

## Module I: Introduction

**Remark I.0.1** This brief module gives an overview for the course.

## Section I.0

### Remark I.0.1 What is Linear Algebra?

Linear algebra is the study of **linear maps**.

- In Calculus, you learn how to approximate any function by a linear function.
- In Linear Algebra, we learn about how linear maps behave.
- Combining the two, we can approximate how any function behaves.

### Remark I.0.2 What is Linear Algebra good for?

- Linear algebra is used throughout several fields in higher mathematics.
- In computer graphics, linear algebra is used to help represent 3D objects in a 2D grid of pixels.
- Linear algebra is used to approximate differential equation solutions in a vast number of engineering applications (e.g. fluid flows, vibrations, heat transfer) whose solutions are very difficult (or impossible) to find precisely.
- Google's search engine is based on its Page Rank algorithm, which ranks websites by computing an eigenvector of a matrix.

### Remark I.0.3 What will I learn in this class?

By the end of this class, you will be able to:

- Solve systems of linear equations. (Module E)
- Identify vector spaces and their properties. (Module V)
- Analyze the structure of vector spaces and sets of vectors. (Module S)
- Use and apply the algebraic properties of linear transformations. (Module A)
- Perform fundamental operations in the algebra of matrices. (Module M)
- Use and apply the geometric properties of linear transformations. (Module G)

## Module E: Solving Systems of Linear Equations

### Standards for this Module

**How can we solve systems of linear equations?** At the end of this module, students will be able to...

**E1. Systems as matrices.** ... translate back and forth between a system of linear equations and the corresponding augmented matrix.

**E2. Row reduction.** ... put a matrix in reduced row echelon form.

**E3. Systems of linear equations.** ... compute the solution set for a system of linear equations.

### Readiness Assurance Outcomes

Before beginning this module, each student should be able to...

- Determine if a system to a two-variable system of linear equations will have zero, one, or infinitely-many solutions by graphing.
- Find the unique solution to a two-variable system of linear equations by back-substitution.
- Describe sets using set-builder notation, and check if an element is a member of a set described by set-builder notation.

### Readiness Assurance Resources

The following resources will help you prepare for this module.

- Systems of linear equations (Khan Academy): <http://bit.ly/2l21etm>
- Solving linear systems with substitution (Khan Academy): <http://bit.ly/1SlMpix>
- Set builder notation: <https://youtu.be/xnfUZ-NTsCE>

## Section E.0

**Definition E.0.1** A **linear equation** is an equation of the variables  $x_i$  of the form

$$a_1x_1 + a_2x_2 + \cdots + a_nx_n = b.$$

A **solution** for a linear equation is a Euclidean vector

$$\begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix}$$

that satisfies

$$a_1s_1 + a_2s_2 + \cdots + a_ns_n = b$$

(that is, a Euclidean vector that can be plugged into the equation).

**Remark E.0.2** In previous classes you likely used the variables  $x, y, z$  in equations. However, since this course often deals with equations of four or more variables, we will often write our variables as  $x_i$ , and assume  $x = x_1, y = x_2, z = x_3, w = x_4$  when convenient.

**Definition E.0.3** A **system of linear equations** (or a **linear system** for short) is a collection of one or more linear equations.

$$\begin{array}{ccccccc} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n & = & b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n & = & b_2 \\ \vdots & & \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n & = & b_m \end{array}$$

Its **solution set** is given by

$$\left\{ \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} \mid \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} \text{ is a solution to all equations in the system} \right\}.$$

**Remark E.0.4** When variables in a large linear system are missing, we prefer to write the system in one of the following standard forms:

Original linear system:

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

Verbose standard form:

$$\begin{array}{rcl} 1x_1 + 0x_2 + 3x_3 & = & 3 \\ 3x_1 - 2x_2 + 4x_3 & = & 0 \\ 0x_1 - 1x_2 + 1x_3 & = & -2 \end{array}$$

Concise standard form:

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

**Definition E.0.5** A linear system is **consistent** if its solution set is non-empty (that is, there exists a solution for the system). Otherwise it is **inconsistent**.

**Fact E.0.6** All linear systems are one of the following:

- **Consistent with one solution:** its solution set contains a single vector, e.g.  $\left\{ \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\}$
- **Consistent with infinitely-many solutions:** its solution set contains infinitely many vectors, e.g.  $\left\{ \begin{bmatrix} 1 \\ 2-3a \\ a \end{bmatrix} \middle| a \in \mathbb{R} \right\}$
- **Inconsistent:** its solution set is the empty set  $\{\} = \emptyset$

**Activity E.0.7** (*~10 min*) All inconsistent linear systems contain a logical **contradiction**. Find a contradiction in this system to show that its solution set is  $\emptyset$ .

$$\begin{aligned} -x_1 + 2x_2 &= 5 \\ 2x_1 - 4x_2 &= 6 \end{aligned}$$

**Activity E.0.8** (*~10 min*) Consider the following consistent linear system.

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

*Part 1:* Find three different solutions for this system.

*Part 2:* Let  $x_2 = a$  where  $a$  is an arbitrary real number, then find an expression for  $x_1$  in terms of  $a$ . Use this to write the solution set  $\left\{ \begin{bmatrix} ? \\ a \end{bmatrix} \middle| a \in \mathbb{R} \right\}$  for the linear system.

**Activity E.0.9** (*~10 min*) Consider the following linear system.

$$\begin{aligned} x_1 + 2x_2 - x_4 &= 3 \\ x_3 + 4x_4 &= -2 \end{aligned}$$

Describe the solution set

$$\left\{ \begin{bmatrix} ? \\ a \\ ? \\ b \end{bmatrix} \middle| a, b \in \mathbb{R} \right\}$$

to the linear system by setting  $x_2 = a$  and  $x_4 = b$ , and then solving for  $x_1$  and  $x_3$ .

**Observation E.0.10** Solving linear systems of two variables by graphing or substitution is reasonable for two-variable systems, but these simple techniques won't usually cut it for equations with more than two variables or more than two equations. For example,

$$-2x_1 - 4x_2 + x_3 - 4x_4 = -8$$

$$x_1 + 2x_2 + 2x_3 + 12x_4 = -1$$

$$x_1 + 2x_2 + x_3 + 8x_4 = 1$$

has the exact same solution set as the system in the previous activity, but we'll want to learn new techniques to compute these solutions efficiently.

## Section E.1

**Remark E.1.1** The only important information in a linear system are its coefficients and constants.

Original linear system:

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

Verbose standard form:

$$\begin{aligned}1x_1 + 0x_2 + 3x_3 &= 3 \\3x_1 - 2x_2 + 4x_3 &= 0 \\0x_1 - 1x_2 + 1x_3 &= -2\end{aligned}$$

Coefficients/constants:

$$\begin{array}{ccc|c}1 & 0 & 3 & 3 \\3 & -2 & 4 & 0 \\0 & -1 & 1 & -2\end{array}$$

**Definition E.1.2** A system of  $m$  linear equations with  $n$  variables is often represented by writing its coefficients and constants in an **augmented matrix**.

$$\begin{array}{l}a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\\vdots \qquad \qquad \qquad \vdots \qquad \qquad \qquad \vdots \qquad \qquad \qquad \vdots \\a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m\end{array} \qquad \left[ \begin{array}{cccc|c}a_{11} & a_{12} & \cdots & a_{1n} & b_1 \\a_{21} & a_{22} & \cdots & a_{2n} & b_2 \\\vdots & \vdots & \ddots & \vdots & \vdots \\a_{m1} & a_{m2} & \cdots & a_{mn} & b_m\end{array} \right]$$

**Example E.1.3** The corresponding augmented matrix for this system is obtained by simply writing the coefficients and constants in matrix form.

Linear system:

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

Augmented matrix:

$$\left[ \begin{array}{ccc|c}1 & 0 & 3 & 3 \\3 & -2 & 4 & 0 \\0 & -1 & 1 & -2\end{array} \right]$$

**Definition E.1.4** Two systems of linear equations (and their corresponding augmented matrices) are said to be **equivalent** if they have the same solution set.

For example, both of these systems share the same solution set  $\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$ .

$$\begin{aligned}3x_1 - 2x_2 &= 1 \\x_1 + 4x_2 &= 5\end{aligned}$$

$$\begin{aligned}3x_1 - 2x_2 &= 1 \\4x_1 + 2x_2 &= 6\end{aligned}$$

Therefore these augmented matrices are equivalent:

$$\left[ \begin{array}{cc|c}3 & -2 & 1 \\1 & 4 & 5\end{array} \right]$$

$$\left[ \begin{array}{cc|c}3 & -2 & 1 \\4 & 2 & 6\end{array} \right]$$

**Activity E.1.5** ( $\sim 10$  min) Following are seven procedures used to manipulate an augmented matrix. Label the procedures that would result in an equivalent augmented matrix as **valid**, and label the procedures that might change the solution set of the corresponding linear system as **invalid**.

- |   |   |
|---|---|
| a) Swap two rows.                         | e) Add a constant multiple of one row to another row. |
| b) Swap two columns.                      |   |
| c) Add a constant to every term in a row. | f) Replace a column with zeros.                       |
| d) Multiply a row by a nonzero constant.  | g) Replace a row with zeros.                          |

**(Instructor Note:)** This activity could be ran as a card sort. Allow 5 additional minutes for intra team discussion.

**Definition E.1.6** The following **row operations** produce equivalent augmented matrices:

1. Swap two rows.
2. Multiply a row by a nonzero constant.
3. Add a constant multiple of one row to another row.

Whenever two matrices  $A, B$  are equivalent (so whenever we do any of these operations), we write  $A \sim B$ .

**Activity E.1.7** ( $\sim 10$  min) Consider the following (equivalent) linear systems.

(A)	(C)	(E)
$-2x_1 + 4x_2 - 2x_3 = -8$	$x_1 - 2x_2 + 2x_3 = 7$	$x_1 - 2x_2 = 1$
$x_1 - 2x_2 + 2x_3 = 7$	$2x_3 = 6$	$x_3 = 3$
$3x_1 - 6x_2 + 4x_3 = 15$	$-2x_3 = -6$	$0 = 0$
(B)	(D)	(F)
$x_1 - 2x_2 + 2x_3 = 7$	$x_1 - 2x_2 + 2x_3 = 7$	$x_1 - 2x_2 + 2x_3 = 7$
$-2x_1 + 4x_2 - 2x_3 = -8$	$x_3 = 3$	$2x_3 = 6$
$3x_1 - 6x_2 + 4x_3 = 15$	$-2x_3 = -6$	$3x_1 - 6x_2 + 4x_3 = 15$

*Part 1:* Find a solution to one of these systems.

*Part 2:* Rank the six linear systems from most complicated to simplest.



**Activity E.1.8** ( $\sim 5$  min) We can rewrite the previous in terms of equivalences of augmented matrices

$$\begin{aligned} \left[ \begin{array}{ccc|c} -2 & 4 & -2 & -8 \\ 1 & -2 & 2 & 7 \\ 3 & -6 & 4 & 15 \end{array} \right] &\sim \left[ \begin{array}{ccc|c} \textcircled{1} & -2 & 2 & 7 \\ -2 & 4 & -2 & -8 \\ 3 & -6 & 4 & 15 \end{array} \right] \sim \left[ \begin{array}{ccc|c} \textcircled{1} & -2 & 2 & 7 \\ 0 & 0 & 2 & 6 \\ 3 & -6 & 4 & 15 \end{array} \right] \\ &\sim \left[ \begin{array}{ccc|c} \textcircled{1} & -2 & 2 & 7 \\ 0 & 0 & 2 & 6 \\ 0 & 0 & -2 & -6 \end{array} \right] \sim \left[ \begin{array}{ccc|c} \textcircled{1} & -2 & 2 & 7 \\ 0 & 0 & \textcircled{1} & 3 \\ 0 & 0 & -2 & -6 \end{array} \right] \sim \left[ \begin{array}{ccc|c} \textcircled{1} & -2 & 0 & 1 \\ 0 & 0 & \textcircled{1} & 3 \\ 0 & 0 & 0 & 0 \end{array} \right] \end{aligned}$$

Determine the row operation(s) necessary in each step to transform the most complicated system's augmented matrix into the simplest.

**Activity E.1.9** ( $\sim 10$  min) A matrix is in **reduced row echelon form (RREF)** if

1. The leading term (first nonzero term) of each nonzero row is a 1. Call these terms **pivots**.
2. Each pivot is to the right of every higher pivot.
3. Each term above or below a pivot is zero.
4. All rows of zeroes are at the bottom of the matrix.

Circle the leading terms in each example, and label it as RREF or not RREF.

<p>(A) <math>\left[ \begin{array}{ccc c} 1 &amp; 0 &amp; 0 &amp; 3 \\ 0 &amp; 0 &amp; 1 &amp; -1 \\ 0 &amp; 0 &amp; 0 &amp; 0 \end{array} \right]</math></p>	<p>(C) <math>\left[ \begin{array}{ccc c} 0 &amp; 0 &amp; 0 &amp; 0 \\ 1 &amp; 2 &amp; 0 &amp; 3 \\ 0 &amp; 0 &amp; 1 &amp; -1 \end{array} \right]</math></p>	<p>(E) <math>\left[ \begin{array}{ccc c} 0 &amp; 1 &amp; 0 &amp; 7 \\ 1 &amp; 0 &amp; 0 &amp; 4 \\ 0 &amp; 0 &amp; 0 &amp; 0 \end{array} \right]</math></p>
<p>(B) <math>\left[ \begin{array}{ccc c} 1 &amp; 2 &amp; 4 &amp; 3 \\ 0 &amp; 0 &amp; 1 &amp; -1 \\ 0 &amp; 0 &amp; 0 &amp; 0 \end{array} \right]</math></p>	<p>(D) <math>\left[ \begin{array}{ccc c} 1 &amp; 0 &amp; 2 &amp; -3 \\ 0 &amp; 3 &amp; 3 &amp; -3 \\ 0 &amp; 0 &amp; 0 &amp; 0 \end{array} \right]</math></p>	<p>(F) <math>\left[ \begin{array}{ccc c} 1 &amp; 0 &amp; 0 &amp; 4 \\ 0 &amp; 1 &amp; 0 &amp; 7 \\ 0 &amp; 0 &amp; 1 &amp; 0 \end{array} \right]</math></p>

**Remark E.1.10** It is important to understand the **Gauss-Jordan elimination** algorithm that converts a matrix into reduced row echelon form.

A video outlining how to perform the Gauss-Jordan Elimination algorithm by hand is available at <https://youtu.be/Cq0Nxx2dhhU>. Practicing several exercises outside of class using this method is recommended.

In the next section, we will learn to use technology to perform this operation for us, as will be expected when applying row-reduced matrices to solve other problems.

## Section E.2

**Activity E.2.1** (*~10 min*) Free browser-based technologies for mathematical computation are available online.

- Go to <http://cocalc.com> and create an account.
- Create a project titled “Linear Algebra Team X” with your appropriate team number. Add all team members as collaborators.
- Open the project and click on “New”
- Give it an appropriate name such as “Class E.2 workbook”. Make a new Jupyter notebook.
- Click on “Kernel” and make sure “Octave” is selected.
- Type `A=[1 3 4 ; 2 5 7]` and press **Shift+Enter** to store the matrix  $\begin{bmatrix} 1 & 3 & 4 \\ 2 & 5 & 7 \end{bmatrix}$  in the variable  $A$ .
- Type `rref(A)` and press **Shift+Enter** to compute the reduced row echelon form of  $A$ .

**Remark E.2.2** If you need to find the reduced row echelon form of a matrix during class, you are encouraged to use CoCalc’s Octave interpreter.

You can change a cell from “Code” to “Markdown” or “Raw” to put comments around your calculations such as Activity numbers.

**Activity E.2.3** (*~10 min*) Consider the system of equations.

$$\begin{aligned} 3x_1 - 2x_2 + 13x_3 &= 6 \\ 2x_1 - 2x_2 + 10x_3 &= 2 \\ -x_1 + 3x_2 - 6x_3 &= 11 \end{aligned}$$

Convert this to an augmented matrix and use CoCalc to compute its reduced row echelon form. Write these on your whiteboard, and use them to write a simpler yet equivalent linear system of equations. Then find its solution set.

**Activity E.2.4** (*~10 min*) Consider our system of equations from above.

$$\begin{aligned} 3x_1 - 2x_2 + 13x_3 &= 6 \\ 2x_1 - 2x_2 + 10x_3 &= 2 \\ -x_1 &\quad - 3x_3 = 1 \end{aligned}$$

Convert this to an augmented matrix and use CoCalc to compute its reduced row echelon form. Write these on your whiteboard, and use them to write a simpler yet equivalent linear system of equations. Then find its solution set.

**Activity E.2.5** (*~10 min*) Consider the following linear system.

$$\begin{aligned}x_1 + 2x_2 + 3x_3 &= 1 \\2x_1 + 4x_2 + 8x_3 &= 0\end{aligned}$$

*Part 1:* Find its corresponding augmented matrix  $A$  and use CoCalc to find  $\text{RREF}(A)$ .

*Part 2:* How many solutions does the corresponding linear system have?

**Activity E.2.6** (*~10 min*) Consider the simple linear system equivalent to the system from the previous problem:

$$\begin{aligned}x_1 + 2x_2 &= 4 \\x_3 &= -1\end{aligned}$$

*Part 1:* Let  $x_1 = a$  and write the solution set in the form  $\left\{ \begin{bmatrix} a \\ ? \\ ? \end{bmatrix} \mid a \in \mathbb{R} \right\}$ .

*Part 2:* Let  $x_2 = b$  and write the solution set in the form  $\left\{ \begin{bmatrix} ? \\ b \\ ? \end{bmatrix} \mid b \in \mathbb{R} \right\}$ .

*Part 3:* Which of these was easier? What features of the RREF matrix  $\left[ \begin{array}{ccc|c} \textcircled{1} & 2 & 0 & 4 \\ 0 & 0 & \textcircled{1} & -1 \end{array} \right]$  caused this?

**Definition E.2.7** Recall that the pivots of a matrix in RREF form are the leading 1s in each non-zero row.

The pivot columns in an augmented matrix correspond to the **bound variables** in the system of equations ( $x_1, x_3$  below). The remaining variables are called **free variables** ( $x_2$  below).

$$\left[ \begin{array}{ccc|c} \textcircled{1} & 2 & 0 & 4 \\ 0 & 0 & \textcircled{1} & -1 \end{array} \right]$$

To efficiently solve a system in RREF form, we may assign letters to free variables and solve for the bound variables.

**Activity E.2.8** (*~10 min*) Find the solution set for the system

$$\begin{aligned}2x_1 - 2x_2 - 6x_3 + x_4 - x_5 &= 3 \\-x_1 + x_2 + 3x_3 - x_4 + 2x_5 &= -3 \\x_1 - 2x_2 - x_3 + x_4 + x_5 &= 2\end{aligned}$$

by row-reducing its augmented matrix, and then assigning letters to the free variables (given by non-pivot columns) and solving for the bound variables (given by pivot columns) in the corresponding linear system.

**Observation E.2.9** The solution set to the system

$$\begin{aligned} 2x_1 - 2x_2 - 6x_3 + x_4 - x_5 &= 3 \\ -x_1 + x_2 + 3x_3 - x_4 + 2x_5 &= -3 \\ x_1 - 2x_2 - x_3 + x_4 + x_5 &= 2 \end{aligned}$$

may be written as

$$\left\{ \left[ \begin{array}{c} 1 + 5a + 2b \\ 1 + 2a + 3b \\ a \\ 3 + 3b \\ b \end{array} \right] \middle| a, b \in \mathbb{R} \right\}.$$

**Remark E.2.10** Don't forget to correctly express the solution set of a linear system, using set-builder notation for consistent systems with infinitely many solutions.

- **Consistent with one solution:** e.g.  $\left\{ \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\}$
- **Consistent with infinitely-many solutions:** e.g.  $\left\{ \left[ \begin{array}{c} 1 \\ 2 - 3a \\ a \end{array} \right] \middle| a \in \mathbb{R} \right\}$
- **Inconsistent:**  $\emptyset$

## Module V: Vector Spaces

### Standards for this Module

**What is a vector space?** At the end of this module, students will be able to...

- V1. Vector property verification.** ... show why an example satisfies a given vector space property, but does not satisfy another given property.
- V2. Vector space identification.** ... list the eight defining properties of a vector space, infer which of these properties a given example satisfies, and thus determine if the example is a vector space.
- V3. Linear combinations.** ... determine if a Euclidean vector can be written as a linear combination of a given set of Euclidean vectors.
- V4. Spanning sets.** ... determine if a set of Euclidean vectors spans  $\mathbb{R}^n$ .
- V5. Subspaces.** ... determine if a subset of  $\mathbb{R}^n$  is a subspace or not.

### Readiness Assurance Outcomes

Before beginning this module, each student should be able to...

- Add Euclidean vectors and multiply Euclidean vectors by scalars.
- Add complex numbers and multiply complex numbers by scalars.
- Add polynomials and multiply polynomials by scalars.
- Perform basic manipulations of augmented matrices and linear systems **E1,E2,E3**.

### Readiness Assurance Resources

The following resources will help you prepare for this module.

- Adding and subtracting Euclidean vectors (Khan Academy): <http://bit.ly/2y8A0wa>
- Linear combinations of Euclidean vectors (Khan Academy): <http://bit.ly/2nK3wne>
- Adding and subtracting complex numbers (Khan Academy): <http://bit.ly/1PE3ZMQ>
- Adding and subtracting polynomials (Khan Academy): <http://bit.ly/2d5SLGZ>

## Section V.0

**Activity V.0.1** (*~20 min*) Consider each of the following vector properties. Label each property with  $\mathbb{R}^1$ ,  $\mathbb{R}^2$ , and/or  $\mathbb{R}^3$  if that property holds for Euclidean vectors/scalars  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  of that dimension.

1. **Addition associativity.**

$$\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}.$$

2. **Addition commutativity.**

$$\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}.$$

3. **Addition identity.**

There exists some  $\mathbf{z}$  where  $\mathbf{v} + \mathbf{z} = \mathbf{v}$ .

4. **Addition inverse.**

There exists some  $-\mathbf{v}$  where  $\mathbf{v} + (-\mathbf{v}) = \mathbf{z}$ .

5. **Addition midpoint uniqueness.**

There exists a unique  $\mathbf{m}$  where the distance from  $\mathbf{u}$  to  $\mathbf{m}$  equals the distance from  $\mathbf{m}$  to  $\mathbf{v}$ .

6. **Scalar multiplication associativity.**

$$a(b\mathbf{v}) = (ab)\mathbf{v}.$$

7. **Scalar multiplication identity.**

$$1\mathbf{v} = \mathbf{v}.$$

8. **Scalar multiplication relativity.**

There exists some scalar  $c$  where either  $c\mathbf{v} = \mathbf{w}$  or  $c\mathbf{w} = \mathbf{v}$ .

9. **Scalar distribution.**

$$a(\mathbf{u} + \mathbf{v}) = a\mathbf{u} + a\mathbf{v}.$$

10. **Vector distribution.**

$$(a + b)\mathbf{v} = a\mathbf{v} + b\mathbf{v}.$$

11. **Orthogonality.**

There exists a non-zero vector  $\mathbf{n}$  such that  $\mathbf{n}$  is orthogonal to both  $\mathbf{u}$  and  $\mathbf{v}$ .

12. **Bidimensionality.**

$$\mathbf{v} = a\mathbf{i} + b\mathbf{j} \text{ for some value of } a, b.$$

**Definition V.0.2** A **vector space**  $V$  is any collection of mathematical objects with associated addition and scalar multiplication operations that satisfy the following properties. Let  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  belong to  $V$ , and let  $a, b$  be scalar numbers.

- **Addition is associative.**

$$\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}.$$

- **Addition is commutative.**

$$\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}.$$

- **Additive identity exists.**

There exists some  $\mathbf{z}$  where  $\mathbf{v} + \mathbf{z} = \mathbf{v}$ .

- **Additive inverses exist.**

There exists some  $-\mathbf{v}$  where  $\mathbf{v} + (-\mathbf{v}) = \mathbf{z}$ .

- **Scalar multiplication is associative.**

$$a(b\mathbf{v}) = (ab)\mathbf{v}.$$

- **1 is a scalar multiplicative identity.**

$$1\mathbf{v} = \mathbf{v}.$$

- **Scalar multiplication distributes over vector addition.**

$$a(\mathbf{u} + \mathbf{v}) = a\mathbf{u} + a\mathbf{v}.$$

- **Scalar multiplication distributes over scalar addition.**

$$(a + b)\mathbf{v} = a\mathbf{v} + b\mathbf{v}.$$

Any **Euclidean vector space**  $\mathbb{R}^n$  satisfies all eight requirements regardless of the value of  $n$ , but we will also study other types of vector spaces.

## Section V.1

**Remark V.1.1** Last time, we defined a **vector space**  $V$  to be any collection of mathematical objects with associated addition and scalar multiplication operations that satisfy the following eight properties for all  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  in  $V$ , and all scalars (i.e. real numbers)  $a, b$ .

- |   |   |
|---|---|
| <ul style="list-style-type: none"> <li>• <b>Addition is associative.</b><br/><math>\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}.</math></li> <li>• <b>Addition is commutative.</b><br/><math>\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}.</math></li> <li>• <b>Additive identity exists.</b><br/>There exists some <math>\mathbf{z}</math> where <math>\mathbf{v} + \mathbf{z} = \mathbf{v}.</math></li> <li>• <b>Additive inverses exist.</b><br/>There exists some <math>-\mathbf{v}</math> where <math>\mathbf{v} + (-\mathbf{v}) = \mathbf{z}.</math></li> </ul> | <ul style="list-style-type: none"> <li>• <b>Scalar multiplication is associative.</b><br/><math>a(b\mathbf{v}) = (ab)\mathbf{v}.</math></li> <li>• <b>1 is a scalar multiplicative identity.</b><br/><math>1\mathbf{v} = \mathbf{v}.</math></li> <li>• <b>Scalar multiplication distributes over vector addition.</b><br/><math>a(\mathbf{u} + \mathbf{v}) = a\mathbf{u} + a\mathbf{v}.</math></li> <li>• <b>Scalar multiplication distributes over scalar addition.</b><br/><math>(a + b)\mathbf{v} = a\mathbf{v} + b\mathbf{v}.</math></li> </ul> |
|---|---|

**Remark V.1.2** The following sets are examples of vector spaces, with the usual/natural operations for addition and scalar multiplication.

- $\mathbb{R}^n$ : Euclidean vectors with  $n$  components.
- $\mathbb{R}^\infty$ : Sequences of real numbers  $(v_1, v_2, \dots)$ .
- $M_{m,n}$ : Matrices of real numbers with  $m$  rows and  $n$  columns.
- $\mathbb{C}$ : Complex numbers.
- $\mathcal{P}^n$ : Polynomials of degree  $n$  or less.
- $\mathcal{P}$ : Polynomials of any degree.
- $C(\mathbb{R})$ : Real-valued continuous functions.

**Activity V.1.3** ( $\sim 20$  min) Consider the set  $V = \{(x, y) \mid y = e^x\}$  with operations defined by

$$(x, y) \oplus (z, w) = (x + z, yw) \quad c \odot (x, y) = (cx, y^c)$$

*Part 1:* Show that  $V$  satisfies the vector distributive property

$$(a + b) \odot \mathbf{v} = (a \odot \mathbf{v}) \oplus (b \odot \mathbf{v})$$

by letting  $\mathbf{v} = (x, y)$  and showing both sides simplify to the same expression.

*Part 2:* Show that  $V$  contains an additive identity element by choosing  $\mathbf{z} = (?, ?)$  such that  $\mathbf{v} \oplus \mathbf{z} = (x, y) \oplus (?, ?) = \mathbf{v}$  for any  $\mathbf{v} = (x, y) \in V$ .

**Remark V.1.4** It turns out  $V = \{(x, y) \mid y = e^x\}$  with operations defined by

$$(x, y) \oplus (z, w) = (x + z, yw) \quad c \odot (x, y) = (cx, y^c)$$

satisfies all eight properties.

- |   |  |
|---|--|
| <ul style="list-style-type: none"> <li>• <b>Addition associativity.</b><br/><math>\mathbf{u} \oplus (\mathbf{v} \oplus \mathbf{w}) = (\mathbf{u} \oplus \mathbf{v}) \oplus \mathbf{w}.</math></li> <li>• <b>Addition commutivity.</b><br/><math>\mathbf{u} \oplus \mathbf{v} = \mathbf{v} \oplus \mathbf{u}.</math></li> <li>• <b>Addition identity.</b><br/>There exists some <math>\mathbf{z}</math> where <math>\mathbf{v} \oplus \mathbf{z} = \mathbf{v}.</math></li> <li>• <b>Addition inverse.</b><br/>There exists some <math>-\mathbf{v}</math> where <math>\mathbf{v} \oplus (-\mathbf{v}) = \mathbf{z}.</math></li> </ul> | <ul style="list-style-type: none"> <li>• <b>Scalar multiplication associativity.</b><br/><math>a \odot (b \odot \mathbf{v}) = (ab) \odot \mathbf{v}.</math></li> <li>• <b>Scalar multiplication identity.</b><br/><math>1 \odot \mathbf{v} = \mathbf{v}.</math></li> <li>• <b>Scalar distribution.</b><br/><math>a \odot (\mathbf{u} \oplus \mathbf{v}) = (a \odot \mathbf{u}) \oplus (a \odot \mathbf{v}).</math></li> <li>• <b>Vector distribution.</b><br/><math>(a + b) \odot \mathbf{v} = (a \odot \mathbf{v}) \oplus (b \odot \mathbf{v}).</math></li> </ul> |
|---|--|

Thus,  $V$  is a vector space.

**Activity V.1.5** ( $\sim 15$  min) Let  $V = \{(x, y) \mid x, y \in \mathbb{R}\}$  have operations defined by

$$(x, y) \oplus (z, w) = (x + y + z + w, x^2 + z^2) \quad c \odot (x, y) = (x^c, y + c - 1).$$

*Part 1:* Show that the scalar multiplication identity holds by simplifying  $1 \odot (x, y)$  to  $(x, y)$ .

*Part 2:* Show that the addition identity property fails by showing that  $(0, -1) \oplus \mathbf{z} \neq (0, -1)$  no matter how  $\mathbf{z} = (z_1, z_2)$  is chosen.

*Part 3:* Can  $V$  be a vector space?



**Definition V.1.6** A **linear combination** of a set of vectors  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$  is given by  $c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_m\mathbf{v}_m$  for any choice of scalar multiples  $c_1, c_2, \dots, c_m$ .

For example, we can say  $\begin{bmatrix} 3 \\ 0 \\ 5 \end{bmatrix}$  is a linear combination of the vectors  $\begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix}$  and  $\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$  since

$$\begin{bmatrix} 3 \\ 0 \\ 5 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} + 1 \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

**Definition V.1.7** The **span** of a set of vectors is the collection of all linear combinations of that set:

$$\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\} = \{c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_m\mathbf{v}_m \mid c_i \in \mathbb{R}\}.$$

For example:

$$\text{span}\left\{\begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}\right\} = \left\{a \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} + b \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \mid a, b \in \mathbb{R}\right\}$$

**Activity V.1.8** ( $\sim 10$  min) Consider  $\text{span}\left\{\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right\}$ .

*Part 1:* Sketch  $1 \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,  $3 \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,  $0 \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ , and  $-2 \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  in the  $xy$  plane.

*Part 2:* Sketch a representation of all the vectors belonging to  $\text{span}\left\{\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right\} = \left\{a \begin{bmatrix} 1 \\ 2 \end{bmatrix} \mid a \in \mathbb{R}\right\}$  in the  $xy$  plane.

**Activity V.1.9** ( $\sim 10$  min) Consider  $\text{span}\left\{\begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix}\right\}$ .

*Part 1:* Sketch the following linear combinations in the  $xy$  plane.

$$\begin{array}{ccc} 1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 0 \begin{bmatrix} -1 \\ 1 \end{bmatrix} & 0 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 1 \begin{bmatrix} -1 \\ 1 \end{bmatrix} & 1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 1 \begin{bmatrix} -1 \\ 1 \end{bmatrix} \\ -2 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 1 \begin{bmatrix} -1 \\ 1 \end{bmatrix} & -1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + -2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} & \end{array}$$

*Part 2:* Sketch a representation of all the vectors belonging to  $\text{span}\left\{\begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix}\right\}$  in the  $xy$  plane.

**Activity V.1.10** ( $\sim 5$  min) Sketch a representation of all the vectors belonging to  $\text{span}\left\{\begin{bmatrix} 6 \\ -4 \end{bmatrix}, \begin{bmatrix} -3 \\ 2 \end{bmatrix}\right\}$  in the  $xy$  plane.

## Section V.2

**Remark V.2.1** Recall these definitions from last class:

- A **linear combination** of vectors is given by adding scalar multiples of those vectors, such as:

$$\begin{bmatrix} 3 \\ 0 \\ 5 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} + 1 \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

- The **span** of a set of vectors is the collection of all linear combinations of that set, such as:

$$\text{span} \left\{ \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \right\} = \left\{ a \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} + b \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \mid a, b \in \mathbb{R} \right\}$$

**Activity V.2.2** (*~15 min*) The vector  $\begin{bmatrix} -1 \\ -6 \\ 1 \end{bmatrix}$  belongs to  $\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix}, \begin{bmatrix} -1 \\ -3 \\ 2 \end{bmatrix} \right\}$  exactly when there exists a solution to the vector equation  $x_1 \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix} + x_2 \begin{bmatrix} -1 \\ -3 \\ 2 \end{bmatrix} = \begin{bmatrix} -1 \\ -6 \\ 1 \end{bmatrix}$ .

*Part 1:* Reinterpret this vector equation as a system of linear equations.

*Part 2:* Find its solution set, using CoCalc.com to find RREF of its corresponding augmented matrix.

*Part 3:* Given this solution set, does  $\begin{bmatrix} -1 \\ -6 \\ 1 \end{bmatrix}$  belong to  $\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix}, \begin{bmatrix} -1 \\ -3 \\ 2 \end{bmatrix} \right\}$ ?

**Fact V.2.3** A vector  $\mathbf{b}$  belongs to  $\text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  if and only if the linear system corresponding to  $[\mathbf{v}_1 \dots \mathbf{v}_n \mid \mathbf{b}]$  is consistent.

Put another way,  $\mathbf{b}$  belongs to  $\text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  exactly when  $\text{RREF}[\mathbf{v}_1 \dots \mathbf{v}_n \mid \mathbf{b}]$  doesn't have a row  $[0 \dots 0 \mid 1]$  representing the contradiction  $0 = 1$ .

**Activity V.2.4** (*~10 min*) Determine if  $\begin{bmatrix} 3 \\ -2 \\ 1 \\ 5 \end{bmatrix}$  belongs to  $\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ -3 \\ 2 \end{bmatrix}, \begin{bmatrix} -1 \\ -3 \\ 2 \\ 2 \end{bmatrix} \right\}$  by row-reducing an appropriate matrix.

**Activity V.2.5** (*~5 min*) Determine if  $\begin{bmatrix} -1 \\ -9 \\ 0 \end{bmatrix}$  belongs to  $\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix}, \begin{bmatrix} -1 \\ -3 \\ 2 \end{bmatrix} \right\}$  by row-reducing an appropriate matrix.

**Activity V.2.6** ( $\sim 10$  min) Does the third-degree polynomial  $3y^3 - 2y^2 + y + 5$  in  $\mathcal{P}^3$  belong to  $\text{span}\{y^3 - 3y + 2, -y^3 - 3y^2 + 2y + 2\}$ ?

*Part 1:* Reinterpret this question as an equivalent exercise involving Euclidean vectors in  $\mathbb{R}^4$ . (Hint: What four numbers must you know to write a  $\mathcal{P}^3$  polynomial?)

*Part 2:* Solve this equivalent exercise, and use its solution to answer the original question.

**Activity V.2.7** ( $\sim 5$  min) Does the matrix  $\begin{bmatrix} 3 & -2 \\ 1 & 5 \end{bmatrix}$  belong to  $\text{span}\left\{\begin{bmatrix} 1 & 0 \\ -3 & 2 \end{bmatrix}, \begin{bmatrix} -1 & -3 \\ 2 & 2 \end{bmatrix}\right\}$ ?

**Activity V.2.8** ( $\sim 5$  min) Does the complex number  $2i$  belong to  $\text{span}\{-3 + i, 6 - 2i\}$ ?

### Section V.3

**Activity V.3.1** ( $\sim 5$  min) How many vectors are required to span  $\mathbb{R}^2$ ? Sketch a drawing in the  $xy$  plane to support your answer.

- (a) 1
- (b) 2
- (c) 3
- (d) 4
- (e) Infinitely Many

**Activity V.3.2** ( $\sim 5$  min) How many vectors are required to span  $\mathbb{R}^3$ ?

- (a) 1
- (b) 2
- (c) 3
- (d) 4
- (e) Infinitely Many

**Fact V.3.3** At least  $n$  vectors are required to span  $\mathbb{R}^n$ .



**Activity V.3.4** ( $\sim 15$  min) Choose a vector  $\begin{bmatrix} ? \\ ? \\ ? \end{bmatrix}$  in  $\mathbb{R}^3$  that is not in  $\text{span} \left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix} \right\}$  by using CoCalc

to verify that  $\text{RREF} \left[ \begin{array}{cc|c} 1 & -2 & ? \\ -1 & 0 & ? \\ 0 & 1 & ? \end{array} \right] = \left[ \begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{array} \right]$ . (Why does this work?)

**Fact V.3.5** The set  $\{\mathbf{v}_1, \dots, \mathbf{v}_m\}$  fails to span all of  $\mathbb{R}^n$  exactly when  $\text{RREF}[\mathbf{v}_1 \dots \mathbf{v}_m]$  has a row of zeros:

$$\begin{bmatrix} 1 & -2 \\ -1 & 0 \\ 0 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & -2 & | & a \\ -1 & 0 & | & b \\ 0 & 1 & | & c \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & | & 0 \\ 0 & 1 & | & 0 \\ 0 & 0 & | & 1 \end{bmatrix} \text{ for some choice of vector } \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

**Activity V.3.6** (*~5 min*) Consider the set of vectors  $S = \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ -4 \\ 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 0 \\ 3 \\ 5 \\ 7 \end{bmatrix}, \begin{bmatrix} 3 \\ 13 \\ 7 \\ 16 \end{bmatrix} \right\}$ . Does  $\mathbb{R}^4 = \text{span } S$ ?

**Activity V.3.7** (*~10 min*) Consider the set of third-degree polynomials

$$S = \{2x^3 + 3x^2 - 1, 2x^3 + 3, 3x^3 + 13x^2 + 7x + 16, -x^3 + 10x^2 + 7x + 14, 4x^3 + 3x^2 + 2\}.$$

Does  $\mathcal{P}^3 = \text{span } S$ ? (Hint: first rewrite the question so it is about Euclidean vectors.)

**Activity V.3.8** (*~5 min*) Consider the set of matrices

$$S = \left\{ \begin{bmatrix} 1 & 3 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \right\}$$

Does  $M_{2,2} = \text{span } S$ ?

**Activity V.3.9** (*~5 min*) Let  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3 \in \mathbb{R}^7$  be three vectors, and suppose  $\mathbf{w}$  is another vector with  $\mathbf{w} \in \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ . What can you conclude about  $\text{span}\{\mathbf{w}, \mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ ?

- (a)  $\text{span}\{\mathbf{w}, \mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is larger than  $\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ .
- (b)  $\text{span}\{\mathbf{w}, \mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} = \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ .
- (c)  $\text{span}\{\mathbf{w}, \mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is smaller than  $\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ .

## Section V.4

**Definition V.4.1** A subset of a vector space is called a **subspace** if it is a vector space on its own.

For example, the span of these two vectors forms a planar subspace inside of the larger vector space  $\mathbb{R}^3$ .



**Fact V.4.2** Any subset  $S$  of a vector space  $V$  satisfies the eight vector space properties automatically, since it is a collection of known vectors.

However, to verify that it's a **subspace**, we need to check that addition and multiplication still make sense using only vectors from  $S$ . So we need to check two things:

- The set is **closed under addition**: for any  $\mathbf{x}, \mathbf{y} \in S$ , the sum  $\mathbf{x} + \mathbf{y}$  is also in  $S$ .
- The set is **closed under scalar multiplication**: for any  $\mathbf{x} \in S$  and scalar  $c \in \mathbb{R}$ , the product  $c\mathbf{x}$  is also in  $S$ .

**Activity V.4.3** ( $\sim 15$  min) Let  $S = \left\{ \begin{bmatrix} x \\ y \\ z \end{bmatrix} \mid x + 2y + z = 0 \right\}$ .

*Part 1:* Let  $\mathbf{v} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$  and  $\mathbf{w} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$  be vectors in  $S$ , so  $x + 2y + z = 0$  and  $a + 2b + c = 0$ . Show that

$\mathbf{v} + \mathbf{w} = \begin{bmatrix} x+a \\ y+b \\ z+c \end{bmatrix}$  also belongs to  $S$  by verifying that  $(x+a) + 2(y+b) + (z+c) = 0$ .

*Part 2:* Let  $\mathbf{v} = \begin{bmatrix} x \\ y \\ z \end{bmatrix} \in S$ , so  $x + 2y + z = 0$ . Show that  $c\mathbf{v}$  also belongs to  $S$  for any  $c \in \mathbb{R}$ .

*Part 3:* Is  $S$  a subspace of  $\mathbb{R}^3$ ?

**Activity V.4.4** ( $\sim 10$  min) Let  $S = \left\{ \begin{bmatrix} x \\ y \\ z \end{bmatrix} \mid x + 2y + z = 4 \right\}$ . Choose a vector  $\mathbf{v} = \begin{bmatrix} ? \\ ? \\ ? \end{bmatrix}$  in  $S$  and a real number  $c = ?$ , and show that  $c\mathbf{v}$  isn't in  $S$ . Is  $S$  a subspace of  $\mathbb{R}^3$ ?

**Remark V.4.5** Since  $0$  is a scalar and  $0\mathbf{v} = \mathbf{z}$  for any vector  $\mathbf{v}$ , a set that is closed under scalar multiplication must contain the zero vector  $\mathbf{z}$  for that vector space.

Put another way, an easy way to check that a subset isn't a subspace is to show it doesn't contain  $\mathbf{0}$ .

**Activity V.4.6** ( $\sim 10$  min) Consider these two subsets of  $\mathbb{R}^4$ :

$$S = \left\{ \begin{bmatrix} a \\ b \\ -b \\ -a \end{bmatrix} \mid a, b \text{ are real numbers} \right\} \quad T = \left\{ \begin{bmatrix} a \\ b \\ b-1 \\ a-1 \end{bmatrix} \mid a, b \text{ are real numbers} \right\}$$

*Part 1:* Which set is not a subspace of  $\mathbb{R}^4$ ?

*Part 2:* Is the set of polynomials

$$S = \{ax^3 + bx^2 + (b-1)x + (a-1) \mid a, b \text{ are real numbers}\}$$

a subspace of  $\mathcal{P}^3$ ?

**Activity V.4.7** ( $\sim 10$  min) Consider the subset  $A$  of  $\mathbb{R}^2$  where at least one coordinate of each vector is  $0$ .



This set contains  $\mathbf{0}$ , and it's not hard to show that for every  $\mathbf{v}$  in  $A$  and scalar  $c \in \mathbb{R}$ ,  $c\mathbf{v}$  is also in  $A$ . Is  $A$  a subspace of  $\mathbb{R}^2$ ? Why?

**(Instructor Note:)** Sketch the sum of two vectors on different axes to give a geometrical argument.

**Activity V.4.8** ( $\sim 5$  min) Let  $W$  be a subspace of a vector space  $V$ . How are  $\text{span } W$  and  $W$  related?

- (a)  $\text{span } W$  is bigger than  $W$
- (b)  $\text{span } W$  is the same as  $W$
- (c)  $\text{span } W$  is smaller than  $W$

**Fact V.4.9** If  $S$  is any subset of a vector space  $V$ , then since  $\text{span } S$  collects all possible linear combinations,  $\text{span } S$  is automatically a subspace of  $V$ .

In fact,  $\text{span } S$  is always the smallest subspace of  $V$  that contains all the vectors in  $S$ .

## Module S: Structure of vector spaces

### Standards for this Module

**What structure do vector spaces have?** At the end of this module, students will be able to...

- S1. Linear independence.** ... determine if a set of Euclidean vectors is linearly dependent or independent.
- S2. Basis verification.** ... determine if a set of Euclidean vectors is a basis of  $\mathbb{R}^n$ .
- S3. Basis computation.** ... compute a basis for the subspace spanned by a given set of Euclidean vectors.
- S4. Dimension.** ... compute the dimension of a subspace of  $\mathbb{R}^n$ .
- S5. Abstract vector spaces.** ... solve exercises related to standards V3-S4 when posed in terms of polynomials or matrices.
- S6. Basis of solution space.** ... find a basis for the solution set of a homogeneous system of equations.

### Readiness Assurance Outcomes

Before beginning this module, each student should be able to...

- Add Euclidean vectors and multiply Euclidean vectors by scalars.
- Perform basic manipulations of augmented matrices and linear systems **E1,E2,E3**.
- Apply linear combinations and spanning sets **V3,V4**.

### Readiness Assurance Resources

The following resources will help you prepare for this module.

- Adding and subtracting Euclidean vectors (Khan Academy): <http://bit.ly/2y8A0wa>
- Linear combinations of Euclidean vectors (Khan Academy): <http://bit.ly/2nK3wne>
- Adding and subtracting complex numbers (Khan Academy): <http://bit.ly/1PE3ZMQ>
- Adding and subtracting polynomials (Khan Academy): <http://bit.ly/2d5SLGZ>



## Section S.1

**Activity S.1.1** (*~10 min*) Consider the two sets

$$S = \left\{ \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 4 \end{bmatrix} \right\} \qquad T = \left\{ \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 4 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ -11 \end{bmatrix} \right\}$$

Which of the following is true?

- (A)  $\text{span } S$  is bigger than  $\text{span } T$ .
- (B)  $\text{span } S$  and  $\text{span } T$  are the same size.
- (C)  $\text{span } S$  is smaller than  $\text{span } T$ .

**Definition S.1.2** We say that a set of vectors is **linearly dependent** if one vector in the set belongs to the span of the others. Otherwise, we say the set is **linearly independent**.



You can think of linearly dependent sets as containing a redundant vector, in the sense that you can drop a vector out without reducing the span of the set. In the above image, all three vectors lay on the same planar subspace, but only two vectors are needed to span the plane, so the set is linearly dependent.

**Activity S.1.3** (*~10 min*) Let  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  be vectors in  $\mathbb{R}^n$ . Suppose  $3\mathbf{u} - 5\mathbf{v} = \mathbf{w}$ , so the set  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is linearly dependent. Which of the following is true of the vector equation  $x\mathbf{u} + y\mathbf{v} + z\mathbf{w} = \mathbf{0}$ ?

- (A) It is consistent with one solution
- (B) It is consistent with infinitely many solutions
- (C) It is inconsistent.

**Fact S.1.4** For any vector space, the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is linearly dependent if and only if  $x_1\mathbf{v}_1 + \dots + x_n\mathbf{v}_n = \mathbf{z}$  is consistent with infinitely many solutions.

**Activity S.1.5** (*~10 min*) Find

$$\text{RREF} \left[ \begin{array}{ccccc|c} 2 & 2 & 3 & -1 & 4 & 0 \\ 3 & 0 & 13 & 10 & 3 & 0 \\ 0 & 0 & 7 & 7 & 0 & 0 \\ -1 & 3 & 16 & 14 & 2 & 0 \end{array} \right]$$

and mark the part of the matrix that demonstrates that

$$S = \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 3 \\ 13 \\ 7 \\ 16 \end{bmatrix}, \begin{bmatrix} -1 \\ 10 \\ 7 \\ 14 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \\ 0 \\ 2 \end{bmatrix} \right\}$$

is linearly dependent (the part that shows its linear system has infinitely many solutions).

**Fact S.1.6** A set of Euclidean vectors  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is linearly dependent if and only if RREF  $[\mathbf{v}_1 \ \dots \ \mathbf{v}_n]$  has a column without a pivot position.

**Activity S.1.7** (*~5 min*) Is the set of Euclidean vectors  $\left\{ \begin{bmatrix} -4 \\ 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 10 \\ 10 \\ 2 \\ 6 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 7 \\ 2 \\ 1 \end{bmatrix} \right\}$  linearly dependent or linearly independent?

**Activity S.1.8** (*~10 min*) Is the set of polynomials  $\{x^3 + 1, x^2 + 2x, x^2 + 7x + 4\}$  linearly dependent or linearly independent?

**Activity S.1.9** (*~5 min*) What is the largest number of vectors in  $\mathbb{R}^4$  that can form a linearly independent set?

- (a) 3
- (b) 4
- (c) 5
- (d) You can have infinitely many vectors and still be linearly independent.

**Activity S.1.10** (*~5 min*) What is the largest number of vectors in

$$\mathcal{P}^4 = \{ax^4 + bx^3 + cx^2 + dx + e \mid a, b, c, d, e \in \mathbb{R}\}$$

that can form a linearly independent set?

- (a) 3
- (b) 4
- (c) 5
- (d) You can have infinitely many vectors and still be linearly independent.

**Activity S.1.11** (*~5 min*) What is the largest number of vectors in

$$\mathcal{P} = \{f(x) \mid f(x) \text{ is any polynomial}\}$$

that can form a linearly independent set?

- (a) 3
- (b) 4
- (c) 5
- (d) You can have infinitely many vectors and still be linearly independent.

## Section S.2

**Definition S.2.1** A **basis** is a linearly independent set that spans a vector space.

The **standard basis** of  $\mathbb{R}^n$  is the set  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  where

$$\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix} \quad \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix} \quad \dots \quad \mathbf{e}_n = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}$$

For  $\mathbb{R}^3$ , these are the vectors  $\mathbf{e}_1 = \hat{i} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ ,  $\mathbf{e}_2 = \hat{j} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ , and  $\mathbf{e}_3 = \hat{k} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ .

**Observation S.2.2** A basis may be thought of as a collection of building blocks for a vector space, since every vector in the space can be expressed as a unique linear combination of basis vectors.

For example, in many calculus courses, vectors in  $\mathbb{R}^3$  are often expressed in their component form

$$(3, -2, 4) = \begin{bmatrix} 3 \\ -2 \\ 4 \end{bmatrix}$$

or in their standard basic vector form

$$3\mathbf{e}_1 - 2\mathbf{e}_2 + 4\mathbf{e}_3 = 3\hat{i} - 2\hat{j} + 4\hat{k}.$$

Since every vector in  $\mathbb{R}^3$  can be uniquely described as a linear combination of the vectors in  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$ , this set is indeed a basis.

**Activity S.2.3** (*~15 min*) Label each of the sets  $A, B, C, D, E$  as

- SPANS  $\mathbb{R}^4$  or DOES NOT SPAN  $\mathbb{R}^4$
- LINEARLY INDEPENDENT or LINEARLY DEPENDENT
- BASIS FOR  $\mathbb{R}^4$  or NOT A BASIS FOR  $\mathbb{R}^4$

by finding RREF for their corresponding matrices.

$$\begin{aligned}
 A &= \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\} & B &= \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \\ 3 \end{bmatrix} \right\} \\
 C &= \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 3 \\ 13 \\ 7 \\ 16 \end{bmatrix}, \begin{bmatrix} -1 \\ 10 \\ 7 \\ 14 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \\ 0 \\ 2 \end{bmatrix} \right\} & D &= \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \\ 5 \end{bmatrix} \right\} \\
 E &= \left\{ \begin{bmatrix} 5 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} -2 \\ 1 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 5 \\ 1 \\ 3 \end{bmatrix} \right\}
 \end{aligned}$$

**Activity S.2.4** (*~10 min*) If  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4\}$  is a basis for  $\mathbb{R}^4$ , that means  $\text{RREF}[\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3 \ \mathbf{v}_4]$  doesn't have a non-pivot column, and doesn't have a row of zeros. What is  $\text{RREF}[\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3 \ \mathbf{v}_4]$ ?

$$\text{RREF}[\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3 \ \mathbf{v}_4] = \begin{bmatrix} ? & ? & ? & ? \\ ? & ? & ? & ? \\ ? & ? & ? & ? \\ ? & ? & ? & ? \end{bmatrix}$$

**Fact S.2.5** The set  $\{\mathbf{v}_1, \dots, \mathbf{v}_m\}$  is a basis for  $\mathbb{R}^n$  if and only if  $m = n$  and  $\text{RREF}[\mathbf{v}_1 \ \dots \ \mathbf{v}_n] = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$ .

That is, a basis for  $\mathbb{R}^n$  must have exactly  $n$  vectors and its square matrix must row-reduce to the so-called **identity matrix** containing all zeros except for a downward diagonal of ones. (We will learn where the identity matrix gets its name in a later module.)

**Observation S.2.6** Recall that a **subspace** of a vector space is a subset that is itself a vector space.

One easy way to construct a subspace is to take the span of set, but a linearly dependent set contains “redundant” vectors. For example, only two of the three vectors in the following image are needed to span the planar subspace.



**Activity S.2.7** ( $\sim 10$  min) Consider the subspace  $W = \text{span} \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ -3 \\ 2 \\ -3 \end{bmatrix}, \begin{bmatrix} 1 \\ 5 \\ -1 \\ 0 \end{bmatrix} \right\}$  of  $\mathbb{R}^4$ .

*Part 1:* Mark the part of RREF  $\begin{bmatrix} 2 & 2 & 2 & 1 \\ 3 & 0 & -3 & 5 \\ 0 & 1 & 2 & -1 \\ 1 & -1 & -3 & 0 \end{bmatrix}$  that shows that  $W$ 's spanning set is linearly dependent.

*Part 2:* Find a basis for  $W$  by removing a vector from its spanning set to make it linearly independent.

**Fact S.2.8** Let  $S = \{\mathbf{v}_1, \dots, \mathbf{v}_m\}$ . The easiest basis describing  $\text{span } S$  is the set of vectors in  $S$  given by the pivot columns of  $\text{RREF}[\mathbf{v}_1 \dots \mathbf{v}_m]$ .

Put another way, to compute a basis for the subspace  $\text{span } S$ , simply remove the vectors corresponding to the non-pivot columns of  $\text{RREF}[\mathbf{v}_1 \dots \mathbf{v}_m]$ .

**Activity S.2.9** ( $\sim 10$  min) Let  $W$  be the subspace of  $\mathbb{R}^4$  given by

$$W = \text{span} \left\{ \begin{bmatrix} 1 \\ 3 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ -1 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 4 \\ 5 \\ 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 2 \\ 2 \\ 1 \end{bmatrix} \right\}$$

Find a basis for  $W$ .

**Activity S.2.10** ( $\sim 10$  min) Let  $W$  be the subspace of  $\mathcal{P}^3$  given by

$$W = \text{span} \{x^3 + 3x^2 + x - 1, 2x^3 - x^2 + x + 2, 4x^3 + 5x^2 + 3x, 3x^3 + 2x^2 + 2x + 1\}$$

Find a basis for  $W$ .

### Section S.3

**Observation S.3.1** In the previous section, we learned that computing a basis for the subspace  $\text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_m\}$ , is as simple as removing the vectors corresponding to the non-pivot columns of  $\text{RREF}[\mathbf{v}_1 \dots \mathbf{v}_m]$ .

For example, since

$$\text{RREF} \begin{bmatrix} 1 & 2 & 3 \\ 0 & -2 & -2 \\ -3 & 1 & -2 \end{bmatrix} = \begin{bmatrix} \textcircled{1} & 0 & 1 \\ 0 & \textcircled{1} & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

the subspace  $W = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix}, \begin{bmatrix} 2 \\ -2 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ -2 \\ -2 \end{bmatrix} \right\}$  has  $\left\{ \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix}, \begin{bmatrix} 2 \\ -2 \\ 1 \end{bmatrix} \right\}$  as a basis.

**Activity S.3.2** ( $\sim 10$  min) Let

$$S = \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ -3 \\ 2 \\ -3 \end{bmatrix}, \begin{bmatrix} 1 \\ 5 \\ -1 \\ 0 \end{bmatrix} \right\} \quad \text{and} \quad T = \left\{ \begin{bmatrix} 2 \\ 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ -3 \\ 2 \\ -3 \end{bmatrix}, \begin{bmatrix} 1 \\ 5 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \\ 1 \end{bmatrix} \right\}$$

*Part 1:* Find a basis for  $\text{span } S$ .

*Part 2:* Find a basis for  $\text{span } T$ .

**Observation S.3.3** Even though we found different bases for them,  $\text{span } S$  and  $\text{span } T$  are exactly the same subspace of  $\mathbb{R}^4$ , since

$$S = \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ -3 \\ 2 \\ -3 \end{bmatrix}, \begin{bmatrix} 1 \\ 5 \\ -1 \\ 0 \end{bmatrix} \right\} = \left\{ \begin{bmatrix} 2 \\ 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ -3 \\ 2 \\ -3 \end{bmatrix}, \begin{bmatrix} 1 \\ 5 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \\ 1 \end{bmatrix} \right\} = T$$

**Fact S.3.4** Any non-trivial vector space has infinitely-many different bases, but all the bases for a given vector space are exactly the same size.

For example,

$$\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\} \quad \text{and} \quad \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\} \quad \text{and} \quad \left\{ \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix}, \begin{bmatrix} 2 \\ -2 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ -2 \\ 5 \end{bmatrix} \right\}$$

are all valid bases for  $\mathbb{R}^3$ , and they all contain three vectors.

**Definition S.3.5** The **dimension** of a vector space is equal to the size of any basis for the vector space.

As you'd expect,  $\mathbb{R}^n$  has dimension  $n$ . For example,  $\mathbb{R}^3$  has dimension 3 because any basis for  $\mathbb{R}^3$  such as

$$\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\} \text{ and } \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\} \text{ and } \left\{ \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix}, \begin{bmatrix} 2 \\ -2 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ -2 \\ 5 \end{bmatrix} \right\}$$

contains exactly three vectors.

**Activity S.3.6** (*~10 min*) Find the dimension of each subspace of  $\mathbb{R}^4$  by finding RREF for each corresponding matrix.

$$\begin{aligned} \text{span} \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \\ 3 \end{bmatrix} \right\} & \quad \text{span} \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 3 \\ 13 \\ 7 \\ 16 \end{bmatrix}, \begin{bmatrix} -1 \\ 10 \\ 7 \\ 14 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \\ 0 \\ 2 \end{bmatrix} \right\} \\ \text{span} \left\{ \begin{bmatrix} 2 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 3 \\ 6 \\ 1 \\ 5 \end{bmatrix} \right\} & \quad \text{span} \left\{ \begin{bmatrix} 5 \\ 3 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} -2 \\ 1 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 5 \\ 1 \\ 3 \end{bmatrix} \right\} \end{aligned}$$

**Fact S.3.7** Every vector space with finite dimension, that is, every vector space  $V$  with a basis of the form  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is said to be **isomorphic** to a Euclidean space  $\mathbb{R}^n$ , since there exists a natural correspondence between vectors in  $V$  and vectors in  $\mathbb{R}^n$ :

$$c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n \leftrightarrow \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

**Observation S.3.8** We've already been taking advantage of the previous fact by converting polynomials and matrices into Euclidean vectors. Since  $\mathcal{P}^3$  and  $M_{2,2}$  are both four-dimensional:

$$4x^3 + 0x^2 - 1x + 5 \leftrightarrow \begin{bmatrix} 4 \\ 0 \\ -1 \\ 5 \end{bmatrix} \leftrightarrow \begin{bmatrix} 4 & 0 \\ -1 & 5 \end{bmatrix}$$

**Observation S.3.9** The space of polynomials  $\mathcal{P}$  (of *any* degree) has the basis  $\{1, x, x^2, x^3, \dots\}$ , so it is a natural example of an infinite-dimensional vector space.

Since  $\mathcal{P}$  and other infinite-dimensional spaces cannot be treated as an isomorphic finite-dimensional Euclidean space  $\mathbb{R}^n$ , vectors in such spaces cannot be studied by converting them into Euclidean vectors. Fortunately, most of the examples we will be interested in for this course will be finite-dimensional.



**Definition S.3.10** A **homogeneous** system of linear equations is one of the form:

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n &= 0 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n &= 0 \\ \vdots & \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n &= 0 \end{aligned}$$

This system is equivalent to the vector equation:

$$x_1\mathbf{v}_1 + \dots + x_n\mathbf{v}_n = \mathbf{0}$$

and the augmented matrix:

$$\left[ \begin{array}{cccc|c} a_{11} & a_{12} & \cdots & a_{1n} & 0 \\ a_{21} & a_{22} & \cdots & a_{2n} & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} & 0 \end{array} \right]$$

**Activity S.3.11** ( $\sim 5$  min) Note that if  $\begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$  and  $\begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$  are solutions to  $x_1\mathbf{v}_1 + \dots + x_n\mathbf{v}_n = \mathbf{0}$  so is

$$\begin{bmatrix} a_1 + b_1 \\ \vdots \\ a_n + b_n \end{bmatrix}, \text{ since}$$

$$a_1\mathbf{v}_1 + \dots + a_n\mathbf{v}_n = \mathbf{0} \text{ and } b_1\mathbf{v}_1 + \dots + b_n\mathbf{v}_n = \mathbf{0}$$

implies

$$(a_1 + b_1)\mathbf{v}_1 + \dots + (a_n + b_n)\mathbf{v}_n = \mathbf{0}.$$

Similarly, if  $c \in \mathbb{R}$ ,  $\begin{bmatrix} ca_1 \\ \vdots \\ ca_n \end{bmatrix}$  is a solution. Thus the solution set of a homogeneous system is...

- a) A basis for  $\mathbb{R}^n$ .                      b) A subspace of  $\mathbb{R}^n$ .                      c) The empty set.

**Activity S.3.12** (*~10 min*) Consider the homogeneous system of equations

$$\begin{aligned}x_1 + 2x_2 + x_4 &= 0 \\2x_1 + 4x_2 - x_3 - 2x_4 &= 0 \\3x_1 + 6x_2 - x_3 - x_4 &= 0\end{aligned}$$

*Part 1:* Find its solution set (a subspace of  $\mathbb{R}^4$ ).

*Part 2:* Rewrite this solution space in the form

$$\left\{ a \begin{bmatrix} ? \\ ? \\ ? \\ ? \end{bmatrix} + b \begin{bmatrix} ? \\ ? \\ ? \\ ? \end{bmatrix} \mid a, b \in \mathbb{R} \right\}.$$

**Fact S.3.13** The coefficients of the free variables in the solution set of a linear system always yield linearly independent vectors.

Thus if

$$\left\{ a \begin{bmatrix} 4 \\ 1 \\ 0 \\ 0 \end{bmatrix} + b \begin{bmatrix} -3 \\ 0 \\ -2 \\ 1 \end{bmatrix} \mid a, b \in \mathbb{R} \right\}$$

is the solution space for a homogeneous system, then

$$\left\{ \begin{bmatrix} 4 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ -2 \\ 1 \end{bmatrix} \right\}$$

is a basis for the solution space.

**Activity S.3.14** (*~10 min*) Consider the homogeneous system of equations

$$\begin{aligned}x_1 - 3x_2 + 2x_3 &= 0 \\2x_1 - 6x_2 + 4x_3 + 3x_4 &= 0 \\-2x_1 + 6x_2 - 4x_3 - 4x_4 &= 0\end{aligned}$$

Find a basis for its solution space.

**Activity S.3.15** ( $\sim 5$  min) Suppose  $W$  is a subspace of  $\mathcal{P}^8$ , and you know that it contains a **linearly independent** set of 3 vectors. What can you conclude about  $W$ ?

- (a) The dimension of  $W$  is at most 3.
- (b) The dimension of  $W$  is exactly 3.
- (c) The dimension of  $W$  is at least 3.

**Activity S.3.16** ( $\sim 5$  min) Suppose  $W$  is a subspace of  $\mathcal{P}^8$ , and you know that it contains a **spanning set** of 3 vectors. What can you conclude about  $W$ ?

- (a) The dimension of  $W$  is at most 3.
- (b) The dimension of  $W$  is exactly 3.
- (c) The dimension of  $W$  is at least 3.

## Module A: Algebraic properties of linear maps

### Standards for this Module

**How can we understand linear maps algebraically?** At the end of this module, students will be able to...

- A1. Linear map verification.** ... determine if a map between vector spaces of polynomials is linear or not.
- A2. Linear maps and matrices.** ... translate back and forth between a linear transformation of Euclidean spaces and its standard matrix, and perform related computations.
- A3. Injectivity and surjectivity.** ... determine if a given linear map is injective and/or surjective.
- A4. Kernel and Image.** ... compute a basis for the kernel and a basis for the image of a linear map.

### Readiness Assurance Outcomes

Before beginning this module, each student should be able to...

- State the definition of a spanning set, and determine if a set of Euclidean vectors spans  $\mathbb{R}^n$  **V4**.
- State the definition of linear independence, and determine if a set of Euclidean vectors is linearly dependent or independent **S1**.
- State the definition of a basis, and determine if a set of Euclidean vectors is a basis **S2,S3**.
- Find a basis of the solution space to a homogeneous system of linear equations **S6**.

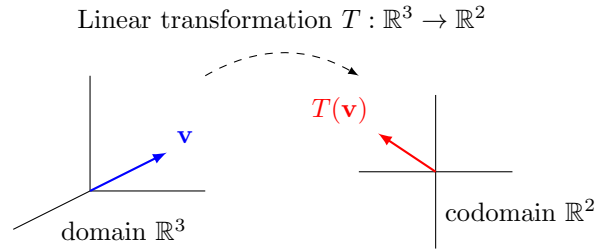
## Section A.1

**Definition A.1.1** A **linear transformation** (also known as a **linear map**) is a map between vector spaces that preserves the vector space operations. More precisely, if  $V$  and  $W$  are vector spaces, a map  $T : V \rightarrow W$  is called a linear transformation if

1.  $T(\mathbf{v} + \mathbf{w}) = T(\mathbf{v}) + T(\mathbf{w})$  for any  $\mathbf{v}, \mathbf{w} \in V$ .
2.  $T(c\mathbf{v}) = cT(\mathbf{v})$  for any  $c \in \mathbb{R}, \mathbf{v} \in V$ .

In other words, a map is linear when vector space operations can be applied before or after the transformation without affecting the result.

**Definition A.1.2** Given a linear transformation  $T : V \rightarrow W$ ,  $V$  is called the **domain** of  $T$  and  $W$  is called the **co-domain** of  $T$ .



**Example A.1.3** Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by

$$T \left( \begin{bmatrix} x \\ y \\ z \end{bmatrix} \right) = \begin{bmatrix} x - z \\ 3y \end{bmatrix}$$

To show that  $T$  is linear, we must verify...

$$\begin{aligned} T \left( \begin{bmatrix} x \\ y \\ z \end{bmatrix} + \begin{bmatrix} u \\ v \\ w \end{bmatrix} \right) &= T \left( \begin{bmatrix} x+u \\ y+v \\ z+w \end{bmatrix} \right) = \begin{bmatrix} (x+u) - (z+w) \\ 3(y+v) \end{bmatrix} \\ T \left( \begin{bmatrix} x \\ y \\ z \end{bmatrix} \right) + T \left( \begin{bmatrix} u \\ v \\ w \end{bmatrix} \right) &= \begin{bmatrix} x-z \\ 3y \end{bmatrix} + \begin{bmatrix} u-w \\ 3v \end{bmatrix} = \begin{bmatrix} (x+u) - (z+w) \\ 3(y+v) \end{bmatrix} \end{aligned}$$

And also...

$$T \left( c \begin{bmatrix} x \\ y \\ z \end{bmatrix} \right) = T \left( \begin{bmatrix} cx \\ cy \\ cz \end{bmatrix} \right) = \begin{bmatrix} cx - cz \\ 3cy \end{bmatrix} \quad \text{and} \quad cT \left( \begin{bmatrix} x \\ y \\ z \end{bmatrix} \right) = c \begin{bmatrix} x - z \\ 3y \end{bmatrix} = \begin{bmatrix} cx - cz \\ 3cy \end{bmatrix}$$

Therefore  $T$  is a linear transformation.

**Example A.1.4** Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^4$  be given by

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x + y \\ x^2 \\ y + 3 \\ y - 2^x \end{bmatrix}$$

To show that  $T$  is not linear, we only need to find one counterexample.

$$T\left(\begin{bmatrix} 0 \\ 1 \end{bmatrix} + \begin{bmatrix} 2 \\ 3 \end{bmatrix}\right) = T\left(\begin{bmatrix} 2 \\ 4 \end{bmatrix}\right) = \begin{bmatrix} 6 \\ 4 \\ 7 \\ 0 \end{bmatrix}$$

$$T\left(\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) + T\left(\begin{bmatrix} 2 \\ 3 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ 0 \\ 4 \\ -1 \end{bmatrix} + \begin{bmatrix} 5 \\ 4 \\ 6 \\ -5 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \\ 10 \\ -6 \end{bmatrix}$$

Since the resulting vectors are different,  $T$  is not a linear transformation.

**Fact A.1.5** A map between Euclidean spaces  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is linear exactly when every component of the output is a linear combination of the variables of  $\mathbb{R}^n$ .

For example, the following map is definitely linear because  $x - z$  and  $3y$  are linear combinations of  $x, y, z$ :

$$T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = \begin{bmatrix} x - z \\ 3y \end{bmatrix} = \begin{bmatrix} 1x + 0y - 1z \\ 0x + 3y + 0z \end{bmatrix}$$

But this map is not linear because  $x^2$ ,  $y + 3$ , and  $y - 2^x$  are not linear combinations (even though  $x + y$  is):

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x + y \\ x^2 \\ y + 3 \\ y - 2^x \end{bmatrix}$$

**Activity A.1.6** ( $\sim 5$  min) Recall the following rules from calculus, where  $D : \mathcal{P} \rightarrow \mathcal{P}$  is the derivative map defined by  $D(f(x)) = f'(x)$  for each polynomial  $f$ .

$$D(f + g) = f'(x) + g'(x)$$

$$D(cf(x)) = cf'(x)$$

What can we conclude from these rules?

- a)  $\mathcal{P}$  is not a vector space
- b)  $D$  is a linear map
- c)  $D$  is not a linear map

**Activity A.1.7** ( $\sim 10$  min) Let the polynomial maps  $S : \mathcal{P}^4 \rightarrow \mathcal{P}^3$  and  $T : \mathcal{P}^4 \rightarrow \mathcal{P}^3$  be defined by

$$S(f(x)) = 2f'(x) - f''(x) \quad T(f(x)) = f'(x) + x^3$$

Compute  $S(x^4 + x)$ ,  $S(x^4) + S(x)$ ,  $T(x^4 + x)$ , and  $T(x^4) + T(x)$ . Which of these maps is definitely not linear?

**Fact A.1.8** If  $L : V \rightarrow W$  is linear, then  $L(\mathbf{z}) = L(0\mathbf{v}) = 0L(\mathbf{v}) = \mathbf{z}$  where  $\mathbf{z}$  is the additive identity of the vector spaces  $V, W$ .

Put another way, an easy way to prove that a map like  $T(f(x)) = f'(x) + x^3$  can't be linear is because

$$T(0) = \frac{d}{dx}[0] + x^3 = 0 + x^3 = x^3 \neq 0.$$

**Activity A.1.9** ( $\sim 15$  min) Continue to consider  $S : \mathcal{P}^4 \rightarrow \mathcal{P}^3$  defined by

$$S(f(x)) = 2f'(x) - f''(x)$$

*Part 1:* Verify that

$$S(f(x) + g(x)) = 2f'(x) + 2g'(x) - f''(x) - g''(x)$$

is equal to  $S(f(x)) + S(g(x))$  for all polynomials  $f, g$ .

*Part 2:* Verify that  $S(cf(x))$  is equal to  $cS(f(x))$  for all real numbers  $c$  and polynomials  $f$ . Is  $S$  linear?

**Activity A.1.10** ( $\sim 20$  min) Let the polynomial maps  $S : \mathcal{P} \rightarrow \mathcal{P}$  and  $T : \mathcal{P} \rightarrow \mathcal{P}$  be defined by

$$S(f(x)) = (f(x))^2 \quad T(f(x)) = 3xf(x^2)$$

*Part 1:* Show that  $S(x + 1) \neq S(x) + S(1)$  to verify that  $S$  is not linear.

*Part 2:* Prove that  $T$  is linear by verifying that  $T(f(x) + g(x)) = T(f(x)) + T(g(x))$  and  $T(cf(x)) = cT(f(x))$ .

**Observation A.1.11** Note that  $S$  in the previous activity is not linear, even though  $S(0) = (0)^2 = 0$ . So showing  $S(0) = 0$  isn't enough to prove a map is linear.

This is a similar situation to proving a subset is a subspace: if the subset doesn't contain  $\mathbf{z}$ , then the subset isn't a subspace. But if the subset contains  $\mathbf{z}$ , you cannot conclude anything.

## Section A.2

**Remark A.2.1** Recall that a linear map  $T : V \rightarrow W$  satisfies

1.  $T(\mathbf{v} + \mathbf{w}) = T(\mathbf{v}) + T(\mathbf{w})$  for any  $\mathbf{v}, \mathbf{w} \in V$ .
2.  $T(c\mathbf{v}) = cT(\mathbf{v})$  for any  $c \in \mathbb{R}, \mathbf{v} \in V$ .

In other words, a map is linear when vector space operations can be applied before or after the transformation without affecting the result.

**Activity A.2.2** ( $\sim 5$  min) Suppose  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  is a linear map, and you know  $T \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$  and

$$T \left( \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} -3 \\ 2 \end{bmatrix}. \text{ Compute } T \left( \begin{bmatrix} 3 \\ 0 \\ 0 \end{bmatrix} \right).$$

(a)  $\begin{bmatrix} 6 \\ 3 \end{bmatrix}$

(c)  $\begin{bmatrix} -4 \\ -2 \end{bmatrix}$

(b)  $\begin{bmatrix} -9 \\ 6 \end{bmatrix}$

(d)  $\begin{bmatrix} 6 \\ -4 \end{bmatrix}$

**Activity A.2.3** ( $\sim 3$  min) Suppose  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  is a linear map, and you know  $T \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$  and

$$T \left( \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} -3 \\ 2 \end{bmatrix}. \text{ Compute } T \left( \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right).$$

(a)  $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$

(c)  $\begin{bmatrix} -1 \\ 3 \end{bmatrix}$

(b)  $\begin{bmatrix} 3 \\ -1 \end{bmatrix}$

(d)  $\begin{bmatrix} 5 \\ -8 \end{bmatrix}$



**Activity A.2.4** ( $\sim 2$  min) Suppose  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  is a linear map, and you know  $T \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$  and

$$T \left( \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} -3 \\ 2 \end{bmatrix}. \text{ Compute } T \left( \begin{bmatrix} -2 \\ 0 \\ -3 \end{bmatrix} \right).$$

(a)  $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$

(c)  $\begin{bmatrix} -1 \\ 3 \end{bmatrix}$

(b)  $\begin{bmatrix} 3 \\ -1 \end{bmatrix}$

(d)  $\begin{bmatrix} 5 \\ -8 \end{bmatrix}$

**Activity A.2.5** ( $\sim 5$  min) Suppose  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  is a linear map, and you know  $T \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$  and

$$T \left( \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} -3 \\ 2 \end{bmatrix}. \text{ Do you have enough information to compute } T(\mathbf{v}) \text{ for any } \mathbf{v} \in \mathbb{R}^3?$$

(a) Yes.

(b) No, exactly one more piece of information is needed.

(c) No, an infinite amount of information would be necessary to compute the transformation of infinitely-many vectors.

**Fact A.2.6** Consider any basis  $\{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  for  $V$ . Since every vector  $\mathbf{v}$  can be written *uniquely* as a linear combination of basis vectors,  $x_1\mathbf{b}_1 + \dots + x_n\mathbf{b}_n$ , we may compute  $T(\mathbf{v})$  as follows:

$$T(\mathbf{v}) = T(x_1\mathbf{b}_1 + \dots + x_n\mathbf{b}_n) = x_1T(\mathbf{b}_1) + \dots + x_nT(\mathbf{b}_n).$$

Therefore any linear transformation  $T : V \rightarrow W$  can be defined by just describing the values of  $T(\mathbf{b}_i)$ . Put another way, the images of the basis vectors **determine** the transformation  $T$ .

**Definition A.2.7** Since linear transformation  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is determined by the standard basis  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ , it's convenient to store this information in the  $m \times n$  **standard matrix**  $[T(\mathbf{e}_1) \ \dots \ T(\mathbf{e}_n)]$ .

For example, let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be the linear map determined by the following values for  $T$  applied to the standard basis of  $\mathbb{R}^3$ .

$$T(\mathbf{e}_1) = T \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 3 \\ 2 \end{bmatrix} \quad T(\mathbf{e}_2) = T \left( \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} -1 \\ 4 \end{bmatrix} \quad T(\mathbf{e}_3) = T \left( \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} 5 \\ 0 \end{bmatrix}$$

Then the standard matrix corresponding to  $T$  is

$$[T(\mathbf{e}_1) \ T(\mathbf{e}_2) \ T(\mathbf{e}_3)] = \begin{bmatrix} 3 & -1 & 5 \\ 2 & 4 & 0 \end{bmatrix}.$$

**Activity A.2.8** ( $\sim 3$  min) Let  $T : \mathbb{R}^4 \rightarrow \mathbb{R}^3$  be the linear transformation given by

$$T(\mathbf{e}_1) = \begin{bmatrix} 0 \\ 3 \\ -2 \end{bmatrix} \quad T(\mathbf{e}_2) = \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} \quad T(\mathbf{e}_3) = \begin{bmatrix} 4 \\ -2 \\ 1 \end{bmatrix} \quad T(\mathbf{e}_4) = \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix}$$

Write the standard matrix  $[T(\mathbf{e}_1) \cdots T(\mathbf{e}_n)]$  for  $T$ .

**Activity A.2.9** ( $\sim 5$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be the linear transformation given by

$$T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = \begin{bmatrix} x + 3z \\ 2x - y - 4z \end{bmatrix}$$

Find the standard matrix for  $T$ .

**Fact A.2.10** Because every linear map  $T : \mathbb{R}^m \rightarrow \mathbb{R}^n$  has a linear combination of the variables in each component, and thus  $T(\mathbf{e}_i)$  yields exactly the coefficients of  $x_i$ , the standard matrix for  $T$  is simply an ordered list of the coefficients of the  $x_i$ :

$$T\left(\begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix}\right) = \begin{bmatrix} ax + by + cz + dw \\ ex + fy + gz + hw \end{bmatrix} \quad A = \begin{bmatrix} a & b & c & d \\ e & f & g & h \end{bmatrix}$$

**Activity A.2.11** ( $\sim 5$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be the linear transformation given by the standard matrix

$$\begin{bmatrix} 3 & -2 & -1 \\ 4 & 5 & 2 \\ 0 & -2 & 1 \end{bmatrix}.$$

Compute  $T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right)$ .

**Activity A.2.12** ( $\sim 5$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be the linear transformation given by the standard matrix

$$\begin{bmatrix} 3 & -2 & -1 \\ 4 & 5 & 2 \\ 0 & -2 & 1 \end{bmatrix}.$$

Compute  $T\left(\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}\right)$ .

**Fact A.2.13** To quickly compute  $T(\mathbf{v})$  from its standard matrix  $A$ , compute the **dot product** (defined in Calculus 3) of each matrix row with the vector. For example, if  $T$  has the standard matrix

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

then for  $\mathbf{v} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$  we will write

$$T(\mathbf{v}) = A\mathbf{v} = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & -2 \\ 2 & -1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 1x + 2y + 3z \\ 0x + 1y - 2z \\ 2x - 1y + 0z \end{bmatrix}$$

and for  $\mathbf{v} = \begin{bmatrix} 3 \\ 0 \\ -2 \end{bmatrix}$  we will write

$$T(\mathbf{v}) = A\mathbf{v} = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & -2 \\ 2 & -1 & 0 \end{bmatrix} \begin{bmatrix} 3 \\ 0 \\ -2 \end{bmatrix} = \begin{bmatrix} 1(3) + 2(0) + 3(-2) \\ 0(3) + 1(0) - 2(-2) \\ 2(3) - 1(0) + 0(-2) \end{bmatrix} = \begin{bmatrix} -3 \\ 4 \\ 6 \end{bmatrix}.$$

**Activity A.2.14** (*~15 min*) Compute the following linear transformations of vectors given their standard matrices.

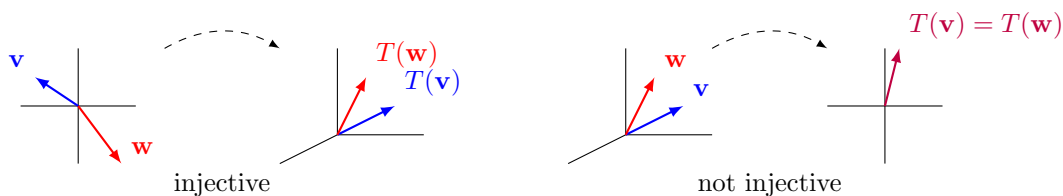
$$T\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right) \text{ for the standard matrix } A = \begin{bmatrix} 4 & 3 \\ 0 & -1 \\ 1 & 1 \\ 3 & 0 \end{bmatrix}$$

$$T\left(\begin{bmatrix} 1 \\ 1 \\ 0 \\ -3 \end{bmatrix}\right) \text{ for the standard matrix } A = \begin{bmatrix} 4 & 3 & 0 & -1 \\ 1 & 1 & 3 & 0 \end{bmatrix}$$

$$T\left(\begin{bmatrix} 0 \\ -2 \\ 0 \end{bmatrix}\right) \text{ for the standard matrix } A = \begin{bmatrix} 4 & 3 & 0 \\ 0 & -1 & 3 \\ 5 & 1 & 1 \\ 3 & 0 & 0 \end{bmatrix}$$

## Section A.3

**Definition A.3.1** Let  $T : V \rightarrow W$  be a linear transformation.  $T$  is called **injective** or **one-to-one** if  $T$  does not map two distinct values to the same place. More precisely,  $T$  is injective if  $T(\mathbf{v}) \neq T(\mathbf{w})$  whenever  $\mathbf{v} \neq \mathbf{w}$ .



**Activity A.3.2** ( $\sim 3$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by

$$T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{with standard matrix} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

Show that  $T$  is not injective by finding two different vectors  $\mathbf{v}, \mathbf{w} \in \mathbb{R}^3$  such that  $T(\mathbf{v}) = T(\mathbf{w})$ .

**Activity A.3.3** ( $\sim 2$  min) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be given by

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix} \quad \text{with standard matrix} \quad \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Is  $T$  injective? If not, find two different vectors  $\mathbf{v}, \mathbf{w} \in \mathbb{R}^3$  such that  $T(\mathbf{v}) = T(\mathbf{w})$ .

**Definition A.3.4** Let  $T : V \rightarrow W$  be a linear transformation.  $T$  is called **surjective** or **onto** if every element of  $W$  is mapped to by an element of  $V$ . More precisely, for every  $\mathbf{w} \in W$ , there is some  $\mathbf{v} \in V$  with  $T(\mathbf{v}) = \mathbf{w}$ .



**Activity A.3.5** ( $\sim 3$  min) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be given by

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix} \quad \text{with standard matrix} \quad \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Show that  $T$  is not surjective by finding a vector in  $\mathbb{R}^3$  that  $T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right)$  can never equal.

**Activity A.3.6** ( $\sim 2$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by

$$T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{with standard matrix} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

Is  $T$  surjective? If not, find a vector in  $\mathbb{R}^2$  that  $T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right)$  can never equal.

**Observation A.3.7** As we will see, it's no coincidence that the RREF of the injective map's standard matrix

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

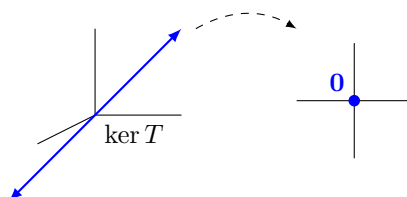
has all pivot columns. Similarly, the RREF of the surjective map's standard matrix

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

has all pivot (zero) rows.

**Definition A.3.8** Let  $T : V \rightarrow W$  be a linear transformation. The **kernel** of  $T$  is an important subspace of  $V$  defined by

$$\ker T = \{\mathbf{v} \in V \mid T(\mathbf{v}) = \mathbf{z}\}$$



**Activity A.3.9** (*~5 min*) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be given by

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix} \quad \text{with standard matrix} \quad \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Which of these subspaces of  $\mathbb{R}^2$  describes  $\ker T$ , the set of all vectors that transform into  $\mathbf{0}$ ?

a)  $\left\{ \begin{bmatrix} a \\ a \end{bmatrix} \mid a \in \mathbb{R} \right\}$

b)  $\left\{ \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right\}$

c)  $\mathbb{R}^2$

**Activity A.3.10** (*~5 min*) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by

$$T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{with standard matrix} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

Which of these subspaces of  $\mathbb{R}^3$  describes  $\ker T$ , the set of all vectors that transform into  $\mathbf{0}$ ?

a)  $\left\{ \begin{bmatrix} 0 \\ 0 \\ a \end{bmatrix} \mid a \in \mathbb{R} \right\}$

b)  $\left\{ \begin{bmatrix} a \\ a \\ 0 \end{bmatrix} \mid a \in \mathbb{R} \right\}$

c)  $\left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right\}$

d)  $\mathbb{R}^3$

**Activity A.3.11** (*~10 min*) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be the linear transformation given by the standard matrix

$$A = \begin{bmatrix} 3 & 4 & -1 \\ 1 & 2 & 1 \end{bmatrix}.$$

*Part 1:* Set  $T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = \begin{bmatrix} ? + ? + ? \\ ? + ? + ? \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$  to find a linear system of equations whose solution set is the kernel.

*Part 2:* Use  $\text{RREF}(A)$  to solve this homogeneous system of equations and find a basis for the kernel of  $T$ .

**Definition A.3.12** Let  $T : V \rightarrow W$  be a linear transformation. The **image** of  $T$  is an important subspace of  $W$  defined by

$$\text{Im } T = \{\mathbf{w} \in W \mid \text{there is some } \mathbf{v} \in V \text{ with } T(\mathbf{v}) = \mathbf{w}\}$$

In the examples below, the left example's image is all of  $\mathbb{R}^2$ , but the right example's image is a planar subspace of  $\mathbb{R}^3$ .



**Activity A.3.13** ( $\sim 5$  min) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be given by

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix} \quad \text{with standard matrix } \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Which of these subspaces of  $\mathbb{R}^3$  describes  $\text{Im } T$ , the set of all vectors that are the result of using  $T$  to transform  $\mathbb{R}^2$  vectors?

- a)  $\left\{ \begin{bmatrix} 0 \\ 0 \\ a \end{bmatrix} \mid a \in \mathbb{R} \right\}$
- b)  $\left\{ \begin{bmatrix} a \\ b \\ 0 \end{bmatrix} \mid a, b \in \mathbb{R} \right\}$
- c)  $\left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \right\}$
- d)  $\mathbb{R}^3$

**Activity A.3.14** ( $\sim 5$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by

$$T \left( \begin{bmatrix} x \\ y \\ z \end{bmatrix} \right) = \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{with standard matrix} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

Which of these subspaces of  $\mathbb{R}^2$  describes  $\text{Im } T$ , the set of all vectors that are the result of using  $T$  to transform  $\mathbb{R}^3$  vectors?

a)  $\left\{ \begin{bmatrix} a \\ a \end{bmatrix} \mid a \in \mathbb{R} \right\}$

b)  $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$

c)  $\mathbb{R}^2$

**Activity A.3.15** ( $\sim 5$  min) Let  $T : \mathbb{R}^4 \rightarrow \mathbb{R}^3$  be the linear transformation given by the standard matrix

$$A = \begin{bmatrix} 3 & 4 & 7 & 1 \\ -1 & 1 & 0 & 2 \\ 2 & 1 & 3 & -1 \end{bmatrix} = [T(\mathbf{e}_1) \quad T(\mathbf{e}_2) \quad T(\mathbf{e}_3) \quad T(\mathbf{e}_4)].$$

Since  $T(\mathbf{v}) = T(x_1\mathbf{e}_1 + x_2\mathbf{e}_2 + x_3\mathbf{e}_3 + x_4\mathbf{e}_4)$ , the set of vectors

$$\left\{ \begin{bmatrix} 3 \\ -1 \\ 2 \end{bmatrix}, \begin{bmatrix} 4 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 7 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix} \right\}$$

a) spans  $\text{Im } T$

b) is a linearly independent subset of  $\text{Im } T$

c) is a basis for  $\text{Im } T$

**Activity A.3.16** ( $\sim 5$  min) Let  $T : \mathbb{R}^4 \rightarrow \mathbb{R}^3$  be the linear transformation given by the standard matrix

$$A = \begin{bmatrix} 3 & 4 & 7 & 1 \\ -1 & 1 & 0 & 2 \\ 2 & 1 & 3 & -1 \end{bmatrix}.$$

Remove vectors from the set

$$\left\{ \begin{bmatrix} 3 \\ -1 \\ 2 \end{bmatrix}, \begin{bmatrix} 4 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 7 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix} \right\}$$

spanning  $\text{Im } T$  to get a basis for  $\text{Im } T$ .



**Observation A.3.17** Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear transformation with standard matrix  $A$ .

- The kernel of  $T$  is the solution set of the homogeneous system given by the augmented matrix  $[A \mid \mathbf{0}]$ . Use the coefficients of its free variables to get a basis for the kernel.
- The image of  $T$  is the span of the columns of  $A$ . Remove the vectors creating non-pivot columns in RREF  $A$  to get a basis for the image.

## Section A.4

**Observation A.4.1** Let  $T : V \rightarrow W$ . We have previously defined the following terms.

- $T$  is called **injective** or **one-to-one** if  $T$  does not map two distinct vectors to the same place.
- $T$  is called **surjective** or **onto** if every element of  $W$  is mapped to by some element of  $V$ .
- The **kernel** of  $T$  is the set of all vectors in  $V$  that are mapped to  $\mathbf{z} \in W$ . It is a subspace of  $V$ .
- The **image** of  $T$  is the set of all vectors in  $W$  that are mapped to by something in  $V$ . It is a subspace of  $W$ .

**Activity A.4.2** ( $\sim 5$  min) Let  $T : V \rightarrow W$  be a linear transformation where  $\ker T$  contains multiple vectors. What can you conclude?

- $T$  is injective
- $T$  is not injective
- $T$  is surjective
- $T$  is not surjective

**Fact A.4.3** A linear transformation  $T$  is injective **if and only if**  $\ker T = \{\mathbf{0}\}$ . Put another way, an injective linear transformation may be recognized by its **trivial** kernel.



**Activity A.4.4** ( $\sim 5$  min) Let  $T : \mathbb{R}^5 \rightarrow \mathbb{R}^5$  be a linear transformation where  $\text{Im } T$  is spanned less than five vectors. What can you conclude?

- $T$  is injective
- $T$  is not injective
- $T$  is surjective
- $T$  is not surjective

**Fact A.4.5** A linear transformation  $T : V \rightarrow W$  is surjective **if and only if**  $\text{Im } T = W$ . Put another way, a surjective linear transformation may be recognized by its identical codomain and image.



**Activity A.4.6** ( $\sim 15$  min) Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear map with standard matrix  $A$ . Sort the following claims into two groups of *equivalent* statements: one group that means  $T$  is **injective**, and one group that means  $T$  is **surjective**.

- (a) The kernel of  $T$  is trivial:  $\ker T = \{\mathbf{0}\}$ .
- (b) The columns of  $A$  span  $\mathbb{R}^m$ .
- (c) The columns of  $A$  are linearly independent.
- (d) Every column of  $\text{RREF}(A)$  has a pivot.
- (e) Every row of  $\text{RREF}(A)$  has a pivot.
- (f) The image of  $T$  equals its codomain:  $\text{Im } T = \mathbb{R}^m$ .
- (g) The system of linear equations given by the augmented matrix  $[A \mid \mathbf{b}]$  has a solution for all  $\mathbf{b} \in \mathbb{R}^m$ .
- (h) The system of linear equations given by the augmented matrix  $[A \mid \mathbf{0}]$  has exactly one solution.

**(Instructor Note:)** This activity may be ran as a card sort.

**Observation A.4.7** The easiest way to show that the linear map with standard matrix  $A$  is injective is to show that  $\text{RREF}(A)$  has all pivot columns.

The easiest way to show that the linear map with standard matrix  $A$  is surjective is to show that  $\text{RREF}(A)$  has all pivot rows.

**Activity A.4.8** ( $\sim 3$  min) What can you immediately conclude about the linear map  $T : \mathbb{R}^5 \rightarrow \mathbb{R}^3$ ?

- a) Its standard matrix has more columns than rows, so  $T$  is not injective.
- b) Its standard matrix has more rows than columns, so  $T$  is not surjective.

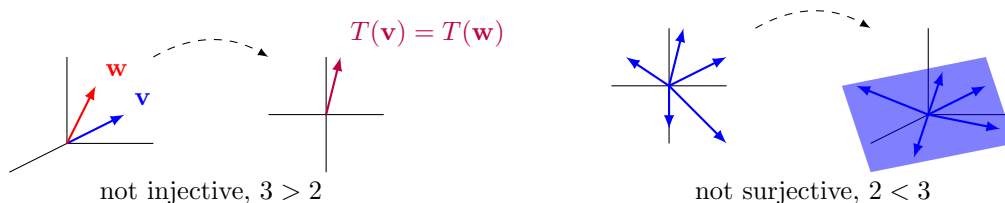
**Activity A.4.9** ( $\sim 2$  min) What can you immediately conclude about the linear map  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^7$ ?

- a) Its standard matrix has more columns than rows, so  $T$  is not injective.
- b) Its standard matrix has more rows than columns, so  $T$  is not surjective.

**Fact A.4.10** The following are true for any linear map  $T : V \rightarrow W$ :

- If  $\dim(V) > \dim(W)$ , then  $T$  is not injective.
- If  $\dim(V) < \dim(W)$ , then  $T$  is not surjective.

Basically, a linear transformation cannot reduce dimension without collapsing vectors into each other, and a linear transformation cannot increase the dimension of its image.



But these results do **not** reverse. For example,  $T : \mathbb{R}^5 \rightarrow \mathbb{R}^4$  might not be surjective, even though  $5 \not< 4$ .

**Definition A.4.11** If  $T : V \rightarrow W$  is both injective and surjective, it is called **bijjective**.

**Activity A.4.12** ( $\sim 5$  min) Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a bijective linear map with standard matrix  $A$ . Label each of the following as true or false.

- The columns of  $A$  form a basis for  $\mathbb{R}^m$
- $\text{RREF}(A)$  is the identity matrix.
- The system of linear equations given by the augmented matrix  $[A \mid \mathbf{b}]$  has exactly one solution for all  $\mathbf{b} \in \mathbb{R}^m$ .

**Observation A.4.13** The easiest way to show that the linear map with standard matrix  $A$  is bijective is to show that  $\text{RREF}(A)$  is the identity matrix.

**Activity A.4.14** ( $\sim 5$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be given by the standard matrix

$$A = \begin{bmatrix} 2 & 1 & -1 \\ 4 & 1 & 1 \\ 6 & 2 & 1 \end{bmatrix}.$$

Which of the following must be true?

- $T$  is neither injective nor surjective
- $T$  is injective but not surjective
- $T$  is surjective but not injective
- $T$  is bijective.

**Activity A.4.15** (*~5 min*) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be given by

$$T \left( \begin{bmatrix} x \\ y \\ z \end{bmatrix} \right) = \begin{bmatrix} 2x + y - z \\ 4x + y + z \\ 6x + 2y \end{bmatrix}.$$

Which of the following must be true?

- (a)  $T$  is neither injective nor surjective
- (b)  $T$  is injective but not surjective
- (c)  $T$  is surjective but not injective
- (d)  $T$  is bijective.

## Module M: Understanding Matrices Algebraically

### Standards for this Module

**What algebraic structure do matrices have?** At the end of this module, students will be able to...

**M1. Matrix Multiplication.** ... multiply matrices.

**M2. Invertible Matrices.** ... determine if a square matrix is invertible or not.

**M3. Matrix inverses.** ... compute the inverse matrix of an invertible matrix.

### Readiness Assurance Outcomes

Before beginning this module, each student should be able to...

- Compose functions of real numbers
- Solve systems of linear equations **E3**
- Find the matrix corresponding to a linear transformation **A1**
- Determine if a linear transformation is injective and/or surjective **A3**
- Interpret the ideas of injectivity and surjectivity in multiple ways

### Readiness Assurance Resources

The following resources will help you prepare for this module.

- Function composition (Khan Academy): <http://bit.ly/2wkz7f3>

**Section M.1**

**Activity M.1.1** (*~5 min*) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by the standard matrix  $B = \begin{bmatrix} 2 & 1 & -3 \\ 5 & -3 & 4 \end{bmatrix}$  and

$S : \mathbb{R}^2 \rightarrow \mathbb{R}^4$  be given by the standard matrix  $A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 3 & 5 \\ -1 & -2 \end{bmatrix}$ .

What is the domain of the composition map  $S \circ T$ ?

(a)  $\mathbb{R}$

(b)  $\mathbb{R}^2$

(c)  $\mathbb{R}^3$

(d)  $\mathbb{R}^4$

**Activity M.1.2** (*~2 min*) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by the standard matrix  $B = \begin{bmatrix} 2 & 1 & -3 \\ 5 & -3 & 4 \end{bmatrix}$  and

$S : \mathbb{R}^2 \rightarrow \mathbb{R}^4$  be given by the standard matrix  $A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 3 & 5 \\ -1 & -2 \end{bmatrix}$ .

What is the codomain of the composition map  $S \circ T$ ?

(a)  $\mathbb{R}$

(b)  $\mathbb{R}^2$

(c)  $\mathbb{R}^3$

(d)  $\mathbb{R}^4$

**Activity M.1.3** ( $\sim 2$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by the standard matrix  $B = \begin{bmatrix} 2 & 1 & -3 \\ 5 & -3 & 4 \end{bmatrix}$  and

$S : \mathbb{R}^2 \rightarrow \mathbb{R}^4$  be given by the standard matrix  $A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 3 & 5 \\ -1 & -2 \end{bmatrix}$ .

The standard matrix of  $S \circ T$  will lie in which matrix space?

- (a)  $4 \times 3$  matrices
- (b)  $4 \times 2$  matrices
- (c)  $3 \times 2$  matrices
- (d)  $2 \times 3$  matrices
- (e)  $2 \times 4$  matrices
- (f)  $3 \times 4$  matrices

**Activity M.1.4** ( $\sim 15$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be given by the standard matrix  $B = \begin{bmatrix} 2 & 1 & -3 \\ 5 & -3 & 4 \end{bmatrix}$  and

$S : \mathbb{R}^2 \rightarrow \mathbb{R}^4$  be given by the standard matrix  $A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 3 & 5 \\ -1 & -2 \end{bmatrix}$ .

*Part 1:* Compute  $(S \circ T)(\mathbf{e}_1)$

*Part 2:* Compute  $(S \circ T)(\mathbf{e}_2)$

*Part 3:* Compute  $(S \circ T)(\mathbf{e}_3)$ .

*Part 4:* Find the standard matrix of  $S \circ T$ .

**Activity M.1.5** ( $\sim 2$  min) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be given by the matrix  $B = \begin{bmatrix} 2 & 3 \\ 1 & -1 \\ 0 & -1 \end{bmatrix}$  and  $S : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be

given by the matrix  $A = \begin{bmatrix} -4 & -2 & 3 \\ 0 & 1 & 1 \end{bmatrix}$ .

What is the domain of the composition map  $S \circ T$ ?

- (a)  $\mathbb{R}$
- (b)  $\mathbb{R}^2$
- (c)  $\mathbb{R}^3$
- (d)  $\mathbb{R}^4$



**Activity M.1.6** ( $\sim 2$  min) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be given by the matrix  $B = \begin{bmatrix} 2 & 3 \\ 1 & -1 \\ 0 & -1 \end{bmatrix}$  and  $S : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be

given by the matrix  $A = \begin{bmatrix} -4 & -2 & 3 \\ 0 & 1 & 1 \end{bmatrix}$ .

What is the codomain of the composition map  $S \circ T$ ?

- (a)  $\mathbb{R}$
- (b)  $\mathbb{R}^2$
- (c)  $\mathbb{R}^3$
- (d)  $\mathbb{R}^4$

**Activity M.1.7** ( $\sim 2$  min) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be given by the matrix  $B = \begin{bmatrix} 2 & 3 \\ 1 & -1 \\ 0 & -1 \end{bmatrix}$  and  $S : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be

given by the matrix  $A = \begin{bmatrix} -4 & -2 & 3 \\ 0 & 1 & 1 \end{bmatrix}$ .

The standard matrix of  $S \circ T$  will lie in which matrix space?

- (a)  $2 \times 2$  matrices
- (b)  $2 \times 3$  matrices
- (c)  $3 \times 2$  matrices
- (d)  $3 \times 3$  matrices

**Activity M.1.8** ( $\sim 10$  min) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be given by the matrix  $B = \begin{bmatrix} 2 & 3 \\ 1 & -1 \\ 0 & -1 \end{bmatrix}$  and  $S : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be

given by the matrix  $A = \begin{bmatrix} -4 & -2 & 3 \\ 0 & 1 & 1 \end{bmatrix}$ .

Find the standard matrix of  $S \circ T$ .

**Activity M.1.9** ( $\sim 5$  min) Let  $T : \mathbb{R}^1 \rightarrow \mathbb{R}^4$  be given by the matrix  $B = \begin{bmatrix} 3 \\ -2 \\ 1 \\ -1 \end{bmatrix}$  and  $S : \mathbb{R}^4 \rightarrow \mathbb{R}^1$  be given

by the matrix  $A = [2 \ 3 \ 2 \ 5]$ .

Find the standard matrix of  $S \circ T$ .

**Definition M.1.10** We define the product of a  $m \times n$  matrix  $A$  and a  $n \times k$  matrix  $B$  to be the  $m \times k$  standard matrix (denoted  $AB$ ) of the composition map of the two corresponding linear functions.

**Fact M.1.11** If  $AB$  is defined,  $BA$  need not be defined, and if it is defined, it is in general different from  $AB$ .

**Activity M.1.12** (*~10 min*) Let  $A = \begin{bmatrix} 3 & 1 & -1 \\ 2 & 0 & 4 \end{bmatrix}$  and  $B = \begin{bmatrix} a & b \\ c & d \\ e & f \end{bmatrix}$ . Compute  $AB$ .

**Activity M.1.13** (*~5 min*) Let  $A = \begin{bmatrix} 3 & 1 & -1 \\ 2 & 0 & 4 \\ -1 & 3 & 5 \end{bmatrix}$  and  $X = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$ . Compute  $AX$

**Observation M.1.14** Consider the system of equations

$$\begin{aligned} 3x + y - z &= 5 \\ 2x + 4z &= -7 \\ -x + 3y + 5z &= 2 \end{aligned}$$

We can interpret this as a **matrix equation**  $AX = B$  where

$$A = \begin{bmatrix} 3 & 1 & -1 \\ 2 & 0 & 4 \\ -1 & 3 & 5 \end{bmatrix} \quad X = \begin{bmatrix} x \\ y \\ z \end{bmatrix} \quad B = \begin{bmatrix} 5 \\ -7 \\ 2 \end{bmatrix}$$

For this reason, we will swap out the use of Euclidean vectors  $\mathbf{x} \in \mathbb{R}^n$  and  $n \times 1$  matrices  $X$  whenever it is convenient.

## Section M.2

**Activity M.2.1** ( $\sim 5$  min) Let  $A = \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix}$ . Find a  $3 \times 3$  matrix  $I$  such that  $IA = A$ , that is,

$$\begin{bmatrix} ? & ? & ? \\ ? & ? & ? \\ ? & ? & ? \end{bmatrix} \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix} = \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix}$$

**Definition M.2.2** The identity matrix  $I_n$  (or just  $I$  when  $n$  is obvious from context) is the  $n \times n$  matrix

$$I_n = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & \vdots \\ \vdots & & \ddots & \\ 0 & \dots & 0 & 1 \end{bmatrix}.$$

It has a 1 on each diagonal element and a 0 in every other position.

**Fact M.2.3** For any square matrix  $A$ ,  $IA = AI = A$ :

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix} = \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix}$$

**Activity M.2.4** ( $\sim 15$  min) Each row operation can be interpreted as a type of matrix multiplication.

*Part 1:* Tweak the identity matrix slightly to create a matrix that doubles the third row of  $A$ :

$$\begin{bmatrix} ? & ? & ? \\ ? & ? & ? \\ ? & ? & ? \end{bmatrix} \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix} = \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 2 & 2 & -2 \end{bmatrix}$$

*Part 2:* Create a matrix that swaps the second and third rows of  $A$ :

$$\begin{bmatrix} ? & ? & ? \\ ? & ? & ? \\ ? & ? & ? \end{bmatrix} \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix} = \begin{bmatrix} 2 & 7 & -1 \\ 1 & 1 & -1 \\ 0 & 3 & 2 \end{bmatrix}$$

*Part 3:* Create a matrix that adds 5 times the third row of  $A$  to the first row:

$$\begin{bmatrix} ? & ? & ? \\ ? & ? & ? \\ ? & ? & ? \end{bmatrix} \begin{bmatrix} 2 & 7 & -1 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix} = \begin{bmatrix} 2+5 & 7+5 & -1-5 \\ 0 & 3 & 2 \\ 1 & 1 & -1 \end{bmatrix}$$

**Fact M.2.5** If  $R$  is the result of applying a row operation to  $I$ , then  $RA$  is the result of applying the same row operation to  $A$ .

This means that for any matrix  $A$ , we can find a series of matrices  $R_1, \dots, R_k$  corresponding to the row operations such that

$$R_1 R_2 \cdots R_k A = \text{RREF}(A).$$

That is, row reduction can be thought of as the result of matrix multiplication.

**Activity M.2.6** ( $\sim 15$  min) Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear map with standard matrix  $A$ . Sort the following items into groups of statements about  $T$ .

- |  |  |
|--|--|
| (a) $T$ is injective (i.e. one-to-one)                               | (g) The columns of $A$ span $\mathbb{R}^m$           |
| (b) $T$ is surjective (i.e. onto)                                    | (h) The columns of $A$ are linearly independent      |
| (c) $T$ is bijective (i.e. both injective and surjective)            | (i) The columns of $A$ are a basis of $\mathbb{R}^m$ |
| (d) $AX = B$ has a solution for all $m \times 1$ matrices $B$        | (j) Every column of $\text{RREF}(A)$ has a pivot     |
| (e) $AX = B$ has a unique solution for all $m \times 1$ matrices $B$ | (k) Every row of $\text{RREF}(A)$ has a pivot        |
| (f) $AX = 0$ has a unique solution.                                  | (l) $m = n$ and $\text{RREF}(A) = I$                 |

**Activity M.2.7** ( $\sim 5$  min) Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear map with matrix  $A$ . If  $T$  is injective, which of the following cannot be true?

- (a)  $A$  has strictly more columns than rows
- (b)  $A$  has the same number of rows as columns (i.e.  $A$  is square)
- (c)  $A$  has strictly more rows than columns

**Activity M.2.8** ( $\sim 5$  min) Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear map with matrix  $A$ . If  $T$  is surjective, which of the following cannot be true?

- (a)  $A$  has strictly more columns than rows
- (b)  $A$  has the same number of rows as columns (i.e.  $A$  is square)
- (c)  $A$  has strictly more rows than columns

**Activity M.2.9** ( $\sim 5$  min) Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear map with matrix  $A$ . If  $T$  is bijective, which of the following cannot be true?

- (a)  $A$  has strictly more columns than rows
- (b)  $A$  has the same number of rows as columns (i.e.  $A$  is square)
- (c)  $A$  has strictly more rows than columns

## Section M.3

**Definition M.3.1** Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be a linear map with standard matrix  $A$ .

- If  $T$  is a bijection and  $B$  is any  $\mathbb{R}^n$  vector, then  $T(X) = AX = B$  has a unique solution  $X$ .
- So we may define an **inverse map**  $T^{-1} : \mathbb{R}^n \rightarrow \mathbb{R}^n$  by setting  $T^{-1}(B) = X$  to be this unique solution.
- Let  $A^{-1}$  be the standard matrix for  $T^{-1}$ . We call  $A^{-1}$  the **inverse matrix** of  $A$ , so we also say that  $A$  is **invertible**.

**Activity M.3.2** ( $\sim 10$  min) Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the bijective linear map defined by  $T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} 2x - 3y \\ -3x + 5y \end{bmatrix}$ .

It can be shown that  $T$  is bijective and has the inverse map  $T^{-1}\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} 5x + 3y \\ 3x + 2y \end{bmatrix}$ .

*Part 1:* Compute  $(T^{-1} \circ T)\left(\begin{bmatrix} -2 \\ 1 \end{bmatrix}\right)$ .

*Part 2:* If  $A$  is the standard matrix for  $T$  and  $A^{-1}$  is the standard matrix for  $T^{-1}$ , what must  $A^{-1}A$  be?

**Observation M.3.3**  $T^{-1} \circ T = T \circ T^{-1}$  is the identity map for any bijective linear transformation  $T$ . Therefore  $A^{-1}A = AA^{-1} = I$  is the identity matrix for any invertible matrix  $A$ .

**Activity M.3.4** ( $\sim 20$  min) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be given by the matrix  $A = \begin{bmatrix} 2 & -1 & -6 \\ 2 & 1 & 3 \\ 1 & 1 & 4 \end{bmatrix}$ .

*Part 1:* Solve  $T(X) = \mathbf{e}_1$  to find  $T^{-1}(\mathbf{e}_1)$ .

*Part 2:* Solve  $T(X) = \mathbf{e}_2$  to find  $T^{-1}(\mathbf{e}_2)$ .

*Part 3:* Solve  $T(X) = \mathbf{e}_3$  to find  $T^{-1}(\mathbf{e}_3)$ .

*Part 4:* Compute  $A^{-1}$ , the standard matrix for  $T^{-1}$ .

**Observation M.3.5** We could have solved these three systems simultaneously by row reducing the matrix  $[A \mid I]$  at once.

$$A = \left[ \begin{array}{ccc|ccc} 2 & -1 & -6 & 1 & 0 & 0 \\ 2 & 1 & 3 & 0 & 1 & 0 \\ 1 & 1 & 4 & 0 & 0 & 1 \end{array} \right] \sim \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & -2 & 3 \\ 0 & 1 & 0 & -5 & 14 & -18 \\ 0 & 0 & 1 & 1 & -3 & 4 \end{array} \right]$$

**Activity M.3.6** ( $\sim 10$  min) Find the inverse  $A^{-1}$  of the matrix  $A = \begin{bmatrix} 1 & 3 \\ 0 & -2 \end{bmatrix}$  by row-reducing  $[A \mid I]$ .

**Activity M.3.7** ( $\sim 10$  min) Is the matrix  $\begin{bmatrix} 2 & 3 & 1 \\ -1 & -4 & 2 \\ 0 & -5 & 5 \end{bmatrix}$  invertible? Give a reason for your answer.

**Observation M.3.8** A matrix  $A \in \mathbb{R}^{n \times n}$  is invertible if and only if  $\text{RREF}(A) = I_n$ .

## Module G: Geometry of Linear Maps

### Standards for this Module

**How can we understand linear maps geometrically?** At the end of this module, students will be able to...

**G1. Row operations.** ... represent a row operation as matrix multiplication, and compute how the operation affects the determinant.

**G2. Determinants.** ... compute the determinant of a square matrix.

**G3. Eigenvalues.** ... find the eigenvalues of a  $2 \times 2$  matrix.

**G4. Eigenvectors.** ... find a basis for the eigenspace of a square matrix associated with a given eigenvalue.

### Readiness Assurance Outcomes

Before beginning this module, each student should be able to...

- Calculate the area of a parallelogram.
- Find the matrix corresponding to a linear transformation of Euclidean spaces **A1**.
- Recall and use the definition of a linear transformation **A2**.
- Find all roots of quadratic polynomials (including complex ones), and be able to use the rational root theorem to find all rational roots of a higher degree polynomial.
- Interpret the statement “ $A$  is an invertible matrix” in many equivalent ways in different contexts.

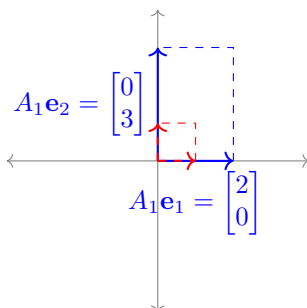
### Readiness Assurance Resources

The following resources will help you prepare for this module.

- Finding the area of a parallelogram (Khan Academy): <http://bit.ly/2B05iWx>
- Factoring quadratics (Khan Academy): <http://bit.ly/1XjfbV2>
- Finding complex roots of quadratics (Khan Academy): <http://bit.ly/1HH3yAA>

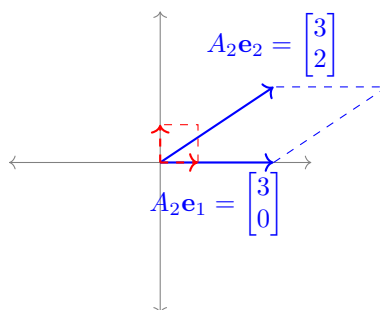
## Section G.1

**Activity G.1.1** ( $\sim 5$  min) The image below illustrates how the linear transformation  $T_1 : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  given by the standard matrix  $A_1 = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$  transforms the unit square.



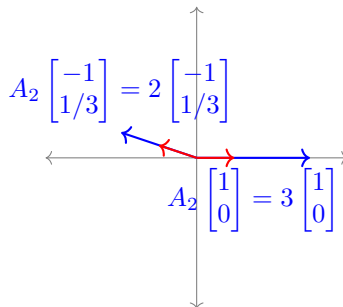
- What is the area of the transformed unit square?
- Find two vectors that were stretched/compressed by the transformation (not sheared), and compute how much those vectors were stretched/compressed.

**Activity G.1.2** ( $\sim 5$  min) The image below illustrates how the linear transformation  $T_2 : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  given by the standard matrix  $A_2 = \begin{bmatrix} 3 & 3 \\ 0 & 2 \end{bmatrix}$  transforms the unit square.



- What is the area of the transformed unit square?
- Find at least one vector that was stretched/compressed by the transformation (not sheared), and compute how much those vectors were stretched/compressed.

**Observation G.1.3** It's possible to find two non-parallel vectors that are stretched by the transformation given by  $A_2$ :



The process for finding such vectors will be covered later in this module.

**Activity G.1.4** ( $\sim 5$  min) Consider the linear transformation given by the standard matrix  $A_3 = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$ .

- Sketch the transformation of the unit square (the parallelogram given by the columns of the standard matrix).
- Compute the area of the transformed unit square.

**Activity G.1.5** ( $\sim 5$  min) Consider the linear transformation given by the standard matrix  $A_4 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ .

- Sketch the transformation of the unit square.
- Compute the area of the transformed unit square.

**Activity G.1.6** ( $\sim 5$  min) Consider the linear transformation given by the standard matrix  $A_5 = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ .

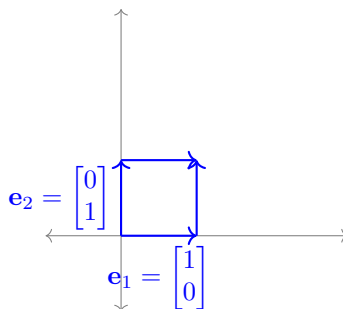
- Sketch the transformation of the unit square.
- Compute the area of the transformed unit square.

**Remark G.1.7** The area of the transformed unit square measures the factor by which all areas are transformed by a linear transformation.

We will define the **determinant** of a square matrix  $A$ , or  $\det(A)$  for short, to be this factor. But what properties must this function satisfy?

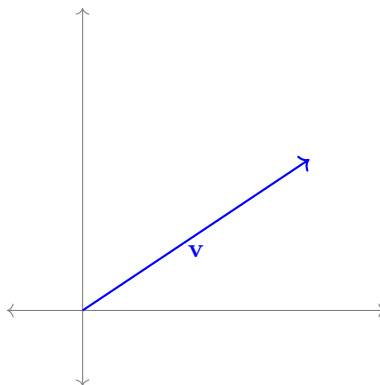


**Activity G.1.8** ( $\sim 2$  min) The transformation of the unit square by the standard matrix  $[\mathbf{e}_1 \ \mathbf{e}_2] = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$  is illustrated below. What is  $\det([\mathbf{e}_1 \ \mathbf{e}_2]) = \det(I)$ , that is, by what factor has the area of the unit square been scaled?



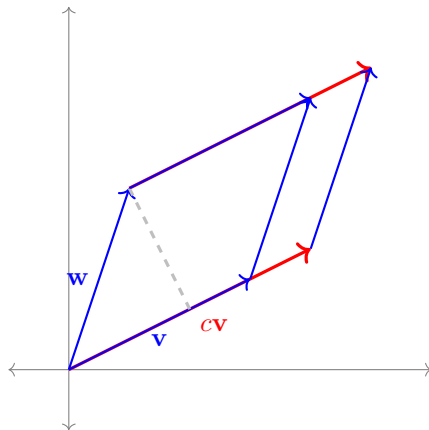
- a) 0
- b) 1
- c) 2
- d) Cannot be determined

**Activity G.1.9** ( $\sim 2$  min) The transformation of the unit square by the standard matrix  $[\mathbf{v} \ \mathbf{v}]$  is illustrated below: both  $T(\mathbf{e}_1) = T(\mathbf{e}_2) = \mathbf{v}$ . What is  $\det([\mathbf{v} \ \mathbf{v}])$ , that is, by what factor has area been scaled?



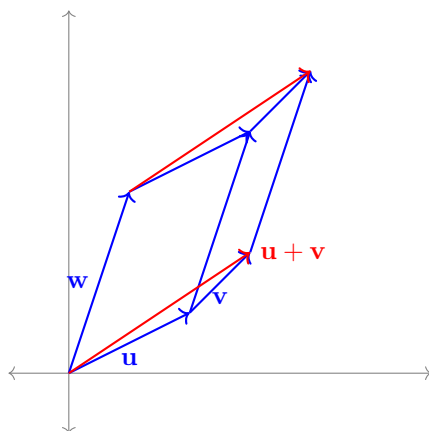
- a) 0
- b) 1
- c) 2
- d) Cannot be determined

**Activity G.1.10** (*~5 min*) The transformations of the unit square by the standard matrices  $[\mathbf{v} \ \mathbf{w}]$  and  $[c\mathbf{v} \ \mathbf{w}]$  are illustrated below. How are  $\det([\mathbf{v} \ \mathbf{w}])$  and  $\det([c\mathbf{v} \ \mathbf{w}])$  related?



- a)  $\det([\mathbf{v} \ \mathbf{w}]) = \det([c\mathbf{v} \ \mathbf{w}])$
- b)  $c + \det([\mathbf{v} \ \mathbf{w}]) = \det([c\mathbf{v} \ \mathbf{w}])$
- c)  $c \det([\mathbf{v} \ \mathbf{w}]) = \det([c\mathbf{v} \ \mathbf{w}])$

**Activity G.1.11** (*~5 min*) The transformations of unit squares by the standard matrices  $[\mathbf{u} \ \mathbf{w}]$ ,  $[\mathbf{v} \ \mathbf{w}]$  and  $[\mathbf{u} + \mathbf{v} \ \mathbf{w}]$  are illustrated below. How is  $\det([\mathbf{u} + \mathbf{v} \ \mathbf{w}])$  related to  $\det([\mathbf{u} \ \mathbf{w}])$  and  $\det([\mathbf{v} \ \mathbf{w}])$ ?



- a)  $\det([\mathbf{u} \ \mathbf{w}]) = \det([\mathbf{v} \ \mathbf{w}]) = \det([\mathbf{u} + \mathbf{v} \ \mathbf{w}])$
- b)  $\det([\mathbf{u} \ \mathbf{w}]) + \det([\mathbf{v} \ \mathbf{w}]) = \det([\mathbf{u} + \mathbf{v} \ \mathbf{w}])$
- c)  $\det([\mathbf{u} \ \mathbf{w}]) \det([\mathbf{v} \ \mathbf{w}]) = \det([\mathbf{u} + \mathbf{v} \ \mathbf{w}])$

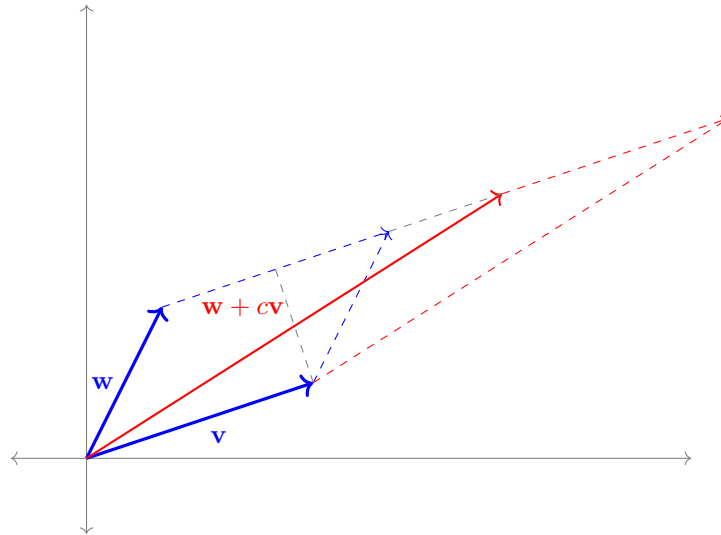
**Definition G.1.12** The **determinant** is the unique function  $\det : \mathbb{R}^{n \times n} \rightarrow \mathbb{R}$  satisfying the following three properties:

P1:  $\det(I) = 1$

P2:  $\det([\mathbf{v}_1 \ \mathbf{v}_2 \ \cdots \ \mathbf{v}_n]) = 0$  whenever two columns of the matrix are identical.

P3:  $\det[\cdots \ c\mathbf{v} + d\mathbf{w} \ \cdots] = c \det[\cdots \ \mathbf{v} \ \cdots] + d \det[\cdots \ \mathbf{w} \ \cdots]$ , assuming all other columns are equal.

**Observation G.1.13** What happens if we had a multiple of one column to another?



The base of both parallelograms is  $\mathbf{v}$ , while the height has not changed. Thus

$$\det([\mathbf{v} \ \mathbf{w} + c\mathbf{v}]) = \det([\mathbf{v} \ \mathbf{w}])$$

**Observation G.1.14** Swapping columns can be obtained from a sequence of adding column multiples.

$$\begin{aligned} \det([\mathbf{v} \ \mathbf{w}]) &= \det([\mathbf{v} + \mathbf{w} \ \mathbf{w}]) \\ &= \det([\mathbf{v} + \mathbf{w} \ \mathbf{w} - (\mathbf{v} + \mathbf{w})]) \\ &= \det([\mathbf{v} + \mathbf{w} \ -\mathbf{v}]) \\ &= \det([\mathbf{v} + \mathbf{w} - \mathbf{v} \ -\mathbf{v}]) \\ &= \det([\mathbf{w} \ -\mathbf{v}]) \\ &= -\det([\mathbf{w} \ \mathbf{v}]) \end{aligned}$$

So swapping two columns results in a negation of the determinant. Therefore, determinants represent a *signed* area, since they are not always positive.

**Fact G.1.15** We've shown that the column versions of the three row-reducing operations a matrix may be used to simplify a determinant:

- (a) Multiplying a column by a scalar multiplies the determinant by that scalar:

$$c \det([\cdots \mathbf{v} \cdots]) = \det([\cdots c\mathbf{v} \cdots])$$

- (b) Swapping two columns changes the sign of the determinant:

$$\det([\cdots \mathbf{v} \cdots \mathbf{w} \cdots]) = -\det([\cdots \mathbf{w} \cdots \mathbf{v} \cdots])$$

- (c) Adding a multiple of a column to another column does not change the determinant:

$$\det([\cdots \mathbf{v} \cdots \mathbf{w} \cdots]) = \det([\cdots \mathbf{v} + c\mathbf{w} \cdots \mathbf{w} \cdots])$$

**Activity G.1.16** ( $\sim 5$  min) The transformation given by the standard matrix  $A$  scales areas by 4, and the transformation given by the standard matrix  $B$  scales areas by 3. How must the transformation given by the standard matrix  $AB$  scale areas?

- (a) 1
- (b) 7
- (c) 12
- (d) Cannot be determined

**Fact G.1.17** Since the transformation given by the standard matrix  $AB$  is obtained by applying the transformations given by  $A$  and  $B$ , it follows that

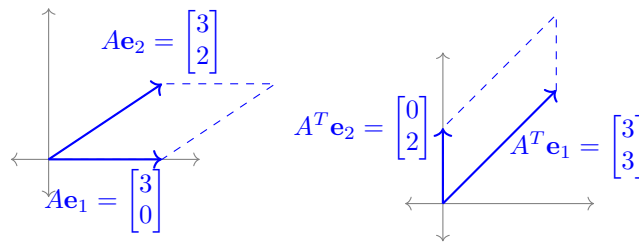
$$\det(AB) = \det(A) \det(B)$$

## Section G.2

**Definition G.2.1** The **transpose** of a matrix is given by rewriting its columns as rows and vice versa:

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}^T = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$$

**Fact G.2.2** It is possible to prove that the determinant of a matrix and its transpose are the same. For example, let  $A = \begin{bmatrix} 3 & 3 \\ 0 & 2 \end{bmatrix}$ , so  $A^T = \begin{bmatrix} 3 & 0 \\ 3 & 2 \end{bmatrix}$ ; both matrices scale the unit square by 6, even though the parallelograms are not congruent.



**Fact G.2.3** We previously figured out that column operations can be used to simplify determinants; since  $\det(A) = \det(A^T)$ , we can also use row operations:

1. Multiplying rows by scalars:  $\det \begin{bmatrix} \vdots \\ cR \\ \vdots \end{bmatrix} = c \det \begin{bmatrix} \vdots \\ R \\ \vdots \end{bmatrix}$

2. Swapping two rows:  $\det \begin{bmatrix} \vdots \\ R \\ \vdots \\ S \\ \vdots \end{bmatrix} = -\det \begin{bmatrix} \vdots \\ S \\ \vdots \\ R \\ \vdots \end{bmatrix}$

3. Adding multiples of rows to other rows:  $\det \begin{bmatrix} \vdots \\ R \\ \vdots \\ S \\ \vdots \end{bmatrix} = \det \begin{bmatrix} \vdots \\ R + cS \\ \vdots \\ S \\ \vdots \end{bmatrix}$

**Activity G.2.4** ( $\sim 10$  min) Compute the determinant of  $\begin{bmatrix} 2 & 4 \\ 2 & 3 \end{bmatrix}$  by row reducing it to a nicer matrix.

For example,  $\det \begin{bmatrix} 2 & 4 \\ 2 & 3 \end{bmatrix} = 2 \det \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix}$ .

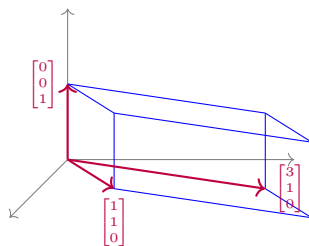
**Fact G.2.5** This same process allows us to prove a more convenient formula:

$$\det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = ad - bc$$

In higher dimensions, the formulas become unreasonable. For example, the formula for  $4 \times 4$  matrices has 24 terms!

**Activity G.2.6** ( $\sim 5$  min) The following image illustrates the transformation of the unit cube by the matrix

$$\begin{bmatrix} 3 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$



This volume is equal to which of the following areas?

(a)  $\det \begin{bmatrix} 3 & 1 \\ 1 & 1 \end{bmatrix}$

(b)  $\det \begin{bmatrix} 3 & 1 \\ 1 & 0 \end{bmatrix}$

(c)  $\det \begin{bmatrix} 3 & 1 \\ 0 & 1 \end{bmatrix}$

(d)  $\det \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$

**Fact G.2.7** If column  $i$  of a matrix is  $\mathbf{e}_i$ , then both column and row  $i$  may be removed without changing the value of the determinant. For example, the second column of the following matrix is  $\mathbf{e}_2$ , so:

$$\det \begin{bmatrix} 3 & 0 & -1 & 5 \\ 2 & 1 & 4 & 0 \\ -1 & 0 & 1 & 11 \\ 3 & 0 & 0 & 1 \end{bmatrix} = \det \begin{bmatrix} 3 & -1 & 5 \\ -1 & 1 & 11 \\ 3 & 0 & 1 \end{bmatrix}$$

Therefore the same holds for the transpose:

$$\det \begin{bmatrix} 3 & 2 & -1 & 3 \\ 0 & 1 & 0 & 0 \\ -1 & 4 & 1 & 0 \\ 5 & 0 & 11 & 1 \end{bmatrix} = \det \begin{bmatrix} 3 & -1 & 3 \\ -1 & 1 & 0 \\ 5 & 11 & 1 \end{bmatrix}$$

Geometrically, this is the fact that if the height is 1, the base  $\times$  height formula reduces to the area/volume/etc. of the  $n - 1$  dimensional base.

**Activity G.2.8** ( $\sim 5$  min) Compute  $\det \begin{bmatrix} 1 & 0 & 0 \\ 1 & 5 & 12 \\ 3 & 2 & -1 \end{bmatrix}$ .

**Activity G.2.9** ( $\sim 5$  min) Compute  $\det \begin{bmatrix} 0 & 3 & -2 \\ 1 & 5 & 12 \\ 0 & 2 & -1 \end{bmatrix}$ .

(a)  $-1$

(b)  $0$

(c)  $1$

**Activity G.2.10** ( $\sim 10$  min) Compute  $\det \begin{bmatrix} 1 & 2 & 3 \\ 1 & -2 & -5 \\ 0 & 3 & 3 \end{bmatrix}$

*Hint:*  $\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ .

(a)  $3$

(b)  $6$

(c)  $9$

(d)  $12$

**Activity G.2.11** ( $\sim 15$  min) Compute  $\det \begin{bmatrix} 2 & 3 & 5 & 0 \\ 0 & 3 & 2 & 0 \\ 1 & 2 & 0 & 3 \\ -1 & -1 & 2 & 2 \end{bmatrix}$ .

**Observation G.2.12**

$$\begin{aligned}
 \det \begin{bmatrix} 2 & 3 & 5 & 0 \\ 0 & 3 & 2 & 0 \\ 1 & 2 & 0 & 3 \\ -1 & -1 & 2 & 2 \end{bmatrix} &= (-1)(0) \det \begin{bmatrix} \cancel{2} & 3 & 5 & 0 \\ \cancel{0} & \cancel{3} & \cancel{2} & \cancel{0} \\ 1 & 2 & 0 & 3 \\ -1 & -1 & 2 & 2 \end{bmatrix} + (1)(3) \det \begin{bmatrix} 2 & \cancel{3} & 5 & 0 \\ \cancel{0} & \cancel{3} & \cancel{2} & \cancel{0} \\ 1 & 2 & 0 & 3 \\ -1 & -1 & 2 & 2 \end{bmatrix} + \\
 &(-1)(2) \det \begin{bmatrix} 2 & 3 & \cancel{5} & 0 \\ \cancel{0} & \cancel{3} & \cancel{2} & \cancel{0} \\ 1 & 2 & 0 & 3 \\ -1 & -1 & \cancel{2} & 2 \end{bmatrix} + (1)(0) \det \begin{bmatrix} 2 & 3 & 5 & \cancel{0} \\ \cancel{0} & \cancel{3} & \cancel{2} & \cancel{0} \\ 1 & 2 & 0 & \cancel{3} \\ -1 & -1 & 2 & \cancel{2} \end{bmatrix} \\
 &= 3 \det \begin{bmatrix} 2 & 5 & 0 \\ 1 & 0 & 3 \\ -1 & 2 & 2 \end{bmatrix} + (-1)(2) \det \begin{bmatrix} 2 & 3 & 0 \\ 1 & 2 & 3 \\ -1 & -1 & 2 \end{bmatrix}
 \end{aligned}$$

This technique is called **Laplace expansion** or **cofactor expansion**.

**Activity G.2.13** (*~10 min*) Compute  $\det \begin{bmatrix} 1 & 2 & 1 & 0 \\ 0 & 3 & 2 & -1 \\ 1 & 2 & 0 & 3 \\ -1 & -3 & 2 & -2 \end{bmatrix}$ .



### Section G.3

**Activity G.3.1** (*~5 min*) An invertible matrix  $M$  and its inverse  $M^{-1}$  are given below:

$$M = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad M^{-1} = \begin{bmatrix} -2 & 1 \\ 3/2 & -1/2 \end{bmatrix}$$

Compute  $\det(M)$  and  $\det(M^{-1})$ .

**Activity G.3.2** (*~5 min*) Suppose the matrix  $M$  is invertible, so there exists  $M^{-1}$  with  $MM^{-1} = I$ . It follows that  $\det(M)\det(M^{-1}) = \det(I)$ .

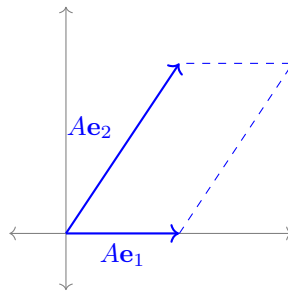
What is the only number that  $\det(M)$  cannot equal?

- (a)  $-1$                       (b)  $0$                       (c)  $1$                       (d)  $\frac{1}{\det(M^{-1})}$

#### Fact G.3.3

- For every invertible matrix  $M$ ,  $\det(M^{-1}) = \frac{1}{\det(M)}$ .
- Furthermore, a square matrix  $M$  is invertible if and only if  $\det(M) \neq 0$ .

**Observation G.3.4** Consider the linear transformation  $A : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  given by the matrix  $A = \begin{bmatrix} 2 & 2 \\ 0 & 3 \end{bmatrix}$



It is easy to see geometrically that

$$A \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

It is less obvious (but easily verified by computation) that

$$A \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 6 \\ 3 \end{bmatrix} = 3 \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

**Definition G.3.5** Let  $A \in \mathbb{R}^{n \times n}$ . An **eigenvector** is a vector  $\mathbf{x} \in \mathbb{R}^n$  such that  $A\mathbf{x}$  is parallel to  $\mathbf{x}$ . In other words,  $A\mathbf{x} = \lambda\mathbf{x}$  for some scalar  $\lambda$ .

We call this  $\lambda$  an **eigenvalue** of  $A$ .

**Observation G.3.6** Since  $\lambda\mathbf{x} = \lambda(I\mathbf{x})$ , we can find the eigenvalues and eigenvectors satisfying  $A\mathbf{x} = \lambda\mathbf{x}$  by inspecting  $(A - \lambda I)\mathbf{x} = \mathbf{0}$ .

- Since we already know that  $(A - \lambda I)\mathbf{0} = \mathbf{0}$  for any value of  $\lambda$ , we are more interested in finding values of  $\lambda$  such that  $A - \lambda I$  has a nontrivial kernel.
- Thus  $\text{RREF}(A - \lambda I)$  must have a non-pivot column, and therefore  $A - \lambda I$  cannot be invertible.
- Since  $A - \lambda I$  cannot be invertible, our eigenvalues must satisfy  $\det(A - \lambda I) = 0$ .

**Definition G.3.7** Computing  $\det(A - \lambda I)$  results in the **characteristic polynomial** of  $A$ .

For example, when  $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ , we have

$$A - \lambda I = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} 1 - \lambda & 2 \\ 3 & 4 - \lambda \end{bmatrix}$$

Thus the characteristic polynomial of  $A$  is

$$\det \begin{bmatrix} 1 - \lambda & 2 \\ 3 & 4 - \lambda \end{bmatrix} = (1 - \lambda)(4 - \lambda) - 6 = \lambda^2 - 5\lambda - 2$$

**Activity G.3.8** (*~15 min*) Compute  $\det(A - \lambda I)$  to find the characteristic polynomial of  $A = \begin{bmatrix} 6 & -2 & 1 \\ 17 & -5 & 5 \\ -4 & 2 & 1 \end{bmatrix}$ .

**Activity G.3.9** (*~15 min*) Let  $A = \begin{bmatrix} 2 & 2 \\ 0 & 3 \end{bmatrix}$ .

*Part 1:* Compute  $\det(A - \lambda I)$  to determine the characteristic polynomial of  $A$ .

*Part 2:* Find the roots of the characteristic polynomial to determine the eigenvalues of  $A$ .

*Part 3:* Compute the kernel of the transformation given by

$$A - 2I = \begin{bmatrix} 2 - 2 & 2 \\ 0 & 3 - 2 \end{bmatrix}$$

to determine all the eigenvectors associated to the eigenvalue 2.

*Part 4:* Compute the kernel of the transformation given by  $A - 3I$  to determine all the eigenvectors associated to the eigenvalue 3.

**Definition G.3.10** The kernel of the transformation given by  $A - \lambda I$  contains all the eigenvectors associated with  $\lambda$ . Since kernel is a subspace of  $\mathbb{R}^n$ , we call this kernel the **eigenspace** associated with the eigenvalue  $\lambda$ .

**Activity G.3.11** (*~15 min*) Find all the eigenvalues and associated eigenspaces for the matrix  $A = \begin{bmatrix} 3 & -2 & 1 \\ 0 & 2 & 8 \\ 0 & 2 & 2 \end{bmatrix}$ .

*Part 1:* Compute  $\det(A - \lambda I)$  to determine the characteristic polynomial of  $A$ .

*Part 2:* Find the roots of the characteristic polynomial  $(3 - \lambda)(\lambda^2 - 4\lambda - 12)$  to determine the eigenvalues of  $A$ .

*Part 3:* Compute the kernels of  $A - \lambda I$  for each eigenvalue  $\lambda \in \{-2, 3, 6\}$  to determine the respective eigenspaces.

**Observation G.3.12** Recall that if  $a$  is a root of the polynomial  $p(\lambda)$ , the **multiplicity** of  $a$  is the largest number  $k$  such that  $p(\lambda) = q(\lambda)(\lambda - a)^k$  for some polynomial  $q(\lambda)$ .

For this reason, the **algebraic multiplicity** of an eigenvalue is its multiplicity as a root of the characteristic polynomial.

**Example G.3.13** If  $A = \begin{bmatrix} 3 & 1 & -1 \\ 0 & 3 & 3 \\ 0 & 0 & -1 \end{bmatrix}$ , the characteristic polynomial is  $p(\lambda) = (\lambda - 3)^2(\lambda + 1)$ .

The eigenvalues are 3 (with algebraic multiplicity 2) and  $-1$  (with algebraic multiplicity 1).

## Section G.4

**Observation G.4.1** Recall from last class:

- To find the eigenvalues of a matrix  $A$ , we need to find values of  $\lambda$  such that  $A - \lambda I$  has a nontrivial kernel. Equivalently, we want values where  $A - \lambda I$  is not invertible, so we want to know the values of  $\lambda$  where  $\det(A - \lambda I) = 0$ .
- $\det(A - \lambda I)$  is a polynomial with variable  $\lambda$ , called the **characteristic polynomial** of  $A$ . Thus the roots of the characteristic polynomial of  $A$  are exactly the eigenvalues of  $A$ .
- Once an eigenvalue  $\lambda$  is found, the **eigenspace** containing all **eigenvectors**  $\mathbf{x}$  satisfying  $A\mathbf{x} = \lambda\mathbf{x}$  is given by  $\ker(A - \lambda I)$ .

**Activity G.4.2** ( $\sim 5$  min) Let  $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ .

*Part 1:* Compute the eigenvalues of  $A$ .

*Part 2:* Sketch a picture of the transformation of the unit square. What about this picture reveals that  $A$  has no real eigenvectors?

**Activity G.4.3** ( $\sim 5$  min) If  $A$  is a  $4 \times 4$  matrix, what is the largest number of eigenvalues  $A$  can have?

- (a) 3
- (b) 4
- (c) 5
- (d) 6
- (e) It can have infinitely many

**Observation G.4.4** An  $n \times n$  matrix may have between 0 and  $n$  real-valued eigenvalues. But the Fundamental Theorem of Algebra implies that if complex eigenvalues are included, then every  $n \times n$  matrix has exactly  $n$  eigenvalues (counting algebraic multiplicities).

**Activity G.4.5** ( $\sim 5$  min) The matrix  $A = \begin{bmatrix} 1 & -2 & 1 \\ -1 & 0 & 1 \\ -1 & -2 & 3 \end{bmatrix}$  has characteristic polynomial  $-\lambda(\lambda - 2)^2$ .

Find the dimension of the eigenspace of  $A$  associated to the eigenvalue 2 (the dimension of the kernel of  $A - 2I$ ).

**Activity G.4.6** ( $\sim 5$  min) The matrix  $B = \begin{bmatrix} -3 & -9 & 5 \\ -2 & -2 & 2 \\ -7 & -13 & 9 \end{bmatrix}$  has characteristic polynomial  $-\lambda(\lambda - 2)^2$ .

Find the dimension of the eigenspace of  $B$  associated to the eigenvalue 2 (the dimension of the kernel of  $B - 2I$ ).

**Observation G.4.7** In the first example, the (2 dimensional) plane spanned by  $\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$  and  $\begin{bmatrix} 4 \\ -2 \\ 0 \end{bmatrix}$  was preserved. In the second example, only the (one dimensional) line spanned by  $\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$  is preserved.

**Definition G.4.8** While the **algebraic multiplicity** of an eigenvalue is its multiplicity as a root of the characteristic polynomial, the **geometric multiplicity** of an eigenvalue is the dimension of its eigenspace.

**Fact G.4.9** As we've seen, the geometric multiplicity may be different than its algebraic multiplicity, but it cannot exceed it.

This fact is explored deeper and explained in Math 316, Linear Algebra II

**Activity G.4.10** (*~20 min*) Consider the  $4 \times 4$  matrix

$$\begin{bmatrix} -3 & 1 & 2 & 1 \\ -9 & 5 & -2 & -1 \\ 31 & -17 & 6 & 3 \\ -69 & 39 & -18 & -9 \end{bmatrix}$$

*Part 1:* Use technology (e.g. Wolfram Alpha) to find its characteristic polynomial.

*Part 2:* Find the algebraic and geometric multiplicities for both eigenvalues.

## Module P: Applications of Linear Algebra

## Section P.1

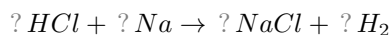
**Definition P.1.1** In chemistry, we learn that when the two substances

- Hydrochloric acid  $HCl$  (formed from 1  $H$  and 1  $Cl$  atom)
- Sodium  $Na$  (formed from 1  $Na$  atom)

react, their atoms rearrange to form the substances

- Salt  $NaCl$  (formed from 1  $Na$  and 1  $Cl$  atom)
- Hydrogen gas  $H_2$  (formed from 2  $H$  atoms).

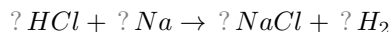
This may be represented by the **chemical equation**



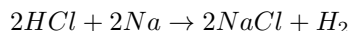
where each  $?$  represents the amount of that substance before/after the reaction.

**Activity P.1.2** ( $\sim 5$  min) The **law of conservation of mass** states that the quantity of atoms before and after a chemical reaction must remain the same.

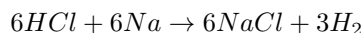
Find positive integers so that both sides of the chemical equation represent the same amount of matter:



**Definition P.1.3** A chemical equation is **balanced** if the given quantities of each substance before and after the reaction are equal and minimal positive integers:

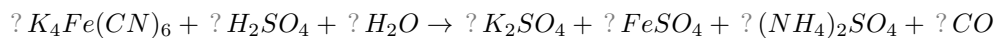
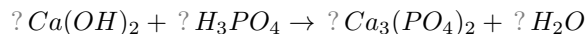
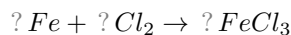


**Observation P.1.4** For example, the following equation isn't balanced because all the integers may be divided by three:



Therefore if a chemical equation can be balanced, there is exactly one correct solution.

**Activity P.1.5** ( $\sim 15$  min) Balance the following chemical equations:

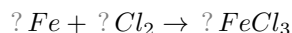


(Note that  $(NH_4)_2SO_4$  represents 2  $N$ , 8  $H$ , 1  $S$ , and 4  $O$ .)

**Observation P.1.6** For the purposes of balancing chemical equations, the set

$$L = \{\mathbf{A} \mid \mathbf{A} \text{ is combination of elements}\}$$

may be treated as a kind of **vector space**. This means that balancing the chemical equation



may be achieved by finding a solution  $\begin{bmatrix} x \\ y \\ z \end{bmatrix}$  to the vector equation

$$x\mathbf{Fe} + y(2\mathbf{Cl}) = z(\mathbf{Fe} + 3\mathbf{Cl}).$$

**Activity P.1.7** ( $\sim 5 \text{ min}$ ) To solve the vector equation

$$x\mathbf{Fe} + y(2\mathbf{Cl}) = z(\mathbf{Fe} + 3\mathbf{Cl})$$

we are only concerned with the subspace  $W = \text{span}\{\mathbf{Cl}, \mathbf{Fe}\}$  of  $L$ . Since the element  $\mathbf{Fe}$  cannot be created from the element  $\mathbf{Cl}$  in a chemical reaction and vice versa, the set  $\{\mathbf{Cl}, \mathbf{Fe}\}$ :

- a) spans  $W$ , but is linearly dependent.
- b) is linearly independent, but does not span  $W$ .
- c) is a basis for  $W$ .

**Observation P.1.8**  $W = \text{span}\{\mathbf{Cl}, \mathbf{Fe}\}$  is a two-dimensional subspace of  $L$ , so as usual we'd rather work with its isomorphic Euclidean space  $\mathbb{R}^2$ .

Thus we should assign a transformation of bases such as:

$$\mathbf{Cl} \leftrightarrow \mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \mathbf{Fe} \leftrightarrow \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$



**Activity P.1.9** (*~10 min*) Rewrite the  $W = \text{span}\{\mathbf{Cl}, \mathbf{Fe}\}$  vector equation

$$x\mathbf{Fe} + y(2\mathbf{Cl}) = z(\mathbf{Fe} + 3\mathbf{Cl})$$

using the transformation of bases

$$\mathbf{Cl} \leftrightarrow \mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \mathbf{Fe} \leftrightarrow \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

and show how it may be simplified to

$$x \begin{bmatrix} 0 \\ 1 \end{bmatrix} + y \begin{bmatrix} 2 \\ 0 \end{bmatrix} - z \begin{bmatrix} 3 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

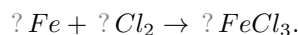
**Activity P.1.10** (*~10 min*) Consider the Euclidean vector equation

$$x \begin{bmatrix} 0 \\ 1 \end{bmatrix} + y \begin{bmatrix} 2 \\ 0 \end{bmatrix} - z \begin{bmatrix} 3 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

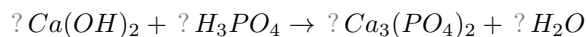
*Part 1:* Find its solution set.

*Part 2:* Find a vector in the solution space that consists of minimal positive integers.

*Part 3:* Balance the chemical equation



**Activity P.1.11** (*~10 min*) Balance the chemical equation



by first converting it into an  $\mathbb{R}^4$  vector equation and finding its solution set.