

Optimization and Computational Linear Algebra

Homework 1 - Fall 2023

New York University

Jiasheng Ni

jn2294

1 Problem 1.1

Are the following sets subspaces of \mathbb{R}^2 ? Draw a picture and justify your answer using the definition of a subspace. If yes, indicate the dimension of the subspace and a basis, add the basis vectors on your drawing.

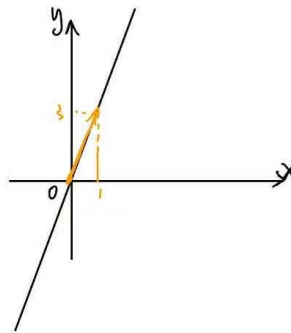
- (a) $E_1 = \{(x, y) \in \mathbb{R}^2 \mid 3x - y = 0\}$.

We first show that it is a subspace.

- $\vec{0} = (0, 0) \in E_1$. Zero vector is in E_1 .
- $\forall \vec{v} = (a_1, a_2), \vec{w} = (b_1, b_2) \in E_1$, we have $3a_1 = a_2, 3b_1 = b_2$. Then $\vec{v} + \vec{w} = (a_1 + b_1, a_2 + b_2) = (a_1 + b_1, 3a_1 + 3b_1) \in E_1$. Thus E_1 is closed under addition.
- $\forall \vec{v} = (a_1, a_2) \in E_1, \alpha \in \mathbb{R}$, we have $3a_1 = a_2$. Then $\alpha\vec{v} = (\alpha a_1, \alpha a_2) \in E_1$. Thus E_1 is closed under scalar multiplication.

Then the basis for it: $\forall \vec{u} = (x, y) \in E_1$, we have $\vec{u} = (x, 3x) = x(1, 3), x \in \mathbb{R}$. So E_1 is a line passing through the origin towards the $(1, 3)$. The basis is then $\left\{ \begin{bmatrix} 1 \\ 3 \end{bmatrix} \right\}$ and the dimension $\dim(E_1) = 1$.

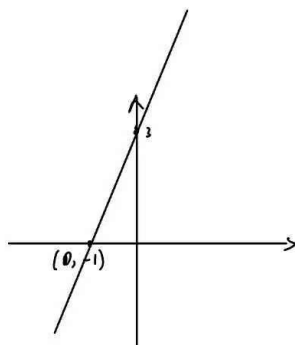
Finally the drawing(basis shown in orange):



- (b) $E_2 = \{(x, y) \in \mathbb{R}^2 \mid 3x - y = -1\}$.

It is **not** a subspace since $\vec{0} \notin E_2$ but it is a translation of subspace(affine subspace). $\forall \vec{u} = (x, y) \in E_2$, we have $\vec{u} = (x, 3x - 1) = x(1, 3) + (0, -1), x \in \mathbb{R}$. So E_2 is a line passing through the point $(0, -1)$ along the direction $(1, 3)$.

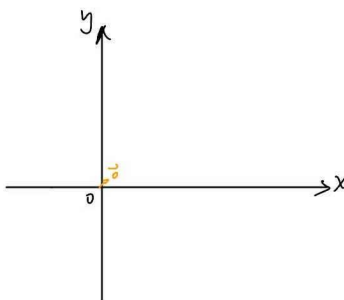
The drawing is:



- (c) $E_3 = \{(x, y) \in \mathbb{R}^2 \mid x^2 + y^2 = 0\}$.

It is a subspace and the only element in E_3 is $\vec{0}$ and the dimension is $\dim(E_3) = 0$.

The drawing is (basis shown in orange):



- (d) $E_4 = \{(x, y) \in \mathbb{R}^2 \mid x + y \geq 0\}$.

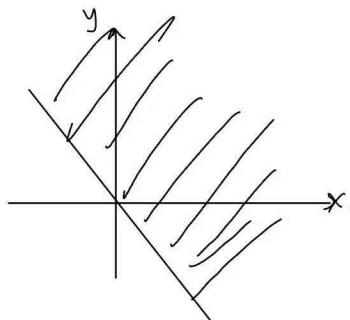
It is a halfspace and thus **not** a subspace.

We verify it as follows:

1. $\vec{0} = (0, 0) \in E_4$. Zero vector is in E_4 .
2. $\forall \vec{v} = (a_1, a_2), \vec{w} = (b_1, b_2) \in E_4$, we have $a_1 + a_2 \geq 0, b_1 + b_2 \geq 0$. Then $\vec{v} + \vec{w} = (a_1 + b_1, a_2 + b_2) \in E_4$. Thus E_4 is closed under addition.
3. $\forall \vec{v} = (a_1, a_2) \in E_4, \alpha \in \mathbb{R}$, we have $a_1 + a_2 \geq 0$ but $\alpha(a_1 + a_2)$ is not necessarily greater or equal to zero if we choose $\alpha < 0$. Then $\alpha\vec{v} = (\alpha \cdot a_1, \alpha \cdot a_2) \notin E_4$. So E_4 is not closed under scalar multiplication.

. Thus E_4 is not a subspace.

The drawing is:



2 Problem 1.2

Are the following sets subspaces of \mathbb{R}^3 ? Justify your answer using the definition of a subspace. If yes, indicate the dimension of the subspace and a basis.

1. (a) $E_5 = \{(x, y, z) \in \mathbb{R}^3 \mid x = 0 \text{ and } y - z = 0\}$.

Yes, and we will prove it by definition:

- 1. Zero Vector: $\vec{0} \in E_5$ in that $\vec{0} = (0, 0, 0) = (x, y, z)$ where $x = 0, y = 0, z = 0$ and $y = z$.
- 2. Addition: $\forall \vec{x} = (a_1, a_2, a_3), \vec{y} = (b_1, b_2, b_3) \in E_5$, we have $a_1 = b_1 = 0, a_2 = a_3, b_2 = b_3$ and that $a_1 + a_2 = 0, a_2 + b_2 = a_3 + b_3$. Then $\vec{x} + \vec{y} = (a_1 + b_1, a_2 + b_2, a_3 + b_3) \in E_5$. So E_5 is closed under addition.
- 3. Scalar Multiplication: $\forall \vec{x} = (a_1, a_2, a_3)$ and $\alpha \in \mathbb{R}$, we have $a_1 = 0, a_2 = a_3$. So $\alpha \cdot a_1 = 0, \alpha \cdot a_2 = \alpha \cdot a_3$. This means $\alpha \vec{x} \in E_5$. Thus, E_5 is closed under scalar multiplication.

Bring all these pieces together, E_5 is a vector space.

For $\forall v = (x, y, z) \in E_5$, we have $v = (0, y, y) = y(0, 1, 1)$, thus the dimension of E_5 is 1 and one valid basis is $(0, 1, 1)$.

2. (b) $E_6 = \{(x, y, z) \in \mathbb{R}^3 \mid x = 0 \text{ or } y - z = 0\}$.

No. And we will show that it doesn't satisfy the closure under addition.

$\forall \vec{x} = (a_1, a_2, a_3) \in E_6$, we have $a_1 = 0$ or $a_2 = a_3$. So if we pick $\vec{u} = (0, 1, 2)$ and $\vec{v} = (2, 3, 3)$, we have $\vec{u} + \vec{v} = (2, 4, 5)$ where $\vec{u} + \vec{v} \notin E_6$. So E_6 is not a subspace.

3 Problem 1.3

Let us define the vectors $e_1, \dots, e_n \in \mathbb{R}^n$ by

$$\begin{aligned} e_1 &= (1, 0, 0, \dots, 0) \\ e_2 &= (0, 1, 0, \dots, 0) \\ &\vdots \\ e_n &= (0, 0, 0, \dots, 1). \end{aligned}$$

1. (a) Verify that the family (e_1, \dots, e_n) is a basis of \mathbb{R}^n . This basis is called the "canonical basis" of \mathbb{R}^n . What is the dimension of \mathbb{R}^n ?

We examine the following systems of linear equations: $a_1 \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + a_2 \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} + \dots + a_n \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$, and we solve

it to get $a_1 = 0, a_2 = 0, \dots, a_n = 0$, which implies that $\{\vec{e}_1, \dots, \vec{e}_n\}$ are linearly independent.

To show $\{\vec{e}_1, \dots, \vec{e}_n\}$ spans \mathbb{R}^n , we notice that $\forall \vec{v} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} \in \mathbb{R}^n$, we have $\vec{v} = \sum_{i=1}^n a_i \vec{e}_i \in \text{Span}(\{\vec{e}_1, \dots, \vec{e}_n\})$,

which ends our proof of basis.

Finally by definition, we have $\dim(\mathbb{R}^n) = n$ since it has a length- n basis.

2. (b) Give an example of hyperplane and an example of a line of \mathbb{R}^n using spans of subsets of (e_1, \dots, e_n)
 - (a) An example of $n - 1$ -dimensional hyperplane can be defined as the $\mathcal{H} = \{\vec{x} \in \mathbb{R}^n | \vec{e}_1^\top \vec{x} = 0\}$ since $\text{Span}(\{\vec{e}_1\}) \oplus \mathcal{H} = \mathbb{R}^n$, where $\dim(\text{Span}(\{\vec{e}_1\})) + \dim(\mathcal{H}) = \dim(\mathbb{R}^n) = n$ and $\text{Span}(\{\vec{e}_1\}) = 1$ and $\dim(\mathcal{H}) = n - 1$.
 - (b) An example of a line of \mathbb{R}^n could be formed by any of the basis vector \vec{e}_i . For example, $S = \{\vec{x} \in \mathbb{R}^n | \vec{x} = \alpha \vec{e}_i, \alpha \in \mathbb{R}\}$.

4 Problem 1.4

1. (a) Consider $v_1, \dots, v_p \in \mathbb{R}^n$. Prove that $\text{Span}(v_1, \dots, v_p)$ is the smallest subspace which contains v_1, \dots, v_p (i.e., it is a subspace which contains v_1, \dots, v_p , and it is contained in any other such subspace).

- We first prove that $\text{Span}(\vec{v}_1, \dots, \vec{v}_n)$ contains $\vec{v}_1, \dots, \vec{v}_n$.

Proof. Since $\vec{v}_i = 0 \cdot \vec{v}_1 + 0 \cdot \vec{v}_2 + \dots + 1 \cdot \vec{v}_i + \dots + 0 \cdot \vec{v}_n \forall i = 1, 2, \dots, n$, which is a linear combination of $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$. Then we know that $\forall \vec{v}_i, \vec{v}_i \in \text{Span}(\vec{v}_1, \dots, \vec{v}_n)$ by definition of span. \square

- We then prove that it is a subspace.

Proof. – Zero Vector: Taking all the coefficients before $\vec{v}_i, i = 1, 2, \dots, n$ to be zero we know that $\vec{0} \in \text{Span}(\vec{v}_1, \dots, \vec{v}_n)$

– Addition: $\forall \vec{v}, \vec{w} \in \text{Span}(\vec{v}_1, \dots, \vec{v}_n)$, by definition we have $\vec{v} = \sum_{i=1}^n \alpha_i \vec{v}_i$ and $\vec{w} = \sum_{i=1}^n \beta_i \vec{v}_i$. Then $\vec{v} + \vec{w} = \sum_{i=1}^n (\alpha_i + \beta_i) \vec{v}_i \in \text{Span}(\vec{v}_1, \dots, \vec{v}_n)$.

– Scalar Multiplication: $\forall \vec{v} \in \text{Span}(\vec{v}_1, \dots, \vec{v}_n), t \in \mathbb{R}$, by definition we have $\vec{v} = \sum_{i=1}^n \alpha_i \vec{v}_i$. Then $\alpha \vec{v} = \sum_{i=1}^n (t\alpha_i) \vec{v}_i \in \text{Span}(\vec{v}_1, \dots, \vec{v}_n)$.

Thus $\text{Span}(\vec{v}_1, \dots, \vec{v}_n)$ is a subspace. \square

- Finally, we prove that it is the smallest such subspaces.

Proof. We know that any subspaces containing $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$ must contain the linear combination of them (which is just $\text{Span}(\vec{v}_1, \dots, \vec{v}_n)$) since any subspaces are closed under addition and scalar multiplication (in short, linear combination). In other words, any subspaces containing $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$ must contain $\text{Span}(\vec{v}_1, \dots, \vec{v}_n)$. Thus, $\text{Span}(\vec{v}_1, \dots, \vec{v}_n)$ is the smallest such subspaces. \square

2. (b) Let V, W be two subspaces of \mathbb{R}^n . Show that $V \cap W$ is a subspace of \mathbb{R}^n .

We will prove by definition:

- Zero Vector: Since V and W are both subspaces, we know $\vec{0} \in V$ and $\vec{0} \in W$, thus $\vec{0} \in V \cap W$ by definition.
- Addition: $\forall \vec{v}, \vec{w} \in V \cap W$, we have $\vec{v}, \vec{w} \in V$ and $\vec{v}, \vec{w} \in W$. Since V, W are both subspaces, we have $\vec{v} + \vec{w} \in V$, and $\vec{v} + \vec{w} \in W$, which implies $\vec{v} + \vec{w} \in V \cap W$ by definition. Thus, $V \cap W$ is closed under addition.
- Scalar Multiplication: $\forall \vec{x} \in V \cap W, \alpha \in \mathbb{R}$, we have $\vec{x} \in V$ and $\vec{x} \in W$. Since V, W are both subspaces, we have $\alpha \vec{x} \in V$, and $\alpha \vec{x} \in W$. Thus, $\alpha \vec{x} \in V \cap W$. So $V \cap W$ is closed under scalar multiplication.

To sum up, $V \cap W$ is a subspace.

3. (c) Let V, W be two subspaces of \mathbb{R}^n . Show that $V \cup W$ may not be a subspace of \mathbb{R}^n with a counter-example.

We will prove by counter-example. Suppose that $V = \{(x, y) \in \mathbb{R}^2 | x = 3y\}, W = \{(x, y) \in \mathbb{R}^2 | x = y\}$. Then if we choose $\vec{v} = (1, 3) \in V$ and $\vec{w} = (1, 1) \in W$, then surely $\vec{v} \in V \cup W$ and $\vec{w} \in V \cup W$. Properties of set. Then $\vec{v} + \vec{w} = (2, 4) \notin V$ and $\vec{v} + \vec{w} \notin W$, which is equivalent to say that $\vec{v} + \vec{w} \notin (V \cup W)$.

5 Problem 1.5

Let V be a vector space of dimension n and let $x_1, \dots, x_n \in V$. Show that:

- (a) If x_1, \dots, x_n are linearly independent, then (x_1, \dots, x_n) is a basis of V .

We will first prove a lemma: If $(x_1, \dots, x_n) \in V$ are linearly independent. Then if $\vec{x} \notin \text{Span}(v_1, \dots, v_n)$, we will have that (v_1, \dots, v_n, x) are linearly independent.

Proof. We prove by contradiction. If (v_1, \dots, v_k, x) are linearly dependent, there exists $(\beta_1, \dots, \beta_{k+1}) \neq \vec{0}$ such that $\beta_1 v_1 + \beta_2 v_2 + \dots + \beta_k v_k + \beta_{k+1} x = 0$ holds.

If $\beta_{k+1} \neq 0$, $x = \frac{\beta_1}{\beta_{k+1}} v_1 + \frac{\beta_2}{\beta_{k+1}} v_2 + \dots + \frac{\beta_k}{\beta_{k+1}} v_k \in \text{Span}(v_1, \dots, v_k)$, which contradicts the assumption that $x \notin \text{Span}(v_1, \dots, v_n)$.

If $\exists \beta_i \neq 0, i = 1, 2, \dots, k$ and $\beta_{k+1} = 0$, $v_i = \frac{\beta_1}{\beta_i} v_1 + \dots + \frac{\beta_{i-1}}{\beta_i} v_{i-1} + \dots + \frac{\beta_{i+1}}{\beta_i} v_{i+1} + \dots + \frac{\beta_k}{\beta_i} v_k$, which contradicts the fact that v_i 's are linearly independent. To sum up (v_1, \dots, v_k, x) are linearly independent. \square

Then we prove our main theorem:

Proof. We prove by contradiction, assume that (x_1, \dots, x_n) is not a basis for V . Then by our criterion of basis, since (x_1, \dots, x_n) are linearly independent, we must have that $\text{Span}(x_1, \dots, x_n) \neq V$. This means $\vec{v} \in V$, s.t. $\vec{v} \notin \text{Span}(x_1, \dots, x_n)$. Then by our lemma, (x_1, \dots, x_n, v) are linearly independent. But we know from lecture that any linear independent list of vectors of V with dimension n should have length $k \leq n$, but (x_1, \dots, x_n, v) has length $n + 1$, which is a contradiction. \square

- (b) If $\text{Span}(x_1, \dots, x_n) = V$, then (x_1, \dots, x_n) is a basis of V .

We will first prove a lemma: Let $x_1, \dots, x_k \in \mathbb{R}^n$. If $x_1 \in \text{Span}(x_2, \dots, x_k)$, then $\text{Span}(x_1, \dots, x_k) = \text{Span}(x_2, \dots, x_k)$.

Proof. We prove by definition of set equivalence.

1. $\forall x \in \text{Span}(x_2, \dots, x_k)$, we have $x = \alpha_2 x_2 + \dots + \alpha_k x_k = 0x_1 + \alpha_2 x_2 + \dots + \alpha_k x_k \in \text{Span}(x_1, \dots, x_k)$. So $\text{Span}(x_2, \dots, x_k) \subseteq \text{Span}(x_1, \dots, x_k)$.

2. $\forall x \in \text{Span}(x_1, \dots, x_k)$, we have $x = \alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_k x_k$. Since $x_1 \in \text{Span}(x_2, \dots, x_k)$, we have $x_1 = \beta_2 x_2 + \dots + \beta_n x_n$, fitting into the equation above we have $x = (\alpha_2 + \alpha_1 \beta_2) x_2 + \dots + (\alpha_k + \alpha_1 \beta_k) x_k \in \text{Span}(x_2, \dots, x_k) \subseteq \text{Span}(x_1, \dots, x_k)$, thus $\text{Span}(x_1, \dots, x_k) \subseteq \text{Span}(x_2, \dots, x_k)$.

To sum up, we have $\text{Span}(x_1, \dots, x_k) = \text{Span}(x_2, \dots, x_k)$. \square

Then we prove our main theorem:

Proof. We prove by contradiction. Assume that (x_1, \dots, x_n) is not a basis of V , then by definition of a basis, (x_1, \dots, x_n) must be linearly dependent. By symmetry, we can assume that $x_i \in \text{Span}((x_j)_{j \neq i})$. Then by our lemma we have $\text{Span}(x_1, \dots, x_n) = \text{Span}((x_j)_{j \neq i}) = V$. However we know from lecture that any spanning list of V with dimension n should have length of at least n , but $\text{Span}((x_j)_{j \neq i})$ has length of just $n - 1$, which is a contradiction. \square