\}$  is orthogonal. We need to show that the equation

$$c_1 u_1 + c_2 u_2 + \dots + c_m u_m = 0$$

has only the trivial solution  $c_1 = c_2 = \dots = c_m = 0$ .

$$0 = u_1 \cdot (c_1 u_1 + c_2 u_2 + \dots + c_m u_m) = c_1(u_1 \cdot u_1) + 0 + 0 + \dots + 0.$$

Hence  $c_1 = 0$ . Similarly for the other  $c_i$ .

### Definition

An **orthogonal basis** for a subspace  $W$  of  $R^n$  is a basis of  $W$  that is an orthogonal set.

**Problem** Show that  $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$  forms orthogonal basis of  $R^2$ .

# Orthogonal Basis

## Example

**Problem** Find an orthogonal basis for the subspace  $W$  of  $\mathbb{R}^3$  given by

$$W = \left\{ \begin{bmatrix} x \\ y \\ z \end{bmatrix} : x - y + 2z = 0 \right\}.$$

We have already studied how to calculate the basis of  $W$ .

$$\begin{bmatrix} y - 2z \\ y \\ z \end{bmatrix} = y \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + z \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix},$$

so,  $\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$  forms basis of  $W$

They are not orthogonal.

## Orthogonal Basis

We want to find a vector of  $W$  that is orthogonal to either  $\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$  or  $\begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$ .

Let  $w = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$  be a vector of  $W$  and is orthogonal to  $\begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$ .

As  $w \in W$ , so we can write

$$x - y + 2z = 0.$$

Orthogonality condition implies

$$-2x + z = 0.$$

Solving above two equations, we get

$$y = 5x, \quad z = 2x.$$

So,

$$w = x \begin{bmatrix} 1 \\ 5 \\ 2 \end{bmatrix}.$$

So,  $\begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 5 \\ 2 \end{bmatrix}$  forms orthogonal basis of  $W$ .

## Orthogonal Basis

### Theorem

Let  $\{x_1, x_2, \dots, x_k\}$  be an orthogonal basis for a subspace  $W$  of  $\mathbb{R}^n$  and let  $w$  be any vector in  $W$ . Then the unique scalars  $c_1, c_2, \dots, c_k$  such that

$$w = c_1 x_1 + c_2 x_2 + \dots + c_k x_k$$

are given by

$$c_i = \frac{w \cdot v_i}{v_i \cdot v_i}, \quad i = 1, 2, \dots, k.$$

## Orthogonal Basis

### Example

**Problem** Show that  $B = \left\{ v_1 = \begin{bmatrix} 4 \\ -2 \end{bmatrix}, v_2 = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right\}$  forms basis of  $\mathbb{R}^2$  and write coordinate vector of  $w = \begin{bmatrix} 1 \\ -3 \end{bmatrix}$  w.r.t  $B$ .

As  $v_1 \cdot v_2 = (4) \cdot (1) + (2)(-2) = 0$ , so,  $v_1$  and  $v_2$  are orthogonal and hence forms basis of  $\mathbb{R}^2$ .

Now, for the coordinate vector

$$w = c_1 v_1 + c_2 v_2.$$

Taking dot product with  $v_1$ , we get

$$\begin{aligned} w \cdot v_1 &= c_1 v_1 \cdot v_1 \\ c_1 &= \frac{w \cdot v_1}{v_1 \cdot v_1} \end{aligned}$$

$$\text{Similarly, } c_2 = \frac{w \cdot v_2}{v_2 \cdot v_2}.$$

As  $v_1 \cdot v_1 = 12$ ,  $v_2 \cdot v_2 = 5$ ,  $w \cdot v_1 = 10$ ,  $w \cdot v_2 = -5$ . So,  $c_1 = \frac{5}{6}$ ,  $c_2 = -1$ . Hence, coordinate vector of  $w$  is

$$\begin{bmatrix} \frac{5}{6} \\ -1 \end{bmatrix}.$$