

Lecture 15 : Subspaces, Bases, and Linear Independence

Math 3013
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Agenda

- ▶ Subspaces
- ▶ Coordinatization of Subspaces and Bases
- ▶ Linear Independence
- ▶ The Row Space of a Matrix

Recap

Definition

A **subspace** of \mathbb{R}^n is a subset $S \subset \mathbb{R}^n$ such that

- ▶ whenever $\mathbf{v} \in S$ and $\lambda \in \mathbb{R}$, $\lambda\mathbf{v} \in S$
- ▶ whenever $\mathbf{v}_1, \mathbf{v}_2 \in S$, $\mathbf{v}_1 + \mathbf{v}_2 \in S$

Two Fundamental Ways Subspaces Arise

- ▶ The **span** of a set of vectors:

$$\begin{aligned} S &= \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_k) \\ &\equiv \{c_1\mathbf{v}_1 + \dots + c_k\mathbf{v}_k \mid c_1, \dots, c_k \in \mathbb{R}\} \end{aligned}$$

Geometrically, this is a hyperplane that passes through the origin $\mathbf{0}$.

- ▶ The **solution set of a homogeneous linear system** $\mathbf{Ax} = \mathbf{0}$

$$a_{11}x_1 + \dots + a_{1m}x_m = 0$$

$$\vdots$$

$$a_{n1}x_1 + \dots + a_{nm}x_m = 0$$

Problem: Finding good coordinates for the vectors lying in a subspace of \mathbb{R}^n

Basic idea: If $W = \text{span}(\mathbf{w}_1, \dots, \mathbf{w}_k)$, use coefficients of elements of W as coordinates

$$\begin{aligned}\mathbf{v} \in W &\implies \mathbf{v} = c_1 \mathbf{w}_1 + \dots + c_k \mathbf{w}_k \\ &\implies \text{use } [c_1, \dots, c_k] \text{ as coordinates for } \mathbf{v} \text{ in } W\end{aligned}$$

Problem: without some restrictions on the generators $\mathbf{w}_1, \dots, \mathbf{w}_k$ such “coordinates” do not specify the vectors in W uniquely.

(Counter-)Example

Consider

$$S = \text{span} \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right)$$

For each choice of numbers c_1, c_2, c_3 ,

$$\mathbf{v}_{c_1, c_2, c_3} = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ -1 \end{bmatrix} + c_3 \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

belongs to S .

Example, Cont'd

But the numbers c_1, c_2, c_3 do not provide good coordinates for the vectors in S since two different choices of c_1, c_2, c_3 can correspond to the same vector. E.g.,

$$\left. \begin{array}{l} c_1 = 1 \\ c_2 = 1 \\ c_3 = 0 \end{array} \right\} \Rightarrow (1) \begin{bmatrix} 1 \\ 1 \end{bmatrix} + (1) \begin{bmatrix} 1 \\ -1 \end{bmatrix} + (0) \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

$$\left. \begin{array}{l} c_1 = 0 \\ c_2 = 0 \\ c_3 = 2 \end{array} \right\} \Rightarrow (0) \begin{bmatrix} 1 \\ 1 \end{bmatrix} + (0) \begin{bmatrix} 1 \\ -1 \end{bmatrix} + (2) \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

Bases

Defining our way out of the problem:

Definition

A **basis** for a subspace S is a set of vectors $\{\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_k\}$ such that **every** vector $\mathbf{v} \in S$ can be **uniquely** expressed as

$$\mathbf{v} = c_1 \mathbf{b}_1 + c_2 \mathbf{b}_2 + \cdots + c_k \mathbf{b}_k$$

(i.e., the above equation is true for exactly one choice of coefficients c_1, c_2, \dots, c_k)

Thus, an element $\mathbf{v} = c_1 \mathbf{b}_1 + \cdots + c_k \mathbf{b}_k \in S$ is uniquely identified by its “coordinate vector” $[c_1, \dots, c_k]$ **if and only if** $\{\mathbf{b}_1, \dots, \mathbf{b}_k\}$ is a basis for S .

Identifying Bases for Subspaces

Definition

A set of vectors $\{\mathbf{w}_1, \dots, \mathbf{w}_k\}$ is **linearly independent** if the only solution of

$$x_1 \mathbf{w}_1 + \dots + x_k \mathbf{w}_k = \mathbf{0}$$

is

$$x_1 = 0, x_2 = 0, \dots, x_k = 0$$

Linear Independence and Bases

Theorem

Suppose $W = \text{span}(\mathbf{w}_1, \dots, \mathbf{w}_k)$. Then $\{\mathbf{w}_1, \dots, \mathbf{w}_k\}$ is a basis for W **if and only if** the vectors $\mathbf{w}_1, \dots, \mathbf{w}_k$ are linearly independent.

Proof.

\implies : To show : If $B = \{\mathbf{w}_1, \dots, \mathbf{w}_k\}$ is a basis for W , then $\{\mathbf{w}_1, \dots, \mathbf{w}_k\}$ are linearly independent.

Since B is a basis for W , every vector in W has a unique expression as a linear combination of $\mathbf{w}_1, \dots, \mathbf{w}_k$.

In particular, since $\mathbf{0} \in W$,

$$\mathbf{0} = c_1\mathbf{w}_1 + \dots + c_k\mathbf{w}_k$$

for exactly one choice of coefficients c_1, \dots, c_k . But clearly, the choice $c_1 = 0, c_2 = 0, \dots, c_k = 0$ works.

Thus,

$$x_1\mathbf{w}_1 + \dots + x_k\mathbf{w}_k = \mathbf{0} \implies x_1 = 0, x_2 = 0, \dots, x_k = 0$$

and so $\mathbf{w}_1, \dots, \mathbf{w}_k$ are linear independent vectors.

Proof of Theorem, Cont'd

\Leftarrow :

To show : If $\mathbf{w}_1, \dots, \mathbf{w}_k$ are linearly independent, then $\{\mathbf{w}_1, \dots, \mathbf{w}_k\}$ is a basis for $W = \text{span}(\mathbf{w}_1, \dots, \mathbf{w}_k)$.

Since $W = \text{span}(\mathbf{w}_1, \dots, \mathbf{w}_k)$, every vector $\mathbf{v} \in W$ is a linear combination of $\mathbf{w}_1, \dots, \mathbf{w}_k$. We need to show that there is only one way of writing a given $\mathbf{v} \in W$ as a linear combination of $\mathbf{w}_1, \dots, \mathbf{w}_k$. Suppose there were two ways:

$$\mathbf{v} = c_1 \mathbf{w}_1 + \dots + c_k \mathbf{w}_k$$

$$\mathbf{v} = d_1 \mathbf{w}_1 + \dots + d_k \mathbf{w}_k$$

Subtracting the second equation from the first yields

$$\mathbf{0} = (c_1 - d_1) \mathbf{w}_1 + \dots + (c_k - d_k) \mathbf{w}_k \quad (*)$$

Since the vectors $\mathbf{w}_1, \dots, \mathbf{w}_k$ are linearly independent,

$$\mathbf{0} = x_1 \mathbf{w}_1 + \dots + x_k \mathbf{w}_k \implies x_1 = 0, \dots, x_k = 0$$

Thus, (*) requires

$$c_1 - d_1 = 0, \dots, c_k - d_k = 0 \implies c_1 = d_1, \dots, c_k = d_k$$

Upshot:

To find a basis for a subspace W , we need to find a set of linearly independent vectors that generate W .

Put another way,

$\{\mathbf{b}_1, \dots, \mathbf{b}_k\}$ is a basis for W **if and only if**

$$(i) \quad W = \text{span}(\mathbf{b}_1, \dots, \mathbf{b}_k)$$

$$(ii) \quad \mathbf{0} = c_1\mathbf{b}_1 + \dots + c_k\mathbf{b}_k \quad \Leftrightarrow \quad c_1 = 0, \dots, c_k = 0$$

Nomenclature

Whenever one has an equation of the form

$$c_1 \mathbf{v}_1 + \cdots + c_k \mathbf{v}_k = \mathbf{0} \quad (1)$$

the solution $c_1 = 0, \dots, c_k = 0$ of (1) is called the **trivial solution** of (1).

If there are other (non-trivial) solutions of (1), then (1) is called a **dependence relation**.

A set of vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ are **linearly dependent** if there exists a dependence relation for them.

Here is tale-tell sign of a set of linearly independent vectors.

Suppose each of the vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ has non-zero component that cannot be cancelled out by a linear combination of the other vectors. Then the vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ are linearly independent.

Example: Consider the vectors $\mathbf{v}_1 = [1, 1, 0, 1]$, $\mathbf{v}_2 = [0, 1, 1, 0]$. A dependence relation between these vectors would be an equation of the form

$$[0, 0, 0, 0] = x_1[1, 1, 0, 1] + x_2[0, 1, 1, 0] = [x_1, x_1 + x_2, x_2, x_1]$$

We must have $x_1 = 0$ because the first component of \mathbf{v}_1 cannot be cancelled by a scalar multiple of \mathbf{v}_2 and we must have $x_2 = 0$ since the third component of \mathbf{v}_2 cannot be cancelled by a scalar multiple of \mathbf{v}_1 .

The Row Vectors of a Matrix in Row Echelon Form

The preceding observation implies

Theorem

Suppose \mathbf{A} is an $n \times m$ matrix in Row Echelon Form. Then the non-zero row vectors of \mathbf{A} are linearly independent.

Proof. If let $\text{Row}_1(\mathbf{A}), \dots, \text{Row}_k(\mathbf{A})$ be the non-zero rows of \mathbf{A} . Consider

$$x_1 \text{Row}_1(\mathbf{A}) + x_2 \text{Row}_2(\mathbf{A}) + \dots + x_k \text{Row}_k(\mathbf{A}) = \mathbf{0} \quad (*)$$

Since \mathbf{A} is in R.E.F., below the pivot in the first row we'll have nothing but 0's; and so we can not cancel out the pivot component of $\text{Row}_1(\mathbf{A})$ using a linear combinations with the other row vectors. Hence, we cannot have (*) without $x_1 = 0$. But once the contribution of $\text{Row}_1(\mathbf{A})$ has been removed from (*), the same argument implies that

$$x_2 \text{Row}_2(\mathbf{A}) + \dots + x_k \text{Row}_k(\mathbf{A}) = \mathbf{0} \implies x_2 = 0$$

(since there is no way to cancel the pivot component of $\mathbf{R}_2(\mathbf{A})$)

The Row Vectors of a Matrix in R.E.F., Cont'd

Thus,

$$\mathbf{0} = x_1 \text{Row}_1(\mathbf{A}) + x_2 \text{Row}_2(\mathbf{A}) + \cdots + x_k \text{Row}_k(\mathbf{A})$$

$$\implies x_1 = 0$$

$$\implies \mathbf{0} = x_2 \text{Row}_2(\mathbf{A}) + \cdots + x_k \text{Row}_k(\mathbf{A}) \implies x_2 = 0$$

$$\vdots$$

$$\mathbf{0} = x_k \text{Row}_k(\mathbf{A}) \implies x_k = 0$$

Thus, (*) can only hold when $x_1 = 0, \dots, x_k = 0$. Hence, the non-zero row vectors of \mathbf{A} are linearly independent. \square .

Finding a Basis for $\text{RowSp}(\mathbf{A})$

Definition

The **Row Space** of an $n \times m$ matrix is the subspace of \mathbb{R}^m generated by the row vectors of \mathbf{A} .

$$\text{RowSp}(\mathbf{A}) = \{c_1 \text{Row}_1(\mathbf{A}) + \cdots + x_n \text{Row}_n(\mathbf{A}) \mid x_1, \dots, x_n \in \mathbb{R}\}$$

Recall

Criteria for a Basis: $\{\mathbf{b}_1, \dots, \mathbf{b}_k\}$ is a basis for a subspace S if

- ▶ $S = \text{span}(\mathbf{b}_1, \dots, \mathbf{b}_k)$
- ▶ $\{\mathbf{b}_1, \dots, \mathbf{b}_k\}$ are linearly independent.

Corollary

If a matrix \mathbf{A} is in R.E.F., its non-zero row vectors provide a basis for $\text{RowSp}(\mathbf{A})$.

What about Matrices that are not in R.E.F.?

Lemma

Suppose \mathcal{R} is an elementary row operation. Then

$$\text{RowSp}(\mathcal{R}(\mathbf{A})) = \text{RowSp}(\mathbf{A})$$

Corollary

Suppose \mathbf{A} is an $n \times m$ matrix and let $\text{R.E.F.}(\mathbf{A})$ be its row echelon form. Then the non-zero rows of $\text{R.E.F.}(\mathbf{A})$ form a basis for $\text{RowSp}(\mathbf{A})$.

Finding a basis for $W = \text{span}(\mathbf{w}_1, \dots, \mathbf{w}_k)$ when $\mathbf{w}_1, \dots, \mathbf{w}_k$ are not linearly independent

Procedure:

- ▶ Write the vectors $\mathbf{w}_1, \dots, \mathbf{w}_k$ as the rows of a matrix \mathbf{A}
- ▶ Row reduce \mathbf{A} to a matrix \mathbf{A}' in R.E.F.
- ▶ The non-zero rows of \mathbf{A}' will be a basis for $W = \text{span}(\mathbf{w}_1, \dots, \mathbf{w}_k)$

This works because

- ▶ Elementary row operations do not change the row space of a matrix
- ▶ The non-zero rows of matrix in R.E.F. are linearly independent and generate its row space.

Example

Consider

$$W = \text{span}([1, 0, 1, 1], [1, 1, -1, 0], [0, -1, 2, 1])$$

Find a basis for W .

- Form a matrix \mathbf{A} using the given vectors as rows

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

Note

$$W = \text{span}([1, 0, 1, 1], [1, 1, -1, 0], [0, -1, 2, 1]) = \text{RowSp}(\mathbf{A})$$

- Row reduce \mathbf{A} to a matrix \mathbf{A}' in R.E.F.

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

Example, Cont'd

- ▶ The non-zero rows of \mathbf{A}' will be a basis for $\text{RowSp}(\mathbf{A}) = W$.
Thus, a suitable basis for W will be

$$B = \{[1, 0, 1, 1], [0, 1, -2, -1]\}$$