Linear Algebra

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Chapter 1

Spring 1

Lecture 1: Introduction

Lecture 7

1.1 Basis and dimension theory

Keep thinking back to \mathbb{R}^n Yo

Definition 1.1 (Basis of V). Let V be a vector space. A subset $B \subseteq V$ is a basis of v if B is:

- linearly independent
- \bullet and spans V

Example. Let $V = M_{m,n}$. This is a vector space. The standard basis is the set

$${E_{ij} \mid 1 \le i \le m, 1 \le j, +n}$$
.

where E_{ij} is the matrix with 0's everywhere except in $(i,j)^{th}$ entry where it equals 1.

For example, in the case M_{22} we have

$$\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$

Linear independece is clear. We can show it spans M_{22}

$$\text{Span: } a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + d \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \in M_{22}$$

Example. Let $V = \mathbb{R}^n$. The standard basis $\{e_1, e_2, \dots, e_n\}$.

Example. Let $V \leq \mathbb{R}^5$ be the space of vectors (x_1, \ldots, x_5) such that

$$x_1 + x_2 - x_3 + x_5 = 0$$

$$x_1 + 2x_2 + x_4 + 3x_5 = 0$$

$$x_2 + x_3 + x_4 + 2x_5 = 0.$$

We write this system as

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

Placing this in reduced row echelon form gives:

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

Assign parameters α, β, γ to x_3, x_4, x_5 :

$$(x_1, x_2, x_3, x_4, x_5) = (2\alpha + \beta + \gamma, -\alpha - \beta - 2\gamma, \alpha, \beta, \gamma)$$

= $\alpha (2, -1, 1, 0, 0) + \beta (1, -1, 0, 1, 0) + \gamma (1, -2, 0, 0, 1)$

Hence a basis for V is $\{(2, -1, 1, 0, 0), (1, -1, 0, 1, 0), (1, -2, 0, 0, 1)\}$

Proposition 1.2. Let V be a finite-dimensional vector space, $V \neq \{0_v\}$. Suppose $S = \{v_1, \ldots, v_s \text{ is a spanning set of } V$. Moreover, if $L \subseteq S$ in any linearly independent set then we can choose B to contain L.

Proof. We only prove the final statement, since the rest follows by setting $L = \emptyset$. Let $B \subseteq S$ be a maximal subset containing L, such that B is linearly independent. We claim B is a basis. We need to show $\langle B \rangle = V$. Suppose not. Hence $\exists v_i \notin \langle B \rangle$. But then $B \cup \{v_i\}$ is a strictly larger linearly independent set.

To show $B \cup \{v_i\}$ is linearly independent:

$$\lambda v_i + \sum \mu_j b_j = 0_v$$

with $b_j \in B$ and $\lambda, \mu_j \in \mathbb{R}$ is a dependence relation.

If $\lambda=0$ this contradicts B being linearly independent. Hence $\lambda\neq 0$. So

$$v_i = \frac{1}{\lambda} \sum \mu_j b_j$$

But this would mean that $v_i \in \langle b \rangle$, again resulting in a contradiction \square

Lecture 8

Remark. Says that every finite-dimensional vector space has a basis.

Lemma 1.3 (Steinitz Exchange Lemma). Let V be a vector space. Let $X = \{v_1, ..., v_N\} \subseteq V$. Suppose $u \in \langle X \rangle$, but $u \notin \langle X \setminus \{v_i\} \rangle$ fro some $1 \le i \le n$. Then let $Y = (X \setminus \{v_i\}) \cup \{u\}$. We have $\langle Y \rangle = \langle X \rangle$.

Proof. • Since $u \in \langle X \rangle \exists \lambda_1, \dots, \lambda_n \in \mathbb{R}$ such that

$$u = \lambda_1 v_1 + \dots +_n v_n \tag{1.1}$$

By assumtion we have $u \notin \langle X \{v_i\} \rangle$. Withought loss of generality assume i = n. Since $u \notin \langle X \{v_i\} \rangle$, so $n \neq 0$. Rearranging (1) we get

$$v_n = \frac{1}{\lambda_n} (u - \lambda_1 v_1 - \dots - \lambda_{n-1} v_{n-1})$$
 (1.2)

- Let $w \in \langle Y \rangle$. We can express w as a linear combination of v_1, \ldots, v_{n-1}, u . We can replace u with (1), hence write w as a linear combination of $v_1, \ldots, v_{n-1}, v_n$. Hence $w \in \langle X \rangle$. So $\langle Y \rangle \subseteq \langle X \rangle$.
- Let $w \in \langle X \rangle$. Then we have an expression for w as a linear combination of v_1, \ldots, v_n . By (2) we can replace v_n with a linear combination of v_1, \ldots, v_{n-1}, k . Hence $w \in \langle Y \rangle$. So $\langle X \rangle \subseteq \langle Y \rangle$.

Example. Let $V = \mathbb{R}^3$. $B = \{e_1, e_2, e_3\}$ the standard basis. Let $u = 2e_1 - 3e_2$. The Steinitz Exchange Lemma tells us

$$B_1 = \{u, e_2, e_3\}$$
 $B_2 + \{e_1, u, e_3\}$

both span V. (You should check this later!)

Although the Steinitz Exchange Lemma says nothing about linear independence, infact both B_1, B_2 are. Hence B_1, B_2 are bases for V. (Again check this!)

What abut if we exchange e_3 for u?

i.e.
$$\{e_1, e_2, u\} = \{(1, 0, 0), (0, 1, 0), (2, -3, 0)\}$$

Since the third coordinate is zero, this cannot span V.

Theorem 1.4. Let V be a vector space and $S,T\subseteq V$ are finite subsets. If S is linearly independent and if $\langle T\rangle=V$, then $\mid S\mid\leq\mid T\mid$.

Proof. • Let $S = \{u_1, \dots, u_m\}, T = \{v_1, \dots, v_n\}$

- We will use Steinitz Exchange Lemma to swap elements of T for those in S. This will finish exhausting S, hence $|S| \leq |T|$.
- Let $T_0 = \{v_1, \ldots, v_n\}$. We have $\langle T_0 \rangle = V_1$, so $u_1 \in \langle v_1, \ldots, v_i \rangle$ but $u_1 \notin \langle v_1, \ldots, v_{i-1} \rangle$. By the Steinitz Exchange Lemma,

$$\langle v_1, \dots, v_i \rangle = \langle u_1, v_1, \dots, v_{i-1} \rangle.$$

Hence $V = \langle u_1, v_1, \dots, v_{i-1}, v_{i+1}, \dots, v_n \rangle$ withought loss of generality we can relable the v_i to get u_1 has been exchanged for v_1 e set

$$T_1 = \{u_1, v_2, \dots, v_n\}, \qquad \langle T_1 \rangle = V.$$

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• Proceeding by induction we have

$$T_k = \{u_1, \dots, u_k, v_{k+1}, \dots, v_n\}, \qquad \langle T_k \rangle = V$$

- At each stage $u_{k+1} \in \langle T_k \rangle$, but $u_{k+1} \notin \langle u_1, \dots, u_k \rangle$ because S is linearly independent. Hence we continue to make substitution. This can only terminate when S is exhausted, $m \leq n$.
- Hence $|S| \leq |T|$.

Lecture 9

Let B_2 be another basis of V. Recall that this means both B_1 and B_2 are spanning and are linearly independent.

$$\begin{array}{c} B_2 \text{ is linearly independent} \\ \langle B_1 \rangle = V \end{array} \right\} \Rightarrow \mid B_2 \mid \leq \mid B_1 \mid \\ B_1 \text{ is linearly independent} \\ \langle B_2 \rangle = V \end{array} \right\} \Rightarrow \mid B_1 \mid \leq \mid B_2 \mid \\ \Rightarrow \mid B_1 \mid = \mid B_2 \mid$$

In particular, B_2 is also finite since B_1 is finite.

Definition 1.5 (Dimension of V). Let V be a finite-dimensional vector space. The dimension of V is defined to be $\dim(V) := |B|$ for some basis B of V

 $S := \{(0,1,2,3), (1,2,3,4), (2,3,4,5)\}$

Example.

$$= \{r_1, r_2, r_3\} \text{ where } r_1, r_2, r_3 \subseteq \mathbb{R}^4$$

$$A = \begin{pmatrix} 0 & 1 & 2 & 3 \\ 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \end{pmatrix} \in M_{34}$$
Then the Rowspan $(A) = \langle S \rangle \subseteq \mathbb{R}^4$

$$= \langle r_1, r_2, r_3 \rangle.$$

$$RRE (A) = \begin{bmatrix} -1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & -2 & 1 \end{bmatrix} A = \begin{bmatrix} 1 & 0 & -1 & -2 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} r'_1 \\ r'_2 \\ 0 \end{bmatrix}$$

$$r'_1 = (1, 0, -1, -2) = -r_1 - r_2 + r_3 \\ r'_2 = (0, 1, 2, 3) = r_1$$

$$\Rightarrow \langle S' \rangle \subseteq \langle S \rangle$$

$$\langle r'_1, r'_2 \rangle \subseteq \langle r_1, r_2, r_3 \rangle$$

$$\begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} = A = \begin{bmatrix} -1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & -2 & 1 \end{bmatrix}^{-1} RRE(A)$$

$$= \begin{bmatrix} 0 & 1 & 0 \\ 1 & 2 & -1 \\ 2 & 3 & -1 \end{bmatrix} \begin{bmatrix} r'_1 \\ r'_2 \\ 0 \end{bmatrix}$$

$$\begin{matrix} r_1 = r'_2 \\ r_2 = r'_1 + 2r'_2 \\ r_3 = 2r'_1 + 3r'_2 \end{matrix} \right\} \Rightarrow \langle S \rangle \leq \langle S' \rangle$$

$$\Rightarrow \langle S \rangle = \langle S' \rangle$$

Notice that $\left\{r_{1}^{'},r_{2}^{'}\right\}$ is spanning $\langle S\rangle.$ Notice that if

$$\lambda_1 r_1' + \lambda_2 r_2' = \underline{0} = (\lambda_1, \lambda_2, -\lambda_1 + 2\lambda_2, -2\lambda_1 + 3\lambda_2) = \underline{0}$$

$$\Rightarrow \lambda_1 - \lambda_2 = 0$$

So $\left\{r_{1}^{'}, r_{2}^{'}\right\}$ is linearly independent. That is, $\left\{r_{1}^{'}, r_{2}^{'}\right\} = S'$ is a basis of Rowspan(A). Hence, dim Rowspan(A) = 2.

Example (Generalisation). Let $A \in M_{nn}$

$$RRE(A) = BA = \begin{bmatrix} r'_1 \\ \vdots \\ r'_k \\ 0 \\ \vdots \\ 0 \end{bmatrix} \text{ with } r'_1 \neq \underline{0}, \dots, r'_k \neq \underline{0}$$

Like above we have $\operatorname{Rowspan}(A) = \langle r_1^{'}, \dots, r_k^{'} \rangle$

$$\begin{array}{ll} \text{row rank of } A &=& \dim \, \operatorname{Rowspan} \, (A) \\ &=& \operatorname{non-zero} \, \operatorname{rows} \, \operatorname{in} \, \operatorname{RRE} \, (A) \\ &=& k \end{array}$$

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1.2 Subspaces and Dimension Theory

Proposition 1.6. Let V be a finite-dimensional vector space and $U \leq V$ is a subspace. Then

- \bullet U is finite-dimensional
- dim $U \leq \dim V$

Furthermore, if dim $U = \dim V$, then U = V.

Proof. Let $S \subseteq U$ be the largest linearly independent subset. Let B be a basis of V, say B with $|B| = n = \dim(V)$

$$S\subseteq V \text{ is linearly independent.} \quad \left. \left. \right\} \Rightarrow |S| \leq |B| \right.$$

Assume towards a contradiction that $\langle S \rangle \not\subseteq U$. Then there exists $u \in U \langle S \rangle$

<u>Claim:</u> $S \cup \{u\}$ is linearly independent.

Proof. Say $S = \{s_1, ..., s_k\}.$

Let $\lambda_1, \ldots, \lambda_k, \lambda \in \mathbb{R}$ such that $\lambda_1 s_1 + \ldots + \lambda_k s_k + \lambda u = \underline{0}$. There are two cases:

• $\lambda = 0$ $\Rightarrow u = \frac{-1}{\lambda} (\lambda_1 s_1 + \ldots + \lambda_k s_k) \in \langle S \rangle$

which is a contradiction.

• $\lambda \neq 0$

$$\Rightarrow \underline{0} = \lambda_1 s_1 + \ldots + \lambda_k s_k + \lambda_u \text{ where } \lambda_u = 0$$
$$= \lambda_1 s_1 + \ldots + \lambda_k s_k$$

Since $s_{1,k}$ are linearly independent so $\lambda_1 = \ldots = \lambda_k = 0$ That is all coefficients are zero.

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Hence, $S \not\subseteq S \cup \{u\}$ is linearly independent. This is a contradiction to S being maximal, which implies $U = \langle S \rangle$ and S is a basis of U, which implies $\dim(U) = |S| \leq |B| = n = \dim(V)$

Lecture 10

Theorem 1.7 (Dimension Formula). Let V be a finite-dimensional vector space. Let $U,W \leq V$ be subspaces. Then

$$\dim (U + W) + \dim (U \cap W) = \dim (U) + \dim (W)$$

Example. Suppose $V = \mathbb{R}_3$, $U = \langle (1,0,0) \rangle$, $W = \langle (1,0,1) \rangle$. Have $\dim(U) = 1$, $\dim W = 1$.

$$U \cap W = \langle (1,0,0) \rangle \cap \langle (1,0,1) \rangle = \{\underline{0}\}$$

Hence $\dim(U \cap W = 0)$. By the dimension formula

$$\dim (U + W) = \dim (U) + \dim (W) - \dim (U \cap W)$$
$$= 1 + 1 - 0$$
$$= 2$$

 \Diamond

Proof. Let $\{v_1, \ldots, v_m\}$ be a basis for $U \cap W$. Have $U \cap W$. Have $UW \leq U$, $U \cap W \leq W$. We can extend our basis to give a basis

$$\{v_1, \ldots, v_m, u_1, \ldots, u_p\}$$
 of U

$$\{v_1, \dots, v_m, w_1, \dots, w_q\}$$
 of W

Hence $\dim(U) = m + p$, $\dim(W) = m + q$

$$\dim (U \cap W) = m$$

We claim $S = \{v_1, \dots, v_m, u_1, \dots, u_p, w_1, \dots, w_q\}$ is a basis for U + W. Hence $\dim(U + W) = m + p + q$ as required.

Spanning:

 $\overline{\text{Take } x \in U} + W$. Then x = u+w, for some $u \in U$, $w \in W$.

$$u = a_1 v_1 + \ldots + a_m v_m + a'_1 u_1 + \ldots + a'_p u_p$$

$$w = b_1 v_1 + \ldots + b_m v_m + b'_1 w_1 + \ldots + b'_a w_a$$

for some $a_i, a'_i, b_i, b'_i \in \mathbb{R}$. Then

$$x = u + w$$

$$= (a_1 + b_1) v_1 + \ldots + (a_m + b_m) v_m + a'_1 u_1 + \ldots + a'_n u_p + b'_1 w_1 + \ldots + b'_a w_1 + \ldots + b'_a w_q$$

Hence $x \in \langle S \rangle$.

Linear Independence:

Suppose

$$a_1v_1 + \ldots + a_mv_m + b_1u_1 + \ldots + b_pu_p + c_1w_1 + \ldots + c_qw_q = 0$$

for some $a_i, b_i, c_i \in \mathbb{R}$. Have

$$(a_1v_1 + \ldots + a_mv_m + b_1u_1 + \ldots + b_pu_p) \in U = -(c_1w_1 + \ldots + c_qw_q) \in W$$

so this lies in $U \cap W$. Since $\{v_1, \ldots, v_m, w_1, \ldots, w_q\}$ is a basis of W, so

$$c_1 = \ldots = c_q = d_1 = \ldots = d_m = 0.$$

Hence we have:

$$a_1v_1 + \ldots + a_mv_m + b_1u_1 + \ldots + b_pu_p = 0$$

Since $\{v_1, \ldots, v_m, u_1, \ldots, u_p\}$ is a basis of U, so

$$a_1 = \ldots = a_m = b_1 = \ldots = b_p = 0.$$

Hence it is linearly independent.

Definition 1.8 (Direct Sum). Let V be a vector space $U, W \leq V$. If $U \cap W = \{\underline{0}\}$ and U + W = V. We write $V = U \bigoplus W$ and we say V is the direct sum of U and W.