Multivariable Calculus

August 15, 2018

Contents

Week 0: Review

Vectors

In this section, we review some basics about vectors. This includes the definition of a vector, basic vector operations, standard basis vectors, and notation.

Vectors

In linear algebra, we often worked with vectors. We begin by recalling the (algebraic) definition of a vector in \mathbb{R}^n .

Definition 1. A vector in \mathbb{R}^n is an ordered n-tuple of real numbers. That is, a vector \vec{v} may be written as

$$\vec{v} = (a_1, a_2, ..., a_n)$$

where $a_1, a_2, ..., a_n$ are real numbers.

We call the numbers a_i the components or entries of the vector. We call n the dimension of the vector \vec{v} , and say that \vec{v} is n-dimensional.

We write the vector with an arrow above it, as \vec{v} , in order to make the distinction between vectors and *scalars*, which are just real numbers. Some other common notations for vectors are \mathbf{v} and \hat{v} . It's important to make this distinction between vectors and scalars, so you should make use of one of these notations for vectors.

Example 1. $\vec{v} = (1,3)$ is a vector in \mathbb{R}^2 .

 $\vec{w} = (-1, 5, 0)$ is a vector in \mathbb{R}^3 .

 $\vec{x} = (1, -2, 3)$ is a vector in \mathbb{R}^3 .

 $\vec{y} = (-6, \pi, 1/24, -0.5, 3)$ is a vector in \mathbb{R}^5 .

It's sometimes convenient to write a vector as a column vector instead (particularly when working with linear transformations, which we'll review in a later section). We could write

$$\vec{v} = (a_1, a_2, ..., a_n) = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}$$

Learning outcomes:

Author(s):

or

$$\vec{v} = (a_1, a_2, ..., a_n) = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix}.$$

The choice between square brackets and parentheses is just a difference in notation, they mean the same thing, and you should feel free to use either.

Example 2. We write the following vectors as column vectors.

$$\vec{v} = (1,3) = \begin{bmatrix} 1 \\ 3 \end{bmatrix}.$$

$$\vec{w} = (-1,5,0) = \begin{bmatrix} -1 \\ 5 \\ 0 \end{bmatrix}.$$

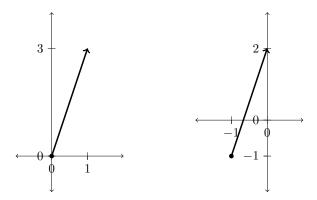
$$\vec{x} = (1,-2,3) = \begin{bmatrix} 1 \\ -2 \\ 3 \end{bmatrix}.$$

$$\vec{y} = (-6,\pi,1/24,-0.5,3) = \begin{bmatrix} -6 \\ \pi \\ 1/24 \\ -0.5 \\ 3 \end{bmatrix}.$$

The Geometric Perspective

We also can think of a vector geometrically, as giving a direction and magnitude, but without a fixed position.

In two or three dimensions, it is useful to visualize a vector as an arrow in \mathbb{R}^n . We might visualize a vector $\vec{v} = (1,3)$ in \mathbb{R}^2 as the arrow starting at the origin and ending at the point (1,3), thus giving a direction and a magnitude. However, we typically don't think of a vector as having a set location. We could also visualize the vector \vec{v} as starting at the point (-1,-1) and ending at the point (0,2). Note that this arrow would have the same direction and magnitude as the one starting at the origin, thus they represent the same vector.



In four or higher dimensions, visualizing anything becomes very difficult. It can still be useful to think of a vector (1, 2, 3, 4, 5) in \mathbb{R}^5 as starting at the origin and ending at the point (1, 2, 3, 4, 5), but you probably won't be able to have a very clear picture of this in your head.

This concept will probably seem more useful once you think about a displacement vector.

Definition 2. Given points $P_1 = (x_1, ..., x_n)$ and $P_2 = (y_1, ..., y_n)$ in \mathbb{R}^n , the displacement vector from P_1 to P_2 is

$$\vec{P_1P_2} = (y_1 - x_1, ..., y_n - x_n).$$

This is the vector that starts at P_1 and ends at P_2 .

Notice that the notation $(a_1, ..., a_n)$ that we use for a vector in \mathbb{R}^n is identical to the notation that we'd use for a point in \mathbb{R}^n . Since both vectors and points in \mathbb{R}^n are defined as n-tuples of points, they are, in some sense, the same thing. The difference between the two comes when we consider the context and geometric significance of the vector or point that we're working with. As we move into multivariable calculus, we'll often blur the distinction between a vector and a point, and sometimes think of a vector as a point and vice versa. This will be greatly simplify notation, and we promise that it won't be as confusing as it sounds!

Vector Operations

Before defining some basic vector operations, we define what it means for two vectors to be equal. This is done by comparing the components of the vectors.

Definition 3. Two vectors $(a_1, a_2, ..., a_n)$ and $(b_1, b_2, ..., b_n)$ in \mathbb{R}^n are equal if their corresponding components are equal, so $a_1 = b_1$, $a_2 = b_2$, ..., $a_n = b_n$.

Notice that, in order to be equal, two vectors must have the same dimension and the same entries in the same order. Thus, the vectors (1,3) and (1,3,0) are not equal.

We now define addition of two vectors of the same dimension, which is done componentwise.

Definition 4. Let $\vec{a} = (a_1, a_2, ..., a_n)$ and $\vec{b} = (b_1, b_2, ..., b_n)$ be vectors in \mathbb{R}^n . We define $\vec{a} + \vec{b}$ to be the vector in \mathbb{R}^n given by

$$\vec{a} + \vec{b} = (a_1 + b_1, a_2 + b_2, ..., a_n + b_n).$$

Note that we can only add two vectors if they have the same dimension.

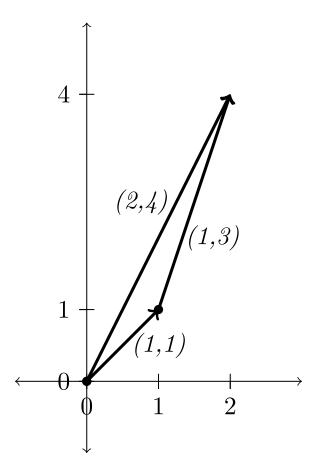
Example 3. Adding the vectors $\vec{a} = (1, -8, 2)$ and $\vec{b} = (3, -1, -2)$, we obtain

$$\vec{a} + \vec{b} = (1, -8, 2) + (3, -1, -2)$$

= $(1 + 3, -8 - 1, 2 - 2)$
= $(4, -9, 0)$.

Geometrically, we can add vectors by placing the start point of the second vector at the end point of the first vector, and drawing an arrow from the start point of the first vector to the end point of the second vector.

Example 4. In this example, we add the vectors (1,1) and (1,3). Adding these vectors algebraically, we obtain (2,4). We can also see this geometrically by placing the start point of the vector (1,3) at the end of the vector (1,1) (so at the point (1,1)), and drawing the vector from the origin to the end point of the vector (1,3), which is now at (2,4).



Another vector operation is scalar multiplication. Here, we multiply a vector by a real number, possibly changing the length of the vector.

Definition 5. Let $\vec{a} = (a_1, ..., a_n)$ be a vector in \mathbb{R}^n , and let r be a real number (also called a scalar). We define the scalar product $r\vec{a}$ to be

$$r\vec{a} = (ra_1, ..., ra_n).$$

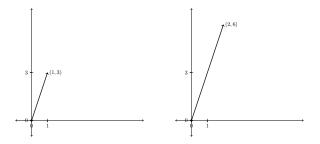
Thus, we see that scalar multiplication is defined by multiplying each component of the vector by the scalar r.

Example 5.
$$3(1,5,-2) = (3,15,-6)$$

 $-1(1,1,1) = (-1,-1,-1)$
 $0(6,2,4) = (0,0,0)$

Now, let's look at what scalar multiplication does geometrically. Consider the vector (1,3).

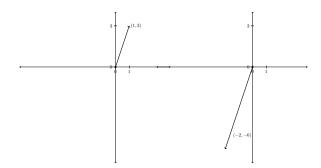
When we multiply (1,3) by 2, we obtain (2,6), which is twice as long as (1,3) and goes in the same direction.



When we multiply (1,3) by $\frac{1}{2}$, we obtain $\left(\frac{1}{2},\frac{3}{2}\right)$, which is half as long as (1,3) and goes in the same direction.



If we multiply (1,3) by -2, we obtain (-2,-6), which is twice as long as (1,3) and goes in the exact opposite direction.



Thus, we have seen that multiplying a vector by a scalar changes the length of a vector, but not the direction (except for reversing it, if the scalar is negative).

Properties

Now, let's recall some useful properties of vector addition and scalar multiplication.

Proposition 1. Suppose $\vec{a}, \vec{b}, \vec{c}$ are vectors in \mathbb{R}^n and k, l are real numbers.

- (a) $\vec{a} + \vec{b} = \vec{b} + \vec{a}$ (vector addition is commutative);
- (b) $\vec{a} + (\vec{b} + \vec{c}) = (\vec{a} + \vec{b}) + \vec{c}$ (vector addition is associative);
- (c) $\vec{a} + \vec{0} = \vec{0} + \vec{a} = \vec{a}$, where $\vec{0} = (0, ..., 0)$ is the zero vector in \mathbb{R}^n ;
- (d) $(k+l)\vec{a} = k\vec{a} + l\vec{a}$;
- (e) $k(\vec{a} + \vec{b}) = k\vec{a} + k\vec{b}$ (with the previous property, scalar multiplication is distributive):
- (f) $k(l\vec{a}) = (kl)\vec{a} = l(k\vec{a});$
- (g) $1\vec{a} = \vec{a}$.

These properties tell us different kinds of algebraic manipulations that we can do with vectors.

Standard Basis Vectors

It's often useful to write things in terms of the standard basis vectors for \mathbb{R}^n .

Definition 6. The vectors $\vec{e}_1 = (1, 0, ...0)$, $\vec{e}_2 = (0, 1, 0, ..., 0)$, ..., $\vec{e}_n = (0, ..., 0, 1)$ in \mathbb{R}^n are called the standard basis vectors for \mathbb{R}^n .

Note that any vector in \mathbb{R}^n can be written uniquely as a linear combination of the standard unit vectors. For example, in \mathbb{R}^4 ,

$$\begin{split} (1,5,-3,6) &= 1(1,0,0,0) + 5(0,1,0,0) - 3(0,0,1,0) + 6(0,0,0,1) \\ &= 1\vec{e}_1 + 5\vec{e}_2 - 3\vec{e}_3 + 6\vec{e}_4. \end{split}$$

In \mathbb{R}^2 , we sometimes write the standard basis vectors as $\mathbf{i} = (1,0)$ and $\mathbf{j} = (0,1)$. This gives us a new notation for vectors, for example we could write

$$(3,4) = 3\mathbf{i} + 4\mathbf{j}.$$

Similarly, in \mathbb{R}^3 , we sometimes write the standard basis vectors as $\mathbf{i} = (1, 0, 0)$, $\mathbf{j} = (0, 1, 0)$, and $\mathbf{k} = (0, 0, 1)$. We can then write

$$(2,3,4) = 2\mathbf{i} + 3\mathbf{j} + 4\mathbf{k}.$$

Summary

In this section, we reviewed some basics about vectors, including the definition of a vector, basic vector operations, standard basis vectors, notation, and the geometric perspective.

The Dot Product

In this section we review the dot product on vectors. This also includes the angle between vectors and the projection of one vector onto another.

The Dot Product

We begin with the definition of the dot product.

Definition 7. The dot product of two vectors $\vec{v} = (v_1, v_2, ..., v_n)$ and $\vec{w} = (w_1, w_2, ..., w_n)$ in \mathbb{R}^n is

$$\vec{v} \cdot \vec{w} = v_1 w_1 + v_2 w_2 + \dots + v_n w_n.$$

Notice that the dot product takes two vectors and outputs a scalar.

Example 6.
$$(1,6) \cdot (-3,-6) = -3 - 36 = -39$$

 $(1,2,3) \cdot (7,-2,4) = 7 - 4 + 12 = 15$
 $(1,7,-3) \cdot (3,0,1) = 3 + 0 - 3 = 0$

We can also compute the dot product using the magnitude (or length) of the vectors and the angle in between them.

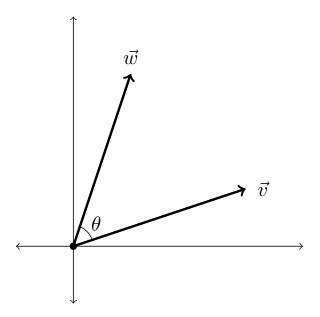
Proposition 2. If \vec{v} and \vec{w} are vectors in \mathbb{R}^n , then

$$\vec{v} \cdot \vec{w} = \|\vec{v}\| \|\vec{w}\| \cos \theta,$$

where $\|\vec{v}\|$ and $\|\vec{w}\|$ are the lengths of the vectors \vec{v} and \vec{w} , respectively, and θ is the angle between \vec{v} and \vec{w} .

This is illustrated in the picture below.

Learning outcomes: Author(s):



This provides us with a geometric interpretation of the dot product: it gives us a measure of "how much" in the same direction two vectors are (taking their lengths into account). This also gives us a useful way to compute the angle between two vectors.

Example 7. Consider the vectors (1,4) and (-2,2). We have

$$(1,4) \cdot (-2,2) = -2 + 8 = 6,$$

$$\|(1,4)\| = \sqrt{1^2 + 4^2} = \sqrt{17},$$

$$\|(-2,2)\| = \sqrt{(-2)^2 + 2^2} = \sqrt{8}.$$

From $\vec{v} \cdot \vec{w} = ||\vec{v}|| \, ||\vec{w}|| \cos \theta$, we then have

$$6 = \sqrt{17}\sqrt{8}\cos\theta.$$

Solving for θ , we obtain the angle between the vectors as

$$\theta = \arccos\left(\frac{6}{\sqrt{17}\sqrt{8}}\right) \approx 59.04^{\circ}$$

Furthermore, note that for nonzero vectors \vec{v} and \vec{w} in \mathbb{R}^n , their dot product is 0 if and only if $\cos(\theta) = 0$. This means that θ would have to be 90° or 270°, meaning that \vec{v} and \vec{w} are perpendicular.

Proposition 3. Two nonzero vectors \vec{v} in \vec{w} in \mathbb{R}^n are perpendicular if and only if $\vec{v} \cdot \vec{w} = 0$.

This provides us with a very useful algebraic method for determining if two vectors are perpendicular.

Example 8. The vectors (1,7,-3) and (3,0,1) in \mathbb{R}^3 are perpendicular, since

$$(1,7,-3) \cdot (3,0,1) = 3 + 0 - 3 = 0.$$

By taking the dot product of a vector with itself, we get an important relationship between the dot product and the length of a vector.

Proposition 4. Let \vec{v} be a vector in \mathbb{R}^n . Then

$$\vec{v} \cdot \vec{v} = ||\vec{v}||^2.$$

This can be shown directly, or using the fact that the angle between \vec{v} and itself is 0.

Projection of one vector onto another

We can also use the dot product to define the projection of one vector onto another.

Definition 8. For vectors \vec{a} and \vec{b} in \mathbb{R}^n , we define the vector projection of \vec{a} onto \vec{b} as

$$proj_{\vec{b}}(\vec{a}) = \frac{\vec{a} \cdot \vec{b}}{\|\vec{b}\|^2} \vec{b} = \frac{\vec{a} \cdot \vec{b}}{\vec{b} \cdot \vec{b}} \vec{b}.$$

Example 9. We can use this to find the projection of (2,4,3) onto (1,-1,1).

$$proj_{(1,-1,1)}(2,4,3) = \frac{(2,4,3) \cdot (1,-1,1)}{(1,-1,1) \cdot (1,-1,1)} (1,-1,1)$$

$$= \frac{2-4+3}{1+1+1} (1,-1,1)$$

$$= \frac{1}{3} (1,-1,1)$$

$$= \left(\frac{1}{3}, -\frac{1}{3}, \frac{1}{3}\right)$$

Summary

In this section we reviewed the dot product on vectors, the angle between vectors, and the projection of one vector onto another.

The Cross Product

In this section, we review the vector cross product, including the geometric perspective of the cross product, the area of a parallelogram, and the volume of parallelepiped.

The Cross Product

The cross product is fundamentally different from the dot product in a couple of ways. The cross product is defined only on vectors in \mathbb{R}^3 , while the dot product is defined in \mathbb{R}^n for any positive integer n. Furthermore, the cross product takes two vectors and produces another vector, while the dot product takes two vectors and produces a scalar.

We now give the algebraic definition of the cross product.

Definition 9. Let $\vec{a} = (a_1, a_2, a_3)$ and $\vec{b} = (b_1, b_2, b_3)$ be vectors in \mathbb{R}^3 . The cross product of \vec{a} and \vec{b} , denoted $\vec{a} \times \vec{b}$, is defined to be

$$\vec{a} \times \vec{b} = \left(\begin{array}{c} a_2b_3 - a_3b_2 \\ a_3b_1 - a_1b_3 \\ a_1b_2 - a_2b_1 \end{array} \right).$$

Equivalently, we can compute the cross product as

$$\vec{a} imes \vec{b} = det \left(egin{array}{ccc} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{array}
ight),$$

where

$$\mathbf{i} = (1, 0, 0),$$

 $\mathbf{j} = (0, 1, 0),$
 $\mathbf{k} = (0, 0, 1).$

Learning outcomes: Author(s):

Example 10.

$$\begin{aligned} (3,2,-1)\times(9,0,2) &= \det\begin{pmatrix} \pmb{i} & \pmb{j} & \pmb{k} \\ 3 & 2 & -1 \\ 9 & 0 & 2 \end{pmatrix} \\ &= (2\cdot2)\mathbf{i} - (0\cdot-1)\mathbf{i} + (-1\cdot9)\mathbf{j} - (2\cdot3)\mathbf{j} + (3\cdot0)\mathbf{k} - (9\cdot-1)\mathbf{k} \\ &= 4\mathbf{i} - 15\mathbf{j} + 9\mathbf{k} \\ &= (4,-15,9) \end{aligned}$$

The cross product has some nice algebraic properties, which can be very useful.

Proposition 5. Let \vec{a} , \vec{b} , and \vec{c} be vectors in \mathbb{R}^3 , and let $k \in \mathbb{R}$ be a scalar. The cross product has the following properties:

- (a) $\vec{a} \times \vec{b} = -\vec{b} \times \vec{a}$ (the cross product is anticommutative);
- (b) $\vec{a} \times (\vec{b} + \vec{c}) = \vec{a} \times \vec{b} + \vec{a} \times \vec{c};$
- (c) $(\vec{a} + \vec{b}) \times \vec{c} = \vec{a} \times \vec{c} + \vec{b} \times \vec{c}$ (with the previous property, the cross product is distributive over vector addition);
- (d) $k(\vec{a} \times \vec{b}) = (k\vec{a}) \times \vec{b} = \vec{a} \times (k\vec{b}).$

In particular, it's important to remember that the cross product is *not* commutative, so the order of the vectors matters!

Geometry of the Cross Product

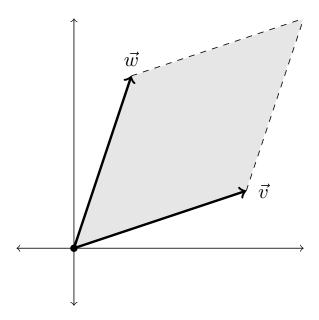
It's often easiest to compute cross products algebraically, but it's easier to understand their significance from a geometric perspective. We now discuss some of the geometric properties of the cross product.

Proposition 6. Let \vec{a} and \vec{b} be vectors in \mathbb{R}^3 , and consider their cross product $\vec{a} \times \vec{b}$.

• The magnitude of the vector $\vec{a} \times \vec{b}$ can be computed as

$$\|\vec{a} \times \vec{b}\| = \|\vec{a}\| \|\vec{b}\| \sin(\theta),$$

where θ is the angle between the vectors \vec{a} and \vec{b} . Furthermore, this magnitude is equal to the area of the parallelogram determined by \vec{a} and \vec{b} .



• The vector $\vec{a} \times \vec{b}$ is always perpendicular to both \vec{a} and \vec{b} , and follows the right-hand rule. That is, if you take your right hand and orient it so you can curl your fingers from the vector \vec{a} to the \vec{b} , your thumb will be pointing in the same direction as the cross product $\vec{a} \times \vec{b}$.

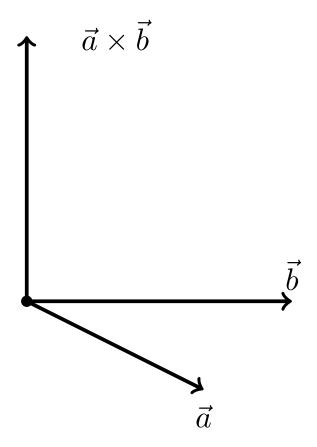
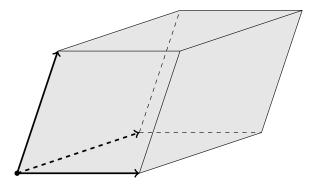


Image this image in \mathbb{R}^3 , so that $\vec{a} \times \vec{b}$ is perpendicular to both \vec{a} and \vec{b} .

Volume of a Parallelepiped

We can use the cross product and dot product together to compute the volume of a parallelepiped.



The volume of the parellelepiped can be computed as the area of the base times the height. We've seen that the area of the base can be computed as the magnitude of a cross product, $\|\vec{a} \times \vec{b}\|$. The height of the parallelepiped can be computed as $\|\vec{c}\| |\cos(\theta)|$, where θ is the angle between the vector \vec{c} and a line perpendicular to the base. We then have that the volume is $\|\vec{a} \times \vec{b}\| \|\vec{c}\| |\cos(\theta)|$, which we can recognize as the absolute value of the dot product of the vectors $\vec{a} \times \vec{b}$ and \vec{c} . Thus we have the following proposition.

Proposition 7. The volume of the parallelepiped determined by the vectors \vec{a} , \vec{b} , and \vec{c} can be computed as $|(\vec{a} \times \vec{b}) \cdot \vec{c}|$.

Summary

We've reviewed the cross product, including its properties and geometric perspective, including its use in finding the area of parallelograms and volume of parallelepipeds.

Matrices

In this section, we review matrices, including the determinant and the linear transformation represented by a matrix.

Matrices

We begin with the definition of a matrix.

Definition 10. An $m \times n$ matrix A is a rectangular array of numbers a_{ij} , with m rows and n columns:

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix},$$

where the a_{ij} are real numbers for i and j integers with $1 \le i \le m$ and $1 \le j \le n$. The numbers a_{ij} are called the entries of the matrix A.

Note that for an entry a_{ij} , the subscript ij describes the location of a_{ij} in the matrix A: i gives the row, and j gives the column.

We can also think of a matrix as a "vector of vectors" in two different ways. If we imagine that the columns of A are vectors in \mathbb{R}^n , then the matrix of A can be viewed as a vector of column vectors. If we imagine that the rows of A are vectors in \mathbb{R}^n , then the matrix A can be viewed as a vector of row vectors.

Matrix Operations

Here, we'll define matrix addition and matrix multiplication.

In order to be able to add two matrices, they need to have the exact same dimensions. That is, they both need to be $m \times n$ matrices for some fixed values of m and n. When we have two matrices with the same dimensions, we define their sum component-wise or entry-wise.

Learning outcomes: Author(s):

Definition 11. Let A and B be two $m \times n$ matrices, with

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix},$$

$$B = \begin{pmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & & \vdots \\ b_{m1} & b_{m2} & \cdots & b_{mn} \end{pmatrix}.$$

Then we define the matrix sum A + B to be

$$A+B=\begin{pmatrix} a_{11}+b_{11} & a_{12}+b_{12} & \cdots & a_{1n}+b_{1n} \\ a_{21}+b_{21} & a_{22}+b_{22} & \cdots & a_{2n}+b_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1}+b_{m1} & a_{m2}+b_{m2} & \cdots & a_{mn}+b_{mn}. \end{pmatrix},$$

That is, A + B is the $m \times n$ matrix obtained by adding the corresponding entries of A and B.

Example 11. We can add the matrices $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$ and $B = \begin{pmatrix} 7 & 8 & 9 \\ 10 & 11 & 12 \end{pmatrix}$ as follows:

$$A + B = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} + \begin{pmatrix} 7 & 8 & 9 \\ 10 & 11 & 12 \end{pmatrix}$$
$$= \begin{pmatrix} 1 + 7 & 2 + 8 & 3 + 9 \\ 4 + 10 & 5 + 11 & 6 + 12 \end{pmatrix}$$
$$= \begin{pmatrix} 8 & 10 & 12 \\ 14 & 16 & 18 \end{pmatrix}$$

Example 12. We cannot add the matrices $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$ and $B = \begin{pmatrix} 7 & 8 \\ 10 & 11 \end{pmatrix}$, because their dimensions don't match.

As you might expect, matrix addition has some nice properties which are inherited from addition of real numbers. We list some of them here.

Proposition 8. Let A, B, and C be $m \times n$ matrices. Then we have

- (a) A + B = B + A (matrix addition is commutative);
- (b) A + (B + C) = (A + B) + C (matrix addition is associative).

Furthermore, there is an $m \times n$ matrix O, called the zero matrix, such that A + O = A for any $m \times n$ matrix A. All of the entries of the zero matrix are the real number 0.

We've seen that matrix addition works in a very natural way, and multiplying a matrix by a scalar (or real number) is similarly nice. We now define scalar multiplication for matrices.

Definition 12. Let

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}$$

be an $m \times n$ matrix, and let $k \in \mathbb{R}$ be a scalar. Then the scalar product of k and A, denoted kA, is

$$kA = \begin{pmatrix} ka_{11} & ka_{12} & \cdots & ka_{1n} \\ ka_{21} & ka_{22} & \cdots & ka_{2n} \\ \vdots & \vdots & & \vdots \\ ka_{m1} & ka_{m2} & \cdots & ka_{mn} \end{pmatrix}.$$

That is, we obtain the scalar product by multiplying each entry in A by the scalar k.

Example 13. We can compute the scalar product of 2 and the matrix $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$ as follows:

$$2A = 2 \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$$
$$= \begin{pmatrix} 2 \cdot 1 & 2 \cdot 2 & 2 \cdot 3 \\ 2 \cdot 4 & 2 \cdot 5 & 2 \cdot 6 \end{pmatrix}$$
$$= \begin{pmatrix} 2 & 4 & 6 \\ 8 & 10 & 12 \end{pmatrix}.$$

We now list some nice properties of scalar multiplication.

Proposition 9. Let A and B be $m \times n$ matrices, and let k and l be scalars in \mathbb{R} . Then

- (a) (k+l)A = kA + lA (scalar multiplication is distributive over scalar addition);
- (b) k(A+B) = kA + kB (scalar multiplication is distributive over matrix addition);
- (c) k(lA) = (kl)A = l(kA).

We'll now define matrix multiplication, which can be a bit trickier to work with than matrix addition or scalar multiplication. Here are some important things to remember about matrix multiplication:

- Not all matrices can be multiplied. In order to compute the product AB of two matrices A and B, the number of columns in A needs to be the same as the number of rows in B.
- Matrix multiplication is *not* commutative. In fact, its possible that the matrix product AB exists but the product BA does not.

Definition 13. Let A be an $m \times n$ matrix, and let B be an $n \times p$ matrix, with

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix},$$

$$B = \begin{pmatrix} b_{11} & b_{12} & \cdots & b_{1p} \\ b_{21} & b_{22} & \cdots & b_{2p} \\ \vdots & \vdots & & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{np} \end{pmatrix}.$$

Note that we are assuming the number of columns in A is the same as the number of rows in B.

We define the matrix product of A and B, denoted AB, to be the $m \times p$ matrix

$$AB = \begin{pmatrix} a_{11}b_{11} + a_{12}b_{21} + \dots + a_{1n}b_{n1} & a_{11}b_{12} + a_{12}b_{22} + \dots + a_{1n}b_{n2} & \dots & a_{11}b_{1p} + a_{12}b_{2p} + \dots + a_{1n}b_{n2} \\ a_{21}b_{11} + a_{22}b_{21} + \dots + a_{2n}b_{n1} & a_{21}b_{12} + a_{22}b_{22} + \dots + a_{2n}b_{n2} & \dots & a_{21}b_{1p} + a_{22}b_{2p} + \dots + a_{2n}b_{n2} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{m1}b_{11} + a_{m2}b_{21} + \dots + a_{mn}b_{n1} & a_{m1}b_{12} + a_{m2}b_{22} + \dots + a_{mn}b_{n2} & \dots & a_{m1}b_{1p} + a_{m2}b_{2p} + \dots + a_{mn}b_{n2} \end{pmatrix}$$

Equivalently, we could define the ijth entry of AB to be the dot product of the ith row of A with the jth column of B. This makes sense, since the number of columns in A is the same as the number of rows in B (both n), which ensures that the ith row of A and the jth column of B are both vectors in \mathbb{R}^n .

This definition can seem a bit convoluted, and it's easier to understand how matrix multiplication works by going through an example.

Example 14. We can compute the product AB of the matrices $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$

$$and \ B = \left(\begin{array}{ccc} 7 & 8 \\ 9 & 10 \\ 11 & 12 \end{array} \right) \ as \ follows:$$

$$AB = \left(\begin{array}{ccc} 1 & 2 & 3 \\ 4 & 5 & 6 \end{array} \right) \left(\begin{array}{ccc} 7 & 8 \\ 9 & 10 \\ 11 & 12 \end{array} \right),$$

$$= \left(\begin{array}{ccc} 1 \cdot 7 + 2 \cdot 9 + 3 \cdot 11 & 1 \cdot 8 + 2 \cdot 10 + 3 \cdot 12 \\ 4 \cdot 7 + 5 \cdot 9 + 6 \cdot 11 & 4 \cdot 8 + 5 \cdot 10 + 6 \cdot 12 \end{array} \right),$$

$$= \left(\begin{array}{ccc} 58 & 64 \\ 139 & 154 \end{array} \right).$$

Note that we multiplied a 2×3 matrix by a 3×2 matrix, and we obtained a 2×2 matrix.

We can also compute the product BA for the same matrices as above.

$$BA = \begin{pmatrix} 7 & 8 \\ 9 & 10 \\ 11 & 12 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix},$$

$$= \begin{pmatrix} 7 \cdot 1 + 8 \cdot 4 & 7 \cdot 2 + 8 \cdot 5 & 7 \cdot 3 + 8 \cdot 6 \\ 9 \cdot 1 + 10 \cdot 4 & 9 \cdot 2 + 10 \cdot 5 & 9 \cdot 3 + 10 \cdot 6 \\ 11 \cdot 1 + 12 \cdot 4 & 11 \cdot 2 + 12 \cdot 5 & 11 \cdot 3 + 12 \cdot 6 \end{pmatrix},$$

$$= \begin{pmatrix} 39 & 54 & 69 \\ 49 & 68 & 81 \\ 59 & 82 & 105 \end{pmatrix}.$$

Note that we multiplied a 3×2 matrix by a 2×3 matrix, and we obtained a 3×3 matrix.

Note that in this case $AB \neq BA$; matrix multiplication is not commutative, so the order of the matrices matters!

Although matrix multiplication is not commutative, it still has some nice algebraic properties. We list some of them here.

Proposition 10. Let A, B, and C be matrices of dimensions such that the following operations are defined, and let k be a scalar. Then

- (a) A(BC) = (AB)C (matrix multiplication is associative);
- (b) k(AB) = (kA)B = A(kB);
- (c) A(B+C) = AB + AC:
- (d) (A + B)C = AC + BC (with the previous property, matrix multiplication is distributive over matrix addition).

Determinants

When we have a square matrix (meaning an $n \times n$ matrix, where the number of rows and number of columns are the same), we can compute an important number, called the determinant of the matrix. It turns out that this single number can tell us some important things about the matrix!

We begin by defining the determinant of a 2×2 matrix.

Definition 14. Consider the 2×2 matrix $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$. We define the determinant of the matrix A to be

$$det(A) = ad - bc.$$

We also sometimes use the notation |A| for the determinant of the matrix A.

Note that the determinate of a 2×2 matrix is just a number, not a matrix. We compute the determinant in a couple of examples.

Example 15. We'll compute the determinant of the matrix $A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$.

$$det(A) = 1 \cdot 4 - 2 \cdot 3$$
$$= -2.$$

We've defined the determinant of 2×2 matrices, but we haven't defined the determinant of a larger square matrix yet. It turns out that the determinant is defined *inductively*. This means that the determinant of a 3×3 matrix is defined using determinants of 2×2 matrices, the determinant of a 4×4 matrix is defined using determinants of 3×3 matrices, the determinant of a 5×5 matrix is defined using determinants of 4×4 matrices, and so on. This means in order to compute the determinant of a large square matrix, we often need to compute the determinants of many smaller matrices.

We now give the definition of the determinant of an $n \times n$ matrix.

Definition 15. Let A be an $n \times n$ matrix, with entries a_{ij} . We defined the determinant of A to be the number computed by

$$det(A) = (-1)^{1+1}a_{11}det(A_{11}) + (-1)^{1+2}a_{12}det(A_{12}) + \dots + (-1)^{1+n}a_{1n}det(A_{1n}),$$

where A_{ij} is the $(n-1) \times (n-1)$ submatrix of A which we obtain by deleting the *i*th row and *j*the column from A.

This definition is pretty confusing if you read through it without seeing an example, but this actually follows a nice pattern. This pattern is easier to see with an example.

Example 16. We compute the determinant of the 4×4 matrix,

$$A = \left(\begin{array}{rrrr} 1 & 4 & 2 & -1 \\ 0 & 0 & -2 & 1 \\ -3 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{array}\right).$$

Note that we begin by writing this in terms of determinants of 3×3 matrices. But in order to compute the determinant of each 3×3 matrix, we write it in terms of 2×2 matrices! This winds up being a lot of determinants to compute.

$$det(A) = (-1)^{1+1} 1 det \begin{pmatrix} 0 & -2 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} + (-1)^{1+2} 4 det \begin{pmatrix} 0 & -2 & 1 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} + (-1)^{1+3} 2 det \begin{pmatrix} 0 & 0 & 1 \\ -3 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} + (-1)^{1+4} (-1) det \begin{pmatrix} 0 & 0 & -2 \\ -3 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

We now compute the determinant of each of the 3×3 submatrices, which we

compute using determinants of 2×2 matrices.

$$det\begin{pmatrix} 0 & -2 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = (-1)^{1+1}0det\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + (-1)^{1+2}(-2)det\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \\ + (-1)^{1+3}1det\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \\ = 1 \cdot 0 \cdot (1 \cdot 1 - 0 \cdot 0) + -1 \cdot (-2) \cdot (0 \cdot 1 - 0 \cdot 0) + 1 \cdot 1 \cdot (0 \cdot 0 - 1 \cdot 0) \\ = 0$$

$$det\begin{pmatrix} 0 & -2 & 1 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = (-1)^{1+1}0det\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + (-1)^{1+2}(-2)det\begin{pmatrix} -3 & 0 \\ 0 & 1 \end{pmatrix} \\ + (-1)^{1+3}1det\begin{pmatrix} -3 & 1 \\ 0 & 0 \end{pmatrix} \\ = 1 \cdot 0 \cdot (1 \cdot 1 - 0 \cdot 0) + -1 \cdot (-2) \cdot (-3 \cdot 1 - 0 \cdot 0) + 1 \cdot 1 \cdot (-3 \cdot 0 - 1 \cdot 0) \\ = 6$$

$$det\begin{pmatrix} 0 & 0 & 1 \\ -3 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} = (-1)^{1+1}0det\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} + (-1)^{1+2}0det\begin{pmatrix} -3 & 0 \\ 0 & 1 \end{pmatrix} \\ + (-1)^{1+3}1det\begin{pmatrix} -3 & 0 \\ 0 & 0 \end{pmatrix} \\ = 1 \cdot 0 \cdot (0 \cdot 1 - 0 \cdot 0) + -1 \cdot 0 \cdot (-3 \cdot 1 - 0 \cdot 0) + 1 \cdot 1 \cdot (-3 \cdot 0 - 0 \cdot 0) \\ = 0$$

$$det\begin{pmatrix} 0 & 0 & -2 \\ -3 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} = (-1)^{1+1}0det\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + (-1)^{1+2}0det\begin{pmatrix} -3 & 1 \\ 0 & 0 \end{pmatrix} \\ + (-1)^{1+3}(-2)det\begin{pmatrix} -3 & 0 \\ 0 & 0 \end{pmatrix} \\ = 1 \cdot 0 \cdot (0 \cdot 0 - 1 \cdot 0) + -1 \cdot 0 \cdot (-3 \cdot 0 - 1 \cdot 0) + 1 \cdot (-2) \cdot (-3 \cdot 0 - 0 \cdot 0) \\ = 0$$

Substituting these in to our computation of the determinant of A, we then obtain

$$det(A) = 1 \cdot 1 \cdot 0 + (-1) \cdot 4 \cdot (6) + 1 \cdot 2 \cdot 0 + (-1) \cdot (-1) \cdot 0$$

= -24.

We sometimes call this method of computing a determinant as "expanding along the first row." This is because we can also compute the determinant of a matrix by similarly expanding along a different row, or even a column.

Proposition 11. We can similarly compute the determinant of an $n \times n$ matrix

A by expanding along any row or column. Expanding along the ith row, we have

$$det(A) = (-1)^{i+1}a_{i1}det(A_{i1}) + (-1)^{i+2}a_{i2}det(A_{i2}) + \dots + (-1)^{i+n}a_{in}det(A_{in}).$$

Expanding along the jth column, we have

$$det(A) = (-1)^{1+j} a_{1j} det(A_{1j}) + (-1)^{2+j} a_{2j} det(A_{2j}) + \dots + (-1)^{n+j} a_{nj} det(A_{nj}).$$

Once again, A_{ij} is the $(n-1) \times (n-1)$ submatrix obtained from A by deleting the *i*th row and *j*th column.

It can be useful to think about which row or column will be easiest to expand along. In particular, choosing a row or column with a lot of zeros greatly simplifies computation.

Example 17. We'll once again compute the determinant of the 4×4 matrix

$$A = \left(\begin{array}{cccc} 1 & 4 & 2 & -1 \\ 0 & 0 & -2 & 1 \\ -3 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{array}\right),$$

this time by expanding along the second column. Note that this column is a good choice, since there's only one nonzero element. We have

$$det(A) = (-1)^{1+2}(4)det \begin{pmatrix} 0 & -2 & 1 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} + (-1)^{2+2}(0)det \begin{pmatrix} 1 & 2 & -1 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} + (-1)^{3+2}(0)det \begin{pmatrix} 1 & 2 & -1 \\ 0 & -2 & 1 \\ 0 & 0 & 1 \end{pmatrix}$$

We'll only compute the determinant of the submatrix $\begin{pmatrix} 0 & -2 & 1 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$; we

won't bother computing the others since their determinants will be multiplied by 0.

$$det \begin{pmatrix} 0 & -2 & 1 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = (-1)^{3+1}(0)det \begin{pmatrix} -2 & 1 \\ 1 & 0 \end{pmatrix} + (-1)^{3+2}(0)det \begin{pmatrix} 0 & 1 \\ -3 & 0 \end{pmatrix} + (-1)^{3+3}(1)det \begin{pmatrix} 0 & -2 \\ -3 & 1 \end{pmatrix} = 0 + 0 + (1)(1)(0 \cdot 1 - (-2) \cdot (-3)),$$

$$= -6.$$

Once again, we don't bother computing the determinants which will be multiplied by zero. Note that we chose to expand across the last row, since it had two zeroes. Expanding along the first column would also have been a reasonable choice.

Returning to our computation of the determinant of A, we have

$$\begin{aligned} \det(A) &= (-1)^{1+2}(4) \det \begin{pmatrix} 0 & -2 & 1 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} + (-1)^{2+2}(0) \det \begin{pmatrix} 1 & 2 & -1 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} + (-1)^{3+2}(0) \det \begin{pmatrix} 1 & 2 & -1 \\ 0 & -2 & 1 \\ 0 & 0 & 1 \end{pmatrix} \\ &= (-1)(4)(-6) + 0 + 0 + 0, \\ &= 24. \end{aligned}$$

Notice that this matching our previous computation, expanding along the first row.

One of the most powerful uses of the determinant is to tell us whether or not a matrix is invertible. Recall that an $n \times n$ matrix A is *invertible* if there is a matrix B such that $AB = BA = I_n$, where I_n is the $n \times n$ identity matrix.

Proposition 12. An $n \times n$ matrix A is invertible if and only if its determinant is nonzero.

This gives us a convenient way to test if a matrix is invertible, without needing to produce an explicit inverse.

Example 18. We found that the determinant of the matrix

$$A = \left(\begin{array}{cccc} 1 & 4 & 2 & -1 \\ 0 & 0 & -2 & 1 \\ -3 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{array}\right)$$

is 24. Since this is nonzero, the matrix A is invertible.

On the other hand, you can verify that the determinant of the matrix

$$B = \left(\begin{array}{rrrr} 1 & 2 & 3 & 4 \\ 1 & 0 & -1 & -2 \\ -3 & 1 & -1 & 0 \\ -1 & 3 & 1 & 2 \end{array}\right)$$

is 0. Thus, the matrix B is not invertible.

Linear Transformations

One of the most important uses of matrices is to represent linear transformations. Recall the definition of a linear transformation.

Definition 16. A function T from \mathbb{R}^n to \mathbb{R}^n is a linear transformation if for all vectors \vec{v} and \vec{w} in \mathbb{R}^n and all scalars $k \in \mathbb{R}$, we have

(a)
$$T(\vec{v} + \vec{w}) = T(\vec{v}) + T(\vec{w});$$

(b)
$$T(k\vec{v}) = kT(\vec{v})$$
.

We can view an $m \times n$ matrix A as representing a linear transformation from \mathbb{R}^n to \mathbb{R}^m as follows. We write vectors as column vectors, or, equivalently, $n \times 1$ or $m \times 1$ matrices. For an input column vector \vec{v} in \mathbb{R}^n , we multiply \vec{v} by A

on the left, using matrix multiplication. This produces an $m \times 1$ matrix, or, equivalently, a column vector in \mathbb{R}^m . Thus, we can define a function

$$T_A(\vec{v}) = A\vec{v}.$$

Using properties of matrix multiplication, we have that this is a linear transformation. Thus, we have the linear transformation associated to a matrix.

Example 19. Consider the linear transformation T_A from \mathbb{R}^3 to \mathbb{R}^2 corresponding to the 2×3 matrix

$$A = \left(\begin{array}{ccc} 1 & -5 & 3 \\ 2 & 0 & -1 \end{array}\right).$$

Let investigate the images of several vectors in \mathbb{R}^3 under the linear transformation T_A .

$$T_A((1,2,3)) = \begin{pmatrix} 1 & -5 & 3 \\ 2 & 0 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$
$$= \begin{pmatrix} 1 \cdot 1 + -5 \cdot 2 + 3 \cdot 3 \\ 2 \cdot 1 + 0 \cdot 2 + -1 \cdot 3 \end{pmatrix}$$
$$= \begin{pmatrix} 0 \\ -1 \end{pmatrix}$$

$$T_A((1,-1,2)) = \begin{pmatrix} 1 & -5 & 3 \\ 2 & 0 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$
$$= \begin{pmatrix} 1 \cdot 1 + -5 \cdot -1 + 3 \cdot 1 \\ 2 \cdot 1 + 0 \cdot -1 + -1 \cdot 1 \end{pmatrix}$$
$$= \begin{pmatrix} 9 \\ 1 \end{pmatrix}$$

$$T_A((1,0,0)) = \begin{pmatrix} 1 & -5 & 3 \\ 2 & 0 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$
$$= \begin{pmatrix} 1 \cdot 1 + -5 \cdot 0 + 3 \cdot 0 \\ 2 \cdot 1 + 0 \cdot 0 + -1 \cdot 0 \end{pmatrix}$$
$$= \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

Matrices

$$T_A((0,1,0)) = \begin{pmatrix} 1 & -5 & 3 \\ 2 & 0 & -1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$
$$= \begin{pmatrix} 1 \cdot 0 + -5 \cdot 1 + 3 \cdot 0 \\ 2 \cdot 0 + 0 \cdot 1 + -1 \cdot 0 \end{pmatrix}$$
$$= \begin{pmatrix} -5 \\ 0 \end{pmatrix}$$

$$T_A((0,0,1)) = \begin{pmatrix} 1 & -5 & 3 \\ 2 & 0 & -1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$
$$= \begin{pmatrix} 1 \cdot 0 + -5 \cdot 0 + 3 \cdot 1 \\ 2 \cdot 0 + 0 \cdot 0 + -1 \cdot 1 \end{pmatrix}$$
$$= \begin{pmatrix} 3 \\ -1 \end{pmatrix}$$

Notice that when we apply the linear transformation to the standard unit vectors \vec{e}_1 , \vec{e}_2 , and \vec{e}_3 , we obtain the columns of A as the output vector. This observation can be used to reconstruct a matrix from a given linear transformation.

Proposition 13. Given any linear transformation T from \mathbb{R}^n to \mathbb{R}^m , there is an $m \times n$ matrix such that $T = T_A$.

Furthermore, the columns of A can be obtained by applying T to the standard unit vectors. More specifically, the jth column of A is given by $T(\vec{e_j})$.

We can see how this is useful through an example.

Example 20. Consider the linear transformation T from \mathbb{R}^2 to \mathbb{R}^2 that rotates a vector by 30° counterclockwise. We can see geometrically that, for the standard unit vectors \vec{e}_1 and \vec{e}_2 in \mathbb{R}^2 , we have

$$T((1,0)) = \left(\frac{\sqrt{3}}{2}, \frac{1}{2}\right),$$

$$T((0,1)) = \left(-\frac{1}{2}, \frac{\sqrt{3}}{2}\right).$$

These tell us the columns of the matrix corresponding to the linear transformation, so we then know that the rotation can be represented by the matrix

$$A = \begin{pmatrix} \frac{\sqrt{3}}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{\sqrt{3}}{2} \end{pmatrix}.$$

Summary

In this section, we reviewed matrix operations and properties, determinants, and linear transformations.

Although we've reviewed some of the most important concepts from linear algebra, there is still a lot of material that we weren't able to include here. Make sure you refer back to your linear algebra textbook if there's anything else you need to review!

Representations of Lines and Planes

In this section, we review the different ways we can represent lines and planes, including parametric representations.

Representations of Lines

When you think of describing a line algebraically, you might think of the standard form

$$y = mx + b,$$

where m is the slope and b is the y-intercept. This is often called slope-intercept form.

In addition to slope-intercept form, there are several other ways to represent lines. For example, you may remember using *point-slope* form in single variable calculus. We can describe a line of slope m going through a point (x_0, y_0) with the equation

$$y - y_0 = m(x - x_0).$$

It's important to note that there are many different possible choices for the point (x_0, y_0) . Because of this, unlike slope-intercept form, point-slope form does not give a unique representation of a line.

In linear algebra, we saw that we could parametrize a line using a vector $\vec{v} = (v_1, v_2)$ giving the direction of the line, and a point (x_0, y_0) that the line passes through. We parametrize the line as

$$\vec{x}(t) = (v_1, v_2)t + (x_0, y_0),$$

= $(v_1t + x_0, v_2t + y_0).$

Note that this representation works a bit differently from the previous two representations. In slope-intercept form and point-slope form, the line was the set of points (x, y) satisfying the given equation. However, in the parametrization, we plug in values for the parameter t in order to get points on the line.

Unlike slope-intercept form and point-slope form, the parametrization of a line can easily be generalized to three or more dimensions. That is, a line in \mathbb{R}^n through the point \vec{a} and in the direction of the vector \vec{v} can be parametrized as

$$\vec{x}(t) = \vec{v}t + \vec{a}$$

Learning outcomes: Author(s):

for $t \in \mathbb{R}$.

If we would like to describe a line in higher dimensions using equations (rather than a parametrization), we would need more than one equation. For example, in \mathbb{R}^3 , we would require two equations to determine a line.

Representations of Planes

We also have multiple ways to represent planes. Here, we'll focus on planes in \mathbb{R}^3 .

Recall that a plane can be determined by two vectors (giving the "direction" of the plane) and a point that the plane passes through. We can use this to give a parametrization for the plane through the point \vec{a} and parallel to the vectors \vec{v} and \vec{w} :

$$\vec{x}(s,t) = \vec{v}s + \vec{w}t + \vec{a},$$

for s and t in \mathbb{R} . Note that we require two parameters for the parametrization of the plane.

We can also describe a plane using a single linear equation in x, y, and z. For example,

$$2x + 4y - z = 9$$

defines a plane. A standard way to do this is using a point on the plane and a normal vector to the plane. Recall that a normal vector is perpendicular to every vector in the plane. If $\vec{n} = (n_1, n_2, n_3)$ is a normal vector to a plane passing through the point $\vec{a} = (a_1, a_2, a_3)$, the plane is defined by the equation

$$\vec{n} \cdot (\vec{x} - \vec{a}) = 0.$$

This can be rewritten as

$$n_1(x-a_1) + n_2(y-a_2) + n_3(z-a_3) = 0.$$

Summary

We reviewed various representations of lines and planes, including parametrizations.

Extra Problems

Online Problems

Problem 1 Compute the following:

$$(1,2,3) + (8,3,6) = \boxed{(9,5,9)}$$

$$4(1,-2,4) = \boxed{(4,-8,16)}$$
$$-12((5,2,6) - (8,2,4)) = \boxed{(36,0,24)}$$

Problem 2 Let h be a constant. Compute the following:

$$(7,2,-1) + (2h,0,h) = (7+2h,2,h-1)$$

$$h(1,8,2) = (h,8h,2h)$$

Problem 3 For each of the following, determine whether the quantity exists or does not exist.

$$(1, 8, 3, 7) + (-1, 7, 2, 7)$$

Multiple Choice:

- (a) Exists. ✓
- (b) Does not exist.

$$(2,8,3)+(1,7)$$

Multiple Choice:

Learning outcomes: Author(s):

- (a) Exists.
- (b) Does not exist. ✓
- (2,7,3)+1

Multiple Choice:

- (a) Exists.
- (b) Does not exist. ✓
- (2,8,3)(1,7,3)

Multiple Choice:

- (a) Exists.
- (b) Does not exist. ✓

2(7,2,3,7,2)

Multiple Choice:

- (a) Exists. ✓
- (b) Does not exist.

Problem 4 For points $P_1 = (2, -3, 7, 1)$ and $P_2 = (-1, 7, 2, 1)$, compute the displacement vector $\vec{P_1P_2}$.

$$\vec{P_1P_2} = (-3, 10, -5, 0)$$

Problem 5 Write the vector $2\mathbf{i} - 5\mathbf{j} + 2\mathbf{k}$ in \mathbb{R}^3 in standard vector notation.

$$2\mathbf{i} - 5\mathbf{j} + 2\mathbf{k} = (2, -5, 2)$$

Problem 6 Compute the dot product.

$$(1,8,3)\cdot(-2,6,0)=\boxed{46}$$

Problem 7 Compute the dot product.

$$(1, -5, 0, 2) \cdot (2, -1, 4, 1) = \boxed{9}$$

Problem 8 Compute the dot product.

$$(1,8,3)\cdot(-3,0,1)=\boxed{0}$$

What can you conclude about the vectors?

Multiple Choice:

- (a) They're perpendicular. ✓
- (b) They aren't perpendicular.

Problem 9 Compute the dot product.

$$(1,2,3)\cdot(-3,-2,-1)=\boxed{-10}$$

What can you conclude about the vectors?

Multiple Choice:

- (a) They're perpendicular.
- (b) They aren't perpendicular. ✓

Problem 10 Compute the dot product.

$$(1,8,3,6)\cdot(3,-3,-1,4)=\boxed{-0}$$

What can you conclude about the vectors?

Multiple Choice:

- (a) They're perpendicular. ✓
- (b) They aren't perpendicular.

Problem 11 For each expression, determine whether it exists or does not exist.

 $(2,8,3)\cdot(1,8,2,4)$

Multiple Choice:

- (a) Exists.
- (b) Does not exist. ✓

 $(2,8,3,1,8) \cdot (1,8,2,4,2)$

Multiple Choice:

- (a) Exists. ✓
- (b) Does not exist.

Problem 12 Compute the angle between the vectors (2,8) and (-8,2) in degrees.

$$\theta = 90^{\circ}$$

Problem 13 Compute the angle between the vectors (2, 1, 3, 1, 1) and (-3, 1, 1, 1, -2) in degrees.

$$\theta = \boxed{104.48}^{\circ}$$

(Give your answer as a positive number to two decimal places.

Problem 14 Suppose you have vectors \vec{v} and \vec{w} such that $||\vec{v}|| = 6$ and $||\vec{w}|| = 2$, and the angle between \vec{v} and \vec{w} is $\frac{\pi}{4}$ radians. Compute the dot product of \vec{v} and \vec{w} .

$$\vec{v} \cdot \vec{w} = \boxed{6\sqrt{2}}$$

Problem 15 Compute the projection of $\vec{v} = (1,7,3)$ onto $\vec{w} = (-3,-2,1)$.

$$proj_{\vec{w}}(\vec{v}) = \boxed{(7/2, 7/3, -7/6)}$$

Problem 16 Compute the projection of the vector $\vec{v} = (1, 2, 3)$ onto the vector $\vec{w} = (-3, 2, -1)$.

$$proj_{\vec{w}}(\vec{v}) = \boxed{(0,0,0)}$$

Problem 16.1 Why does your answer make sense?

Multiple Choice:

- (a) The vectors are parallel.
- (b) The vectors are perpendicular. \checkmark
- (c) They are the same length.

Problem 17 Compute the cross product.

$$(1,7,3) \times (2,8,-1) = (-31,7,-6)$$

Problem 18 Compute the cross product.

$$(-1,7,2) \times (2,8,3) = (5,7,-22)$$

Problem	19	$For \ each$	of the	following,	determine	whether	the	expression	ex-
ists or doe	s not	exist.							

 $(-1,7,3) \times (1,7,2)$

Multiple Choice:

- (a) Exists. ✓
- (b) Does not exist.

 $(-1,7,3,8) \times (1,7,2,0)$

Multiple Choice:

- (a) Exists.
- (b) Does not exist. ✓

 $(2,0,0) \times (1,7,2,0)$

Multiple Choice:

- (a) Exists.
- (b) Does not exist. ✓

 $(0,0,0) \times (0,0,0)$

Multiple Choice:

- (a) Exists. ✓
- (b) Does not exist.

Problem 20 Compute the area of the parallelogram determined by (1,6) and (1,0).

 $Area = \boxed{6}$

Problem 21 Compute the volume of the parallelepiped determined by (1,6,2), (-1,2,0), and (0,3,1).

 $Volume = \boxed{2}$

Problem 22 Suppose \vec{v} and \vec{w} are unit vectors in the xy-plane, and we know that they are perpendicular. What is $\vec{v} \times \vec{w}$?

Multiple Choice:

- (a) (0,0,1)
- (b) (0,0,-1)
- (c) Not enough information. \checkmark

Problem 23 Find a parametrization $\vec{x}(t)$ of the line parallel to the vector (1,6,3) and through the point (1,3,2), such that $\vec{x}(0)=(1,3,2)$. $\vec{x}(t)=\boxed{(t+1,6t+3,3t+2)}$

Problem 24 Find a parametrization $\vec{x}(s,t)$ of the plane containing vectors (1,6,2) and (1,3,2), and passing through the point (1,0,0), such that $\vec{x}(0,0) = (1,0,0)$ and $\vec{x}(1,0) = (1,6,2)$.

$$\vec{x}(s,t) = (s+t+1,6s+3t,2s+2t)$$

Problem 25 Give an equation which describes the plane perpendicular to the vector (1,7,3) and through the point (-2,4,1).

$$0 = (x+2) + 7(y-4) + 3(z-1)$$

Written Problems

Problem 26 For any vector \vec{v} in \mathbb{R}^n , prove that $\vec{v} \cdot \vec{v} = ||\vec{v}||^2$.

Problem 27 Prove that vectors \vec{v} and \vec{w} in \mathbb{R}^n are perpendicular if and only if $\operatorname{proj}_{\vec{w}}(\vec{v})$ is the zero vector.

Problem 28 For any vector \vec{v} in \mathbb{R}^3 , prove that $\vec{v} \times \vec{v}$ is the zero vector.

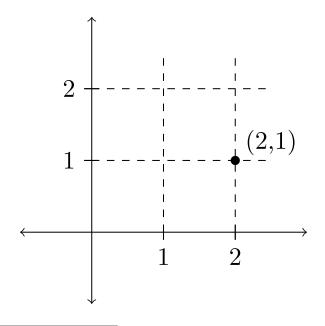
Week 1: Coordinate Systems Review of Coordinate Systems

In this activity, we review coordinate systems that you've seen before, in preparation for introducing new coordinate systems in subsequent sections.

Cartesian Plane

The coordinates that you're probably most comfortable with are standard twodimensional coordinates, also called Cartesian coordinate system on the plane.

In Cartesian coordinates, we describe a point using an x-coordinate and a y-coordinate. We write a point as (x, y), where the x-coordinate describes the horizontal displacement of the point, and the y-coordinate describes the vertical displacement of the point.



Learning outcomes: Author(s):

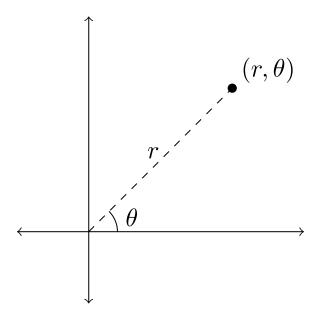
Polar Coordinates

You've also seen polar coordinates.

In polar coordinates, we describe a point with an r-coordinate and a θ -coordinate. The r coordinate gives the distance between the point and the origin, and the θ -coordinate gives the angle (in radians) between the positive x-axis and the segment connecting the origin and the point.

We can switch between cartesian and polar coordinates using the equations

$$x = r\cos\theta$$
$$y = r\sin\theta$$



Problem 29 Write the point $(r, \theta) = (5, \pi/3)$ in cartesian coordinates.

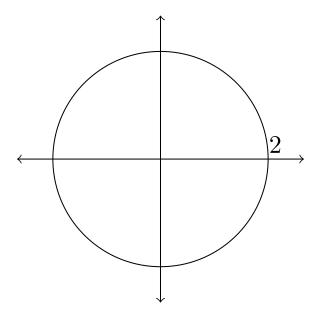
$$(x,y) = (5/2, 5\sqrt{3}/2)$$

Write the point (x, y) = (-2, 2) in polar coordinates.

$$(r,\theta) = \boxed{(\sqrt{8}, 3\pi/4)}$$

Example 21. Recall that we can describe a circle of radius 2 using Cartesian points as the set of points (x, y) satisfying

$$x^2 + y^2 = 4.$$



We would like to write describe this circle using polar coordinates.

By definition, the circle of radius 2 centered at the origin consists of the points which are distance 2 from the origin. Because of this, for any point on the circle, we have

$$r = \boxed{2}$$
.

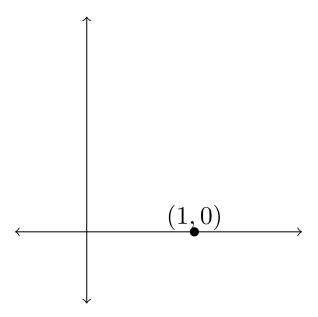
There are points on the circle making every possible angle with the positive x-axis, so we don't need any restrictions on θ . If, however, we only wanted part of the circle, we would accomplish this by restricting θ .

Thus, in polar coordinates, the circle of radius 2 centered at the origin can be described as the set of points (r, θ) such that

$$r=2.$$

There's an important difference between Cartesian coordinates and polar coordinates: Cartesian coordinates are unique, while polar coordinates are not. This means that, given a point P in the plane, there's only one way to describe this point as (x, y) using Cartesian coordinates. However, there are many ways to write the point as (r, θ) , using polar coordinates.

Take, for example, the point (1,0), written in Cartesian coordinates.



This point is on the x-axis and is distance 1 from the origin. Thus, perhaps the most obvious way to represent this point in polar coordinates is as $(r, \theta) = (1,0)$ (coincidentally, the same as in Cartesian coordinates). But we could also describe the angle as 2π , 4π , -2π , etc. So, we could also write the point in polar coordinates as $(r, \theta) = (1, 2\pi)$, and so on.

Perhaps more surprisingly, we can describe this point as $(-1, \pi)$. Imagine making an angle of π with the positive x-axis (so we're on the negative x-axis), then going backwards past the origin. This also gets you to our point. Using equivalent angles, we can also represent the point as $(-1, 3\pi)$, $(-1, -\pi)$, and so on.

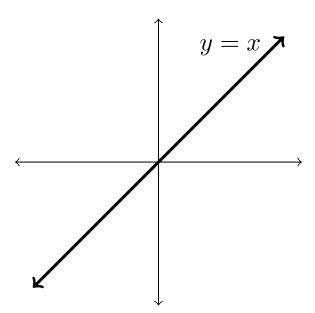
There are some times in working with polar coordinates when we would like to be able to represent points uniquely, and in these situations, we often make restrictions

$$0 \le r, \\ 0 \le \theta < 2\pi.$$

However, even with these restrictions, there still is a point that has multiple representations! Namely, the origin can be written as $(r, \theta) = (0, \theta)$ for any angle θ .

Depending on the situation and context, different people may use different restrictions or conventions for their ranges for r and θ . For this reason, it's good to specify what values you're allowing, to avoid being misunderstood!

Example 22. Let's consider the line described in Cartesian coordinates as the set of points (x, y) such that y = x. We'll figure out how to describe this line in polar coordinates.



Let's restrict our polar coordinates to $0 \le r$ and $0 \le \theta < 2\pi$. Perhaps your first guess is to describe the line as the points (r, θ) such that

$$\theta = \pi/4$$
.

Which shape does this describe?

Multiple Choice:

- (a) A point.
- (b) Half of the line. \checkmark
- (c) The whole line.
- (d) A different line.
- (e) A circle.

Describing the line as $\theta = \pi/4$ is a reasonable first guess, as we can see that many of the points make an angle $\pi/4$ with the positive x-axis. However, with the restriction that $r \geq 0$, this leaves out half of the line! In order to describe the entire line, we have a couple of options. One option would be to relax our restriction on r, and allow negative values as well. This would certainly give us the whole line. If, however, we would like to keep this restriction that $r \geq 0$, we could also include points with $\theta = 5\pi/4$, which will give us the other half of the line.

Which of the following describe the line y = x in polar coordinates? Select all that work.

Select All Correct Answers:

- (a) The points (r, θ) such that $\theta = \pi/4$, where $r \ge 0$.
- (b) The points (r, θ) such that $\theta = \pi/4$, where r can be any real number. \checkmark
- (c) The points (r, θ) such that $\theta = \pi/4$ or $\theta = -\pi/4$, where $r \ge 0$.
- (d) The points (r, θ) such that $\theta = \pi/4$ or $\theta = -\pi/4$, where r can be any real number.
- (e) The points (r, θ) such that $\theta = \pi/4$ or $\theta = 5\pi/4$, where $r \ge 0$.
- (f) The points (r, θ) such that $\theta = \pi/4$ or $\theta = 5\pi/4$, where r can be any real number. \checkmark

Example 23. Consider the set of points (r, θ) such that $r = 2\cos\theta$. What does this set of points look like?

It's not very clear from $r = 2\cos\theta$ what shape this is describing, so let's try converting this to Cartesian coordinates, and see if we get something we recognize.

Recall that the relationship between Cartesian and polar coordinates:

$$x = \boxed{r\cos\theta},$$
$$y = \boxed{r\sin\theta}.$$

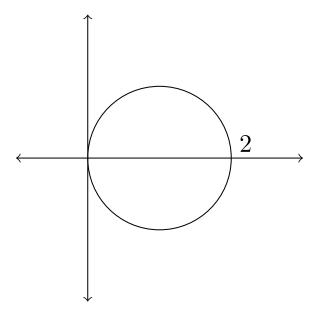
From this, we have that $r^2 = x^2 + y^2$, in terms of x and y, and $\cos \theta = \frac{x}{r}$. Making substitutions using these facts, we have:

$$r = 2\cos\theta$$
$$r = 2\frac{x}{r}$$
$$r^{2} = 2x$$
$$x^{2} + y^{2} = 2x$$

We now have an equation solely in terms of x and y, but maybe it isn't quite recognizable yet. But if we do a bit more algebra...

$$x^{2} + y^{2} = 2x$$
$$(x^{2} - 2x + 1) + y^{2} = 1$$
$$(x - 1)^{2} + y^{2} = 1$$

Now, we can see that this is a circle of radius 1 centered at (1,0).



Linear Change of Coordinates

In Linear Algebra, we saw how different coordinate systems arose through linear change of coordinates. You may remember this referred to as "slanty space."

When we write a point in Cartesian coordinates as (x, y), we can think of this as a linear combination of the standard basis vectors:

$$(x,y) = x(1,0) + y(0,1).$$

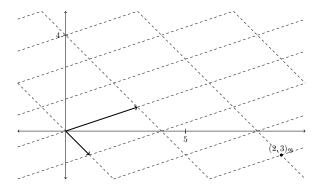
Of course, we can just as well write a point as a linear combination of vectors from a different basis, say (3,1) and (1,-1). Let's call this basis \mathfrak{B} For example, we can write the vector (9,-1) as

$$(9,-1) = 2(3,1) + 3(1,-1).$$

Taking the coefficients, in \mathfrak{B} -coordinates, we would write this point as

$$(2,3)_{\mathfrak{B}}$$
.

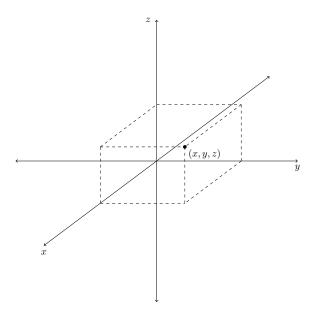
Note that we write $\mathfrak B$ in the subscript, in order to remind us that these are $\mathfrak B$ -coordinates, rather than standard Cartesian coordinates.



With linear changes of coordinates, it's easy to make a mistake and forget which coordinates you're using. Make sure to keep careful track!

Three-Dimensional Coordinates

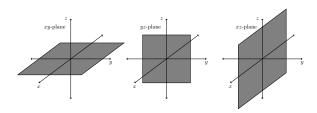
In Linear Algebra, we also worked in three-dimensional Cartesian coordinates, (x,y,z) in $\mathbb{R}^3.$



It's important to remember that the x, y, and z axes follow the right hand rule. That is, if you take your right hand, and point your pointer finger in the direction of the positive x-axis, point your middle finger in the direction of the positive y-axis, then your thumb points in the direction of the positive z-axis.

Another way to say this is that if you point the fingers of your right hand in the direction of the positive x-axis and curl them to point in the direction of the positive y-axis, your thumb points in the direction of the positive z-axis.

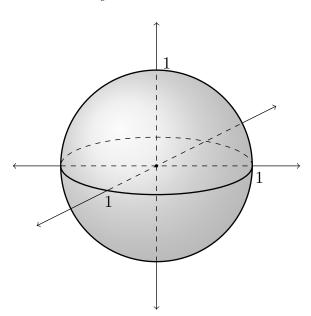
We'll often refer to the *coordinate planes* in \mathbb{R}^3 . These are the three planes we obtain by setting each of the coordinates to be zero.



More precisely, the xy-plane is the set of points (x, y, z) such that z = 0, the yz-plane is the set of points such that x = 0, and the xz-plane is the set of points such that y = 0.

Similarly to in the plane, we can describe sets of points in \mathbb{R}^3 using equations.

Example 24. The set of points (x, y, z) such that $x^2 + y^2 + z^2 = 1$ is the sphere of radius 1 centered at the origin in \mathbb{R}^3 .



Conclusion

In this activity, we reviewed coordinate systems that you've seen before: standard two-dimensional coordinates, polar coordinates, coordinates with respect to a given set of basis vectors, and three-dimensional coordinates.

Cylindrical Coordinates

In this activity, we introduce cylindrical coordinates, a new coordinate system on \mathbb{R}^3 .

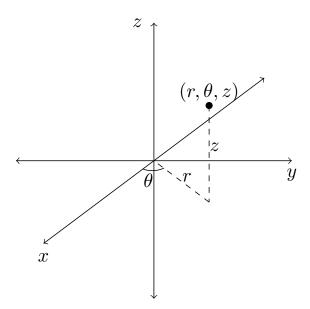
Cylindrical Coordinates

We've seen how points in \mathbb{R}^2 can be written using polar coordinates. Polar coordinates can be useful for describing shapes that are difficult to describe in Cartesian coordinates.

We'd now like to extend this idea to \mathbb{R}^3 , using a coordinate system called *cylindrical coordinates*. Like polar coordinates, cylindrical coordinates will be useful for describing shapes in \mathbb{R}^3 that are difficult to describe using Cartesian coordinates. Later in the course, we will also see how cylindrical coordinates can be useful in multivariable Calculus, when differentiating or integrating in Cartesian coordinates is difficult or impossible.

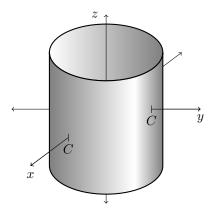
Cylindrical coordinates really are just a simple extension of polar coordinates. For points in the xy-plane, we describe them using r and θ , where r is the distance from the origin and θ is the angle with the positive x-axis. We then tack on a z-coordinate, the exact same as the z-coordinate in Cartesian coordinates, which tells us the vertical displacement of the point.

Learning outcomes: Author(s):



Example 25. Let's look at what happens in cylindral coordinates when we set each of the coordinates r, θ, z to be constant.

That is, we'll begin by examining the set of points (C, θ, z) , where C is a constant. We have that r = C is constant, which means that the distance between any such point and the z axis is constant, C. Also, θ and z can be anything. This will give us the cylinder of radius C, centered at the z-axis.



 $\theta = C$

z = C

Converting between Cartesian and cylindrical coordinates

Perhaps not surprisingly, converting between Cartesian coordinates and cylindrical coordinates is very similar to how we converted between Cartesian coordinates and polar coordinates. That is, we can use the equations:

$$x = r \cos \theta,$$

$$y = r \sin \theta,$$

$$z = z.$$

It can also be very useful to use the fact that $x^2 + y^2 = r^2$.

Spherical Coordinates

nothing to see here...

Learning outcomes: Author(s):