1. Subspace Drills

Determine if the following describe subspaces.

(a)
$$\{\vec{x} = \begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix}^T : x_i \ge 0 \ \forall i = 1, \cdots, n \}$$

Solution:

We proceed to test if the set is closed under vector addition and scalar multiplication. Let's check if this set is closed under vector addition and scalar multiplication.

• Vector addition: Let's consider two vectors \vec{v} and \vec{u} , both in the set.

$$\vec{v} \doteq \begin{bmatrix} v_1 & \cdots & v_n \end{bmatrix}^T$$

$$\vec{u} \doteq \begin{bmatrix} u_1 & \cdots & u_n \end{bmatrix}^T$$

where $v_i, u_i \ge 0 \ \forall i = 1, \dots, n$, i.e. all the elements of \vec{v}, \vec{u} are nonnegative. To see that V is closed under vector addition, we need $\vec{v} + \vec{u}$ to be contained within the set as well.

$$\vec{v} + \vec{u} = \begin{bmatrix} v_1 + u_1 & \cdots & v_n + u_n \end{bmatrix}^T$$

Because $v_i, u_i \ge 0$, then $v_i + u_i \ge 0$, and so $\vec{v} + \vec{u}$ is closed under vector addition.

• *Scalar multiplication*: Suppose we have a vector $\vec{v} \in V$ and a number $\alpha \in \mathbb{R}$. Notice that α can be negative!

To prove something false, we need only come up with a counterexample; so without loss of generality, let's say $\alpha = -1$. This means $\alpha \vec{v} = -\vec{v} = \begin{bmatrix} -v_1 & \cdots & -v_n \end{bmatrix}^T$, which does *not* necessarily have nonnegative entries! Thus, the set is not closed under scalar multiplication.

The set does not describe a vector subspace because it is not closed under scalar multiplication \square

(b) $\{\vec{0}\}$

Solution:

- *Vector addition*: Since $\vec{0}$ is the only vector in this set, $\vec{0} + \vec{0} = \vec{0} \in V$.
- *Scalar multiplication*: Since $a \cdot \vec{0} = \vec{0} \ \forall a \in \mathbb{R}$, we have that $a \cdot \vec{0} \in V$.

The set is closed under vector addition and scalar multiplication, and so it is a vector space! \Box

(c)
$$\left\{ \begin{bmatrix} 1\\0 \end{bmatrix}, \begin{bmatrix} 0\\1 \end{bmatrix} \right\}$$

Solution:

• Vector addition: We can see that

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

which is not in the set of vectors.

• Scalar multiplication: Taking either of the vectors and scaling them by $\alpha \neq 1$ returns a vector not in the set, and so it is not closed under scalar multiplication.

This is *not* a vector subspace. \Box

(d) span
$$\begin{pmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \end{pmatrix}$$

Solution: By definition, the span of N vectors is the set of the linear combinations of those N vectors. For this problem, we can rewrite the set as:

$$\left\{ a \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} a \\ b \end{bmatrix} : a, b \in \mathbb{R} \right\}$$

With the new definition, we see that the span of this set is \mathbb{R}^2 , which is a vector subspace. For the sake of completeness, however, we'll go through the standard tests:

• *Vector addition*: Let's define the following $\vec{v_1}$ and $\vec{v_2}$, both in the set.

$$\vec{v}_i = a_i \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b_i \begin{bmatrix} 0 \\ 1 \end{bmatrix}, i = 1, 2$$

$$\vec{v}_1 + \vec{v}_2 = \left(a_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b_1 \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) + \left(a_2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right)$$

$$= (a_1 + a_2) \begin{bmatrix} 1 \\ 0 \end{bmatrix} + (b_1 + b_2) \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$= a \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

• Scalar multiplication:

$$\vec{v} = a \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$
$$\alpha \vec{v} = \alpha a \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \alpha b \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$
$$= \hat{a} \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \hat{b} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

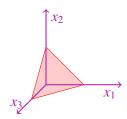
Hence, the set is a vector subspace \square

(e) (PRACTICE)

$$\left\{ \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \in \mathbb{R}^3 : x_1 + x_2 + x_3 = 1 \right\}$$
Solution:

Graphical Understanding

This set describes all points on the plane that passes through (1,0,0), (0,1,0), and (0,0,1).



(If it's not immediately clear why this is true, try thinking of the line $x_1 + x_2 = 1$.)

We know from the properties of vector spaces that one implication of closure under addition and scalar multiplication is that the vector space must contain the zero vector. Furthermore, using any two vectors in the space and adding them together disproves closure under vector addition; scaling a point in the plane by $\alpha \neq 1$ disproves closure under scalar multiplication.

Quick Counterexample Using the endpoints

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

and testing closure under scalar multiplication and addition is sufficient.

(f) (PRACTICE)

$$\left\{ \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \in \mathbb{R}^3 : x_1 + x_2 + x_3 = 0 \right\}$$

Solution: Similar to part (e), we can gain some understanding of what this is by considering the fact that it describes a plane which passes through the origin. However, this is not sufficient to prove that this (spoiler alert) is a vector space. We can do this by seeing if we can find a set of vectors whose span is the set.

The process is identical to finding a null space given a set of linear equations.

Looking at the equation we're given

$$x_1 + x_2 + x_3 = 0$$

we have one equation but three unknowns, meaning we can choose two free variables. Any two will work, but we'll say that x_1 and x_2 are free. That means we want to write x_3 in terms of the other two variables:

$$x_3 = -x_1 - x_2$$

So rewriting the original vector in terms of our free variables x_1 and x_2 :

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ -x_1 - x_2 \end{bmatrix}$$
$$= x_1 \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$$

We can now rewrite the set as span $(\begin{bmatrix} 1 & 0 & -1 \end{bmatrix}^T, \begin{bmatrix} 0 & 1 & -1 \end{bmatrix}^T)$. We know from part (d) that this set is indeed a vector space! \Box

(g) (CHALLENGE PRACTICE)

Let V be the vector space of $n \times n$ matrices, and $M \in V$ a fixed matrix. We define the set

$$U = \{ N \in V \mid NM = MN \}$$

as the *centralizer* of M in V. Show that U is a subspace of V. Solution:

• Vector Addition:

Note that the "vectors" in this case are actually matrices! Suppose we have $2 n \times n$ matrices $A, B \in U$, we want to show that $A + B \in U$, meaning that it must satisfy (A + B)M = M(A + B). Starting with (A + B)M, we expand it out into AM + BM. Now, since both A and B are in U, this means that AM = MA, BM = MB. Hence, (A + B)M = AM + BM = MA + MB = M(A + B).

• Scalar Multiplication:

Suppose we have $N \in U$, and $c \in \mathbb{R}$. Similar to the process above, we want to in the end show that (cN)M = M(cN).

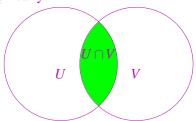
Since $N \in U$, we have that NM = MN. Since c is a real scalar, by the rule of matrix multiplication with scalars, we can move the scalar around within the expression. Hence, (cN)M = cNM = c(NM) = c(MN) = cMN = McN = M(cN).

Hence, we can now conclude that U is indeed a subspace of $V! \square$

(h) (CHALLENGE PRACTICE)

Suppose U and V are both subspaces of a vector space S, show that $U \cap V$ is also a subspace of S. Solution:

• *Vector Addition*: Consider 2 vectors $\vec{x}, \vec{y} \in U \cap V$. We want to show that $\vec{x} + \vec{y} \in U \cap V$. To show that this is true, it seems like a direct approach might be a bit hard since it seems unclear what exactly $U \cap V$ contains in terms of their properties. However, making use of the fact that $U \cap V \subset U, V$ (**The intersection of** U **and** V **is a subset of** U **and** V) will be crucial to the proof. It may help to consider this graphically:



We've marked in green the intersection between sets U and V.

If $\vec{x}, \vec{y} \in U \cap V$, both fall in the center region of the venn diagram. This means $\vec{x}, \vec{y} \in U$ as well as V! The implication goes both ways; if some vector \vec{z} falls in both U and V (the left and right circles), then it necessarily falls into their intersection, $U \cap V$.

We're already told that V and U are vector spaces, meaning $\vec{x} + \vec{y} \in U$ and V separately, so $\vec{x} + \vec{y} \in U \cap V$.

• Scalar Multiplication:

Consider a vector $\vec{x} \in U \cap V$, and a real-number scalar $c \in \mathbb{R}$.

Again, since $\vec{x} \in U \cap V \Longrightarrow \vec{x} \in V$, and V is already a vector subspace, so we know that by the property of scalar multiplication for a subspace, it must be true that $c\vec{x} \in V$.

Applying the same logic again for U, we can see $c\vec{x} \in U$.

Since we've shown that the set is closed under vector addition and scalar multiplication, $U \cap V$ is also a vector subspace of $S! \square$

2. Null Space Drill

In this question, we explore intuition about null spaces and a recipe to compute them. Recall that the nullspace of a matrix **M** is the set of all vectors, \vec{x} such that $\mathbf{M}\vec{x} = \vec{0}$.

(a) First, we begin by proving that a null space is indeed a subspace. Show that any nullspace of a matrix M with n rows and n columns is a subspace.

Solution:

- i. A nullspace of a matrix with n rows and n columns must contain vectors of n elements. These vectors clearly form a subset of \mathbb{R}^n .
- ii. \mathbb{R}^n is a known vector space.
- iii. Closures and 0
 - i. Consider two elements, $\vec{x_1}$ and $\vec{x_2}$ in the nullspace of **M**. By definition, we know that $\mathbf{M}\vec{x_1} = \vec{0}$ and $\mathbf{M}\vec{x_2} = \vec{0}$. Now consider the vector $\vec{x_1} + \vec{x_2}$. Is $\mathbf{M}(\vec{x_1} + \vec{x_2}) \stackrel{?}{=} \vec{0}$. $\mathbf{M}\vec{x_1} + \mathbf{M}\vec{x_2} = \vec{0} + \vec{0} = \vec{0}$. Done.
 - ii. Consider another element, $\vec{x_3}$ in the nullspace of **M**. Consider a scalar $a \in \mathbb{R}$. Is $a\vec{x_2}$ in the nullspace of **M**, i.e., is $\mathbf{M}(a\vec{x_3}) \stackrel{?}{=} \vec{0}$. Yes, because $a\mathbf{M}\vec{x_3} = 0$ since $\vec{x_3}$ is in the nullspace.
 - iii. Is $\vec{0}$ in the nullspace of M? Yes, because $\vec{M0} = \vec{0}$.

Therefore, a nullspace is indeed a subspace.

(b) Now we will explore a recipe to compute null spaces. Let's start with some 3×3 matrices.

$$\mathbf{A} = \begin{bmatrix} 1 & -3 & 1 \\ 2 & -8 & 8 \\ -6 & 3 & -15 \end{bmatrix}$$

A' is the row reduced matrix A.

$$\mathbf{A}' = \begin{bmatrix} 1 & -3 & 1 \\ 0 & -1 & 3 \\ 0 & 0 & -18 \end{bmatrix}$$

Compute the nullspace of A.

Solution: Since the row reduced matrix A' has a pivot in every column, the matrix has a trivial nullspace. The nullspace is the vector $\vec{0}$.

Let's look at this in more depth, however. Remember that a nullspace is the set of vectors such that $\mathbf{A}\vec{x} = \vec{0}$. Let's solve this as linear equations.

$$\begin{bmatrix} 1 & -3 & 1 \\ 2 & -8 & 8 \\ -6 & 3 & -15 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

To solve this, we row reduce. This results in

$$\begin{bmatrix} 1 & -3 & 1 \\ 0 & -1 & 3 \\ 0 & 0 & -18 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Let's convert this back to linear equations:

$$x_1 - 3x_2 + x_3 = 0$$
$$-x_2 + x_3 = 0$$
$$-18x_3 = 0$$

The third equation is only satisfied by $x_3 = 0$. The second equation implies that $x_2 = x_3 = 0$. And finally, the first equation is also only satisfied by $x_1 = 0$. Therefore, $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$ is the only vector which satisfies these equations

(c) Consider another matrix

$$\mathbf{B} = \begin{bmatrix} 1 & -1 & 2 \\ 4 & 4 & -2 \\ -2 & 2 & -4 \end{bmatrix}$$

 \mathbf{B}' is row reduced \mathbf{B} .

$$\mathbf{B}' = \begin{bmatrix} 1 & -1 & 2 \\ 0 & 8 & -10 \\ 0 & 0 & 0 \end{bmatrix}$$

What is the null space of **B**? What is the dimension of the row space of **B**?

Solution: Think of this as linear equations once again. Let the first column correspond to x, the second to y and the third to z. In equation form, the row reduced matrix becomes

$$x - y + 2z = 0 \tag{1}$$

$$8y - 10z = 0 (2)$$

$$0x + 0y + 0z = 0 (3)$$

Equation 3 gives us no information – it is always true. So we ignore it.

Equation 2 says that 4y = 5z. Let's set z = t (let z be a free variable that can take on any value). Then $y = \frac{5}{4}t$.

Equation 1 is then $x - \frac{5}{4}t + 2t = 0 \implies x = \frac{-3}{4}t$. The nullspace is then all vectors of the form $t \begin{bmatrix} -\frac{3}{4} \\ \frac{5}{4} \\ 1 \end{bmatrix}$, where t is any real number. Another way to say this is that the nullspace is spanned by the vector

$$\begin{bmatrix} \frac{-3}{4} \\ \frac{5}{4} \\ 1 \end{bmatrix} \tag{4}$$

The dimension of the nullspace, i.e., the minimum number of vectors required to span it is 1. From the rank-nullity theorem, we know that Dim(Rowspace(B)) + Dim(Nullspace(B)) = Number of columns in B. Therefore, the dimension of the rowspace of B is 2.

(d) In the previous part, we chose one of the variables and set it to be a free variable. Can we choose any variable as our free variable?

Solution: Let's investigate this question by choosing each variable as a free variable. We know z works from the solution to the previous part.

Let's consider y. If we set y = t instead, then we get, from Equation (2) $5z = 4t \implies z = \frac{4}{5}t$.

Equation (1) then gives us
$$x - t + 2\frac{4}{5}t = 0 \implies x - \frac{5t}{5} + \frac{8t}{5} = 0 \implies x = \frac{-3t}{5}$$
.

The nullspace is then spanned by the vector
$$\begin{bmatrix} \frac{-3}{5} \\ 1 \\ \frac{4}{5} \end{bmatrix}$$

Note that this vector is $\frac{4}{5}$ times the vector we found in (4).

Solution:

Now, let's see what happens if we set x = t instead. We can't use Equation (2) yet, so let's try using Equation (1). t - y + 2z = 0. Now what? How do we find the value for y or z in terms of t? ...We can't. So x does not work...

(e) How can we know which variables can be used as free variables?

Solution: Pick your free variables are by looking at columns with no pivots. Although, sometimes, other variables might work (like *y* above), the variables with no pivots will always work!

(f) Now consider another matrix, $\mathbf{C} = \begin{bmatrix} 1 & -2 & -6 & 12 \\ 2 & 4 & 12 & -17 \\ 1 & -4 & -12 & 22 \end{bmatrix}$ Without doing any math, will this matrix

have a trivial nullspace, i.e. consisting of only $\vec{0}$?

Solution: No! A 3x4 matrix can simply not have 4 pivots. So at least one of the variables will need to be free!

(g) (PRACTICE)

Consider another matrix,
$$\mathbf{D} = \begin{bmatrix} 1 & -2 & -6 & 12 \\ 0 & -2 & -6 & 10 \\ 0 & 0 & 0 & -1 \end{bmatrix}$$
. Find vector(s) that span the nullspace.

Solution: In terms of equations, let the variables for cols 1-4 be a to d respectively. Column 3 does not have a pivot. So c is free. Let c = t. At this point, we should feel comfortable reading the matrix as its equations without explicitly writing the equations!

Row 3 of the matrix says that -d = 0, or that d = 0.

Row 2 says that
$$-2b-6c+10d=0 \implies -2b-6t=0 \implies b=-3t$$
.

Row 1 says that
$$a-2b-6c+12d=0 \implies a+6t-6t=0 \implies a=0$$
.

The vector that spans this nullspace is
$$\begin{bmatrix} 0 \\ -3 \\ 1 \\ 0 \end{bmatrix}$$

(h) (PRACTICE)

Consider one final matrix, $\mathbf{E} = \begin{bmatrix} 1 & -2 & -6 & 12 \\ 0 & -2 & -6 & 10 \\ 0 & 0 & 0 & 0 \end{bmatrix}$. What are the vector(s) that span this nullspace?

Solution: Again, let the variables for the columns be *a* to *d* respectively. Columns 3 and 4 don't have pivots. So let's set both of them to be free!

Let c = t, d = s.

Row 2 says
$$-2b - 6c + 10d = 0 \implies -2b = 6t - 10s \implies b = -3t + 5s$$
.

Row 1 says $a - 2b - 6c + 10d = 0 \implies a = 0$.

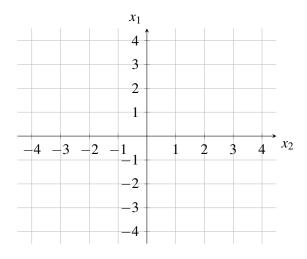
The general form of vectors in the nullspace is then $\begin{bmatrix} 0 \\ -3t + 5s \\ t \\ s \end{bmatrix}$. This needs to be rewritten by splitting

the free variables $s \begin{bmatrix} 0 \\ 5 \\ 0 \\ 1 \end{bmatrix} + t \begin{bmatrix} 0 \\ -3 \\ 1 \\ 0 \end{bmatrix}$.

Finally we conclude that the vectors that span the nullspace are $\begin{bmatrix} 0 \\ 5 \\ 0 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 0 \\ -3 \\ 1 \\ 0 \end{bmatrix}$.

Observation: notice that the number of free variables = number of columns without pivot = number of vectors required to span the nullspace = dimension of the nullspace!

3. Range Intuition



$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}, \quad \vec{x} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

(a) Draw the space on the figure above that is represented by Col(A). Also draw the space for the Row(A) (which is the same as $Col(A^{\top})$. What dimension are these spaces?

Solution: The 1 dimensional space for the column space is a line on the $x_2 = x_1$ axis (so a diagonal line that goes perfectly northeast, intersecting the origin along the way). The 1 dimensional space for the row space is the line $x_2 = 2x_1$

(b) Plot the point \vec{x} , then plot $A\vec{x}$

Solution: A \vec{x} should be on the line that represents Col(A), specifically it should be $\begin{bmatrix} 5 \\ 5 \end{bmatrix}$

(c) Consider some arbitrary vector $\vec{v} = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$ Write out the product $\mathbf{A}\vec{v}$ in terms of v_1 , v_2 , and the columns of \mathbf{A} .

Solution:

$$\mathbf{A}\vec{v} = v_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + v_2 \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

(d) We have talked about how matrices like **A** have no inverse. Give a geometric explanation for why this is the case.

Solution: If we are given some point on the line for the colspace from part (a), we do not know where it came from. For example, if you gave the point given by $\mathbf{A}\vec{x}$, you have no way of knowing that it came from \vec{x}

(e) Consider all points \vec{y} such that $A\vec{y} = 0$ Draw the space that the \vec{y} 's will make up. What do you notice geometrically? What is the dimension of this space?

Solution: This line should be a straight line that is perpendicular to the line for the row space from part (a). It is a space of dimension 1

4. Mechanical Eigenvectors and Eigenvalues

(a) Solve for the eigenvalue-eigenvector pairs for the following 2 by 2 matrix:

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

Solution: To solve for eigenvalues and eigenvectors, let's go back and review the definition of eigenvectors and eigenvalues:

If \vec{x} and λ are the eigenvector and eigenvalue of **A**, respectively, then the following equation holds:

$$\mathbf{A}\vec{x} = \lambda \vec{x}$$

Since the (appropriately sized) identity matrix is analogous to multiplying by 1 in arithmetic, we can say:

$$\mathbf{A}\vec{x} = \lambda \mathbf{I}\vec{x}$$

Rearranging, we get:

$$\mathbf{A}\vec{x} - (\lambda \mathbf{I})\vec{x} = \vec{0}$$
$$(\mathbf{A} - \lambda \mathbf{I})\vec{x} = \vec{0}$$

What does this look like? It looks similar to solving for the nullspace of $(\mathbf{A} - \lambda \mathbf{I})!$ Assuming that there is a nontrivial nullspace, that also means that $\det(\mathbf{A} - \lambda \mathbf{I}) = 0!$ Let's solve for λ first:

$$(\mathbf{A} - \lambda \mathbf{I}) = \begin{bmatrix} 3 & 2 \\ 1 & 4 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix}$$
$$= \begin{bmatrix} 3 - \lambda & 2 \\ 1 & 4 - \lambda \end{bmatrix}$$
$$\mathbf{det}(\mathbf{A} - \lambda \mathbf{I}) = (3 - \lambda)(4 - \lambda) - 2$$
$$= 10 - 7\lambda + \lambda^2$$
$$= (\lambda - 5)(\lambda - 2)$$

By factoring:

$$\lambda = 5.2$$

Let's check: We've just solved for the eigenvalues. But what about the eigenvectors? To do that, we plug in λ into $(\mathbf{A} - \lambda \mathbf{I})$ and solve for the nullspace! For $\lambda = 5$:

$$(\mathbf{A} - \lambda \mathbf{I})\vec{x} = \vec{0}$$

$$\begin{bmatrix} 3 - \lambda & 2 \\ 1 & 4 - \lambda \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \vec{0}$$

$$\begin{bmatrix} -2 & 2 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \vec{0}$$

By row reduction:

$$\begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \vec{0}$$
$$x_1 = x_2$$
$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} x_2$$

So the first pair is

$$\lambda = 5, \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Repeating for $\lambda = 2$,

$$\begin{bmatrix} 3 - \lambda & 2 \\ 1 & 4 - \lambda \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \vec{0}$$
$$\begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \vec{0}$$
$$\begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \vec{0}$$
$$x_1 = -2x_2$$
$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -2 \\ 1 \end{bmatrix} x_2$$

So, the second pair is

$$\lambda = 2, \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

(b) Find the eigenvectors for matrix A given that we know that $\lambda_1 = 4, \lambda_2 = \lambda_3 = -2$ and that

$$\mathbf{A} = \begin{bmatrix} 1 & -3 & 3 \\ 3 & -5 & 3 \\ 6 & -6 & 4 \end{bmatrix}$$

Solution: Once the eigenvalues of a matrix have been found, we can find the eigenvectors by Gaussian Elimination.

Step 1: For each eigenvalue λ , we have

$$(\mathbf{A} - \lambda \mathbf{I})\vec{x} = 0$$

where \vec{x} is the eigenvector associated with eigenvalue λ .

Step 2: Find in the nullspace of $(A - \lambda I)$ by plugging in a value of λ and using Gaussian elimination to solve.

Case 1: $\lambda = 4$. First, form the matrix $\mathbf{A} - 4\mathbf{I}$:

$$\mathbf{A} - 4\mathbf{I} = \begin{bmatrix} -3 & -3 & 3 \\ 3 & -9 & 3 \\ 6 & -6 & 0 \end{bmatrix}$$

To make our numbers nicer, first let's divide our first row by -3

$$R_{1} = R_{1} \cdot \frac{-1}{3}$$

$$\begin{bmatrix} 1 & 1 & -1 \\ 3 & -9 & 3 \\ 6 & -6 & 0 \end{bmatrix}$$

$$R_{2} = R_{2} - 3 \cdot R_{1}$$

$$R_{3} = R_{3} - 6 \cdot R_{1}$$

$$\begin{bmatrix} 1 & 1 & -1 \\ 0 & -12 & 6 \\ 0 & -12 & 6 \end{bmatrix}$$

$$R_{3} = R_{3} - R_{2}$$

$$R_{2} = R_{2} \cdot \frac{1}{6}$$

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

Now, we see that we have reached a row of 0s, which means that our last variable x_3 is the free variable in our system. Now, we can expand this matrix by putting it into a system of linear equations and solving for all the variables in terms of our free variable x_3

$$x_1 + x_2 - x_3 = 0$$

$$-2x_2 + x_3 = 0$$

$$x_2 = \frac{x_3}{2}$$

$$x_1 + \frac{x_3}{2} - x_3 = 0$$

$$x_1 = \frac{x_3}{2}$$

$$\vec{x} = \begin{bmatrix} \frac{x_3}{2} \\ \frac{x_3}{2} \\ x_3 \end{bmatrix}$$

$$= x_3 \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ 1 \end{bmatrix}, \forall x_3 \in \mathbb{R}$$

So the eigenvector for when $\lambda = 4$ is $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$. Now, let's use this same technique to find the eigenvector for $\lambda = -2$

Solution: Case 2: Now let's plug in $\lambda = -2$ into $\mathbf{A} - \lambda \mathbf{I}$ to get

$$\mathbf{A} + 2\mathbf{I} = \begin{bmatrix} 3 & -3 & 3 \\ 3 & -3 & 3 \\ 6 & -6 & 6 \end{bmatrix}$$

And, just like before, let's use Gaussian elimination to reduce the matrix. We can see that this will only take a few steps.

$$R_{2} = R_{2} - R_{1}$$

$$R_{3} = R_{3} - 2 \cdot R_{1}$$

$$R_{1} = R_{1} \cdot \frac{1}{3}$$

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

As we can see here, we have two rows of 0s, which means that we have two free variables (x_2 and x_3). Now we can take this matrix and write it as a linear system to get

$$x_1 + x_2 - x_3 = 0 \Rightarrow x_1 = x_3 - x_2$$

Thus,

$$\vec{x} = \begin{bmatrix} x_3 - x_2 \\ x_2 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + x_2 \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

Which are the two eigenvectors associated with $\lambda = -2$

(c) Find the eigenvalues for matrix **A** given that we know that $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \vec{v}_3 = \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}$ are the eigenvectors of **A**, and that

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

Solution: There are 2 ways to go about solving this problem. Either you can plug each eigenvector \vec{v}_i into $\mathbf{A}v = \lambda v$ or the nullspace equation to come up with 3 equations and solve. As you have had a lot of practice with the latter, we will use the former to try to answer this question.

Let's plug in the first eigenvector and solve for the first eigenvalue.

$$\mathbf{A}\vec{v}_{1} = \lambda_{1}\vec{v}_{1}$$

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

So, we can see that $\lambda_1 = 1$. Similarly, we can do this for the other two eigenvectors.

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

So, we can see that $\lambda_2 = 2$.

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

So, we can see that $\lambda_3 = 3$.

5. (PRACTICE) Nullspaces and Projections

Assume that the vector $\vec{x} = \begin{bmatrix} x_0 \\ x_1 \end{bmatrix}$. For each of the following matrices $\mathbf{A} \in \mathbb{R}^{n \times m}$, answer the following:

- Compute the matrix product $A\vec{x}$. Explain in words how the matrix transforms the vector.
- Suppose you know that A transforms \vec{x} to give \vec{y} . Given \vec{y} , can you find what the original vector \vec{x} was?
- Is the matrix **A** invertible? How do you know? If it is invertible, find the inverse.
- Verify that (dimension of nullspace) + (dimension of column space) = min(n, m)

(a)
$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \vec{y} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

Solution:

- $\mathbf{A}\vec{x} = \begin{bmatrix} x_0 \\ 0 \end{bmatrix}$. We can see that this matrix keeps x_0 and turns x_1 to 0. In other words, the matrix **projects** the vector \vec{x} to the *x*-axis.
- No. We can figure out x_0 , since the value does not change, but the transformation converts all the x_1 values to 0. Given 0 as an output, x_1 could have been any value, so we cannot determine the original vector \vec{x} . As a rule of thumb, if there isn't a one-to-one mapping of inputs to outputs for the elements of \vec{x} , we cannot find the original vector after the transformation.
- No, A is not invertible, since the columns are linearly dependent, or there is no way to turn it into an upper triangular matrix using Gaussian elimination. Intuitively, we cannot retrieve a vector to its original state after applying the matrix, and therefore we cannot "invert" the operation.
- The dimension of the nullspace is 1, or the number of elements that cannot be retrieved after the matrix transformation. The column space is dimension 1, since there is only one pivot in the matrix. Adding both together, we get 1 + 1 = 2, which is equal to $\min(2, 2) = 2$.

(b)
$$\mathbf{A} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}, y = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Solution:

• $\mathbf{A}\vec{x} = \frac{1}{2}\begin{bmatrix} x_0 + x_1 \\ x_0 + x_1 \end{bmatrix}$ The first and second entries of the resulting vector have the same value; $x_0 = x_1$. We can consider the first entry to be equivalent to "x" and the second to "y", Hence, it is a projection onto y = x. Note that the constants are $\frac{1}{2}$ so that the x vector is not scaled.

- No, we cannot retrieve the original values because when we write out the equations represented by the transformation, the two equations that are the same (linearly dependent), so we only have one equation to work with. This is not enough information to retrieve the original information; we are trying to solve for two variables with one equation.
- No; this goes hand-in-hand with the whether or not we can retrieve the original values; the matrix has linearly dependent columns so we cannot invert it. We also know that we cannot invert matrices with determinant equal to 0, and it is clear that this transformation matrix has determinant 0.
- The column space of the matrix is 1 since we retrieved one linearly independent equation from earlier. To calculate the nullspace, we first row reduce the matrix, giving us $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$. Then we solve for $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} x_0 \\ x_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$. This gives us the equation $x_0 + x_1 = 0$. Solving this equation, we get $x_0 = -x_1$, so the nullspace is

This gives us the equation $x_0 + x_1 = 0$. Solving this equation, we get $x_0 = -x_1$, so the nullspace is $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$. Hence, we have one vector in the column space and one vector in the nullspace; 1 + 1 = 2, the rank of the original transformation, so everything is accounted for.

(c)
$$\mathbf{A} = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}, \vec{y} = \begin{bmatrix} \sqrt{2} \\ 0 \end{bmatrix}$$

Solution:

- $\mathbf{A}\vec{x} = \frac{1}{\sqrt{2}} \begin{bmatrix} x_0 x_1 \\ x_0 + x_1 \end{bmatrix}$. However, this does not offer us much insight into what the matrix actually **does**. For that, we notice that this is actually the rotation matrix corresponding to an angle of $\theta = \frac{\pi}{4}$. $\mathbf{A} = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} = \begin{bmatrix} \cos\frac{\pi}{4} & -\sin\frac{\pi}{4} \\ \sin\frac{\pi}{4} & \cos\frac{\pi}{4} \end{bmatrix}$ So, we can interpret this matrix as being a transformation that takes a vector and shifts it counterclockwise by 45°.
- Yes. We know that $x_0 x_1 = 2$ and $x_0 + x_1 = 0$. Solving these, we get $\vec{x} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$. This makes sense, since we can think of the vector $\begin{bmatrix} \sqrt{2} \\ 0 \end{bmatrix}$ as being a 45-degree-rotated version of the vector $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$.
- Yes, the matrix is invertible. We know this because the columns are linearly independent (replace row 2 with row 2 row 1 to get an upper triangular matrix). This also makes sense intuitively, because we know that we can reverse a rotation by applying its inverse rotation (a clockwise rotation by 45 degrees). So, in this case, the inverse matrix will be $\mathbf{A}^{-1} = \begin{bmatrix} \cos -\frac{\pi}{4} & -\sin -\frac{\pi}{4} \\ \sin -\frac{\pi}{4} & \cos -\frac{\pi}{4} \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}$. You can verify that $\begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \cdot \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$
- Because this matrix is invertible, it has a nullspace of dimension 0. And since it is full rank (rows are lin. ind.), its column space has dimension 2. Their sum is equal to min(2, 2) = 2. Thus the equation is verified.

(d)
$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}, \vec{y} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$
Solution:

- $\mathbf{A}\vec{x} = \begin{bmatrix} x_0 \\ 2x_1 \end{bmatrix}$. This is a diagonal matrix, and performs component-wise scaling. We can interpret this as a transformation that scales the first component by 1, and the second component by 2.
- Yes. We know that $x_0 = 2$ and $2x_1 = 4$. Solving these, we get $\vec{x} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$. This makes sense, because the components are getting scaled by 1 and 2 respectively.
- Yes, the matrix is invertible. We know this because the columns are linearly independent (the matrix is already in upper triangular form). We also know that all diagonal matrices with nonzero entries are invertible. This also makes sense intuitively, because we know that we can reverse each component's scaling by applying its inverse scaling. So, in this case, the inverse matrix will be $A^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & \frac{1}{2} \end{bmatrix}.$ You can verify that $\begin{bmatrix} 1 & 0 \\ 0 & \frac{1}{2} \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$
- Because this matrix is invertible, it has a nullspace of dimension 0. And since it is full rank (rows are lin. ind.), its column space has dimension 2. Their sum is equal to $\min(2, 2) = 2$. Thus the equation is verified.