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Chapter 1

Matrices

1.0.1 Gaussian elimination

We can define three elementary operations that do not change the solution of the system.

Definition (Elementary row operations): Consider a matrix $A \in \mathcal{M}_{m \times n}(\mathbb{F})$, we define three elementary row operations on A:

- 1. multiplication of one row by a non-zero scalar c, denoted by $e_r(c)$.
- 2. replacement of $r_{\rm th}$ row of A by row r plus c times row $s \neq r$ that is $r_{\rm new} = r + cs$ and it is denoted by $e_{rs}(c)$
- 3. Interchange two rows e_{rs} .

Theorem 1.1. To each elementary row operation e there corresponds an elementary row operation e^{-1} , of the same type as e, such that $e^{-1}(e(A)) = e(e^{-1}(A)) = A$ for any A. Furthermore, each elementary operation has a $n \times n$ matrix representation.

Proof. It is easily verified by hand that the inverses exist and are of the same kind.

At each step of Gaussian elimination, one takes an element to be the **pivot** and eliminate the element from the other equations. At the end, the system has been reduced to a **upper triangle** (maybe with permuation). For **Gauss-Jordan method** the pivot needs to be one and the element must be eliminate from all equations.

1.0.2 LU decomposition

With some assumption one can transform matrix A to an upper triangular matrix U via elementary row operations. That is, there exists an elementary matrix E such that EA = U. It can be verified that E is lower triangular matrix and hence $L = E^{-1}$ is lower triangular and A = LU. The assumptions made can be simplified if we allow permuations. That is

Theorem 1.2. There exists a permuation of a matrix A that has a LU decomposition. In other words, there exists a permuation matrix P such that

$$PA = LU$$

Theorem 1.3. If A is strongly invertible matrix that is, every submatrix A(1:i,1:i) is invertible (leading minors are non-zero) then there exists a unique lower unitriangular matrix L and a unique upper triangular matrix U such that A = LU.

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Theorem 1.4. If the decomposition A = LU exist where L is a lower unitriangular matrix, then the decomposition A = LDU exists where D is a diagonal matrix and L and U are lower and upper unitriangular matrices respectively. Furthermore, if such LDU decomposition exists and $A = A^T$ then $A = LDL^T$

Chapter 2

Vector Spaces

A vector space, also called linear space, consist of the followings:

- 1. a field \mathbb{F} of scalars.
- 2. a set V of vectors.
- 3. a vector addition +, with the following properties:
 - (a) V is closed under addition.
 - (b) addition is commutative.
 - (c) addition is associative.
 - (d) addition has a unique identity element 0.
 - (e) for each vector $\alpha \in V$, $\exists \beta \in V$ s.t. a + b = 0.
- 4. a scalar multiplication with the following properties:
 - (a) V is closed under scalar multiplication.
 - (b) $(c_1c_2)\alpha = c_1(c_2\alpha)$.
 - (c) $c(\alpha + \beta) = c\alpha + c\beta$.
 - (d) $(c_1 + c_2)\alpha = c_1\alpha + c_2\alpha$
 - (e) scalar multiplication has a unique identity element 1.

Example 2.1. $V = \mathbb{R}$ and $\mathbb{F} = \mathbb{R}$ is a vector space. Furthermore, $V = \mathbb{R}^n$ over $\mathbb{F} = \mathbb{R}$ is a vector space with the scalar multiplication $c(x_1, x_2, \dots, x_n) = (cx_1, cx_2, \dots, cx_n)$.

Example 2.2. $V = \mathbb{R}$ and $\mathbb{F} = \mathbb{Z}$ is not a vector space as \mathbb{Z} is not a field.

Definition: A vector $\beta \in V$ is said to be a **linear combination** of the vectors $\alpha_1, \alpha_2, \dots, \alpha_n$ if there exists scalars $c_1, c_2, \dots, c_n \in \mathbb{F}$ such that:

$$\beta = c_1 \alpha_1 + c_2 \alpha_2 + \dots + c_n \alpha_n = \sum_{i=1}^n c_i \alpha_i$$

2.1 Subspaces

Let V be a vector space over field \mathbb{F} . A susbspace of V is a subset W of V which is itself a vector space over \mathbb{F} with the operations of vector addition and scalar multiplication on V.

Theorem 2.1. A non-empty subset W of V is a subspace of V if and only if for each pair of vectors $\alpha, \beta \in W$ and each scalar $c \in \mathbb{F}$ the vector $c\alpha + \beta \in W$.

Proof. Necessity: Suppose W is a non-empty subset of V with the above property. Since W is not empty then there exists a vector $\alpha \in W$ and therefore, $(-1)\alpha + \alpha = 0 \in W$. For each $c \in \mathbb{F}$, $c\alpha + 0 = c\alpha \in W$. Finally, if $\beta \in W$ as well then $1\alpha + \beta \in W$. Therefore, W satisfies all the conditions and is a linear subspace of V.

Sufficiency: If W is a subspace of V and $\alpha, \beta \in W$ with $c \in \mathbb{F}$ then $c\alpha + \beta \in W$.

Corollary 2.2. Let V be a vector space over \mathbb{F} . The intersection of any collection subspaces of V is a subspace of V.

Theorem 2.3. Let V be a vector space and W_1 and W_2 be two subspaces of V such that $W_1 \cap W_2$ is a subspace of V. Then $W_1 \subset W_2$ or $W_2 \subset W_1$.

Proof. For the sake of contradiction assume neither $W_1 \subset W_2$ nor $W_2 \subset W_1$. Then, there are vectors α_1 and α_2 such that $\alpha_1 \in W_1$ but $\alpha_2 \notin W_2$ and similarly $\alpha_2 \in W_2$ but $\alpha_2 \notin W_1$. Since $\alpha_1 + \alpha_2 \in W_1 \cup W_2$, then $\alpha_1 + \alpha_2 \in W_1$ or $\alpha_1 + \alpha_2 \in W_2$ which is a contradiction.

Theorem 2.4. Let V be a vector space over the inifinite field \mathbb{F} . If W_1, W_2, \ldots, W_n are subspaces of V and $V \subset \bigcup W_i$, then there exists k such that $V = W_k$.

Proof. too long, do it urself bitch.

2.2 Span

Definition: Let S be a set of vector in a vector space V. The subspace spanned by S is defined to be the intersection of all subsapces of V which contains S and is denoted by span S. That is, span $S = \bigcap_{S \subset W} W$. By Corollary 2.2, span S is a linear subspace. Obviously, span S is the smallest subspace containing S because if there were $S \subset K \subset \operatorname{span} S$ since by definition $\operatorname{span} S = \cap W \subset K$, then $K = \operatorname{span} S$.

Example 2.3. Let $S = \{0\}$ then span $S = \bigcap_{\{0\} \subset W} W = \{0\}$. Moreover:

$$\operatorname{span} \emptyset = \{0\} \qquad \qquad \operatorname{span} V = V$$

Theorem 2.5. Let V is a vector space and $S \neq \emptyset$

$$\operatorname{span} S = \{c_1 \alpha_1 + \dots + c_n \alpha_n \mid \alpha_i \in S, \ c_i \in \mathbb{F}, \ n \in \mathfrak{p}\}\$$

Proof. Let L be the set describe above. Clearly, $S \subset L$ and L is a subsapce of V hence span $S \subset L$. Since $S \subset \operatorname{span} S$ and span S is closed under addition and scalar multiplication then $c_1\alpha_1 + \cdots + c_n\alpha_n \in \operatorname{span} S$ hence $L \subset \operatorname{span} S$.

Definition: If S_1, S_2, \ldots, S_k are subsets of a vector space V, the set of all sums

$$\alpha_1 + \alpha_2 + \cdots + \alpha_k$$

where $\alpha_i \in S_i$ is called the **sum** of subsets S_1, S_2, \ldots, S_k and is denoted by

$$S_1 + \dots + S_k = \{\alpha_1 + \alpha_2 + \dots + \alpha_k \mid \alpha_i \in S_i\}$$

Theorem 2.6. Let W_1, W_2, \ldots, W_k be subspaces of vector space V. Then

$$W = W_1 + W_2 + \dots + W_k$$

is a subspaces of V. Moerover, $W = \operatorname{span} \cup W_i$.

Proof. Let $\alpha \in \operatorname{span} \cup W_i$ and $\beta \in W$. By Theorem 2.5, $\alpha = \alpha_1 + \dots + \alpha_n$ where $\alpha_i \in \cup W_i$. Define $\alpha_1' = \sum_{\alpha_i \in W_1} \alpha_i$, $\alpha_2' = \sum_{\alpha_i \in W_2 \setminus W_1} \alpha_i$ and so on. Clearly, $\alpha = \alpha_1' + \alpha_2' + \dots + \alpha_k'$ thus $\alpha \in W$ and $\operatorname{span} \cup W_i \subset W$. By definition, $\beta = \beta_1 + \dots + \beta_k$ where $\beta_i \in W_i$ and therefore, $\beta_i \in \cap W_i \subset \operatorname{span} \cup W_i$. Since $\operatorname{span} \cup W_i$ is a subspace of V then $\beta \in \operatorname{span} \cup W_i$ which implies $W = \operatorname{span} \cup W_i$.

2.3 Basis and Dimension

Definition (Linearly dependent): A set S of vector space V is said to be **linearly dependent**if there are $c_1, \ldots, c_n \in \mathbb{F}$ where at least one of the c_i is non-zero and $\alpha_1, \ldots, \alpha_n$ such that

$$c_1\alpha_1 + \dots + c_n\alpha_n = 0$$

Furthermore, S is **linearly independent** if it is not linearly dependent. That is, if $c_1\alpha_1 + \cdots + c_n\alpha_n = 0$ then $c_i = 0$ for all $i = 1, \ldots, n$.

Corollary 2.7. If S is linearly dependent and $S \subset S'$ then S' is linearly dependent as well. If S is linearly independent and $S' \subset S$ then S' is linearly independent.

Definition (Basis): Let V be a vector space over field \mathbb{F} . $S \subset V$ is a **basis** of V if it is linearly independent and spans V. Additionally, V is finite dimensional if V has a finite basis.

Theorem 2.8. Let V be a vector space which is spanned by a β_1, \ldots, β_m . Then any independent subset of V contains at most m elements.

Proof. needs matrices maybe.

Corollary 2.9. If V is a finite dimensional vector space then any two bases of V have the same finite number of elements.

Proof. Let α, β be two bases of V. By Theorem 2.8, $|\alpha| \leq |\beta|$ and $|\beta| \leq |\alpha|$ therefore they must have the same number of elements. Since, there exists a finite basis γ then α and β must be finite as well.

Definition (Dimension): Dimension of a finite dimensional vector space V, denoted by $\dim V$ is the number of elements of one of its basis.

Example 2.4. $\dim\{0\} = 0$ since span $\emptyset = \{0\}$ and therefore $\dim\{0\} = |\emptyset| = 0$.

Corollary 2.10. Let V be a finite dimensional vector space and let $n = \dim V$. Then

- 1. any subset of V whichi has more then n elements is linearly dependent.
- 2. no subset of V which contains fewer than n vectors can span V.

Proof. 1. It is an immediate consequence of Theorem 2.8.

2. If there was a set $\beta = \{\beta_1, \dots, \beta_m\}$ with m < n such that $V = \operatorname{span} \beta$ then there must be a basis β' such that $|\beta'| \le m$. Which contradicts Corollary 2.9

Lemma 2.11. Let S be a linearly independent subset of vector space V. Suppose β is a vector in V not spanned by S. Then the set $S \cap \{\beta\}$ is linearly independent.

Proof. Suppose $\alpha_1, \ldots, \alpha_m$ are distinct vectors in S and that

$$c_1\alpha_1 + \dots + c_m\alpha_m + b\beta = 0 \tag{2.1}$$

then if $b \neq 0$

$$\beta = -\frac{1}{b}(c_1\alpha_1 + \dots + c_m\alpha_m)$$

and thus $\beta \in \text{span } S$ which is a contradiction.

Theorem 2.12. If W is a subspace of a finite dimensional vector space V then every linearly independent subset of W is finite and part of a basis of V.

Proof. Clearly, since $W \subset V$ then every linearly independent subset S of W must be finite and $|S| \leq \dim V$. Let $S = \{\alpha_1, \alpha_m\} \subset W$ be linearly independent. If $\operatorname{span} S = V$ we're done otherwise, there is $\alpha_{m+1} \in V$ such that $\alpha_{m+1} \notin \operatorname{span} S$. By Lemma 2.11, $S_1 = S \cup \{\alpha_{m+1}\}$ is linearly independent. Now if $\operatorname{span} S_1 = V$ we are done. If not we continue in similar fashion. Since V is finite and any independent set can not have more than $\dim V$ elements, then in $\dim V - m$ steps the set $S \cap \{\alpha_{m+1}, \ldots, \alpha_n\}$ is a basis for V.

Corollary 2.13. A proper subspace W of a finite dimensional vector space V is finite dimensional and $\dim W < \dim V$.

Proof. Any basis W is linearly independent subset of V and therefore it is finite. Suppose W has basis $\{\beta_1, \ldots, \beta_m\}$ with $m \leq \dim V$. Since span $\beta = W \subsetneq V$ then there is αinV that is not in W therefore, $\beta \cup \{\alpha\}$ is a linearly independent subset of V by Lemma 2.11. Simply

$$|\beta| = |\beta \cup \{\alpha\}| + 1 \le n$$

and thus $\dim W < \dim V$.

Theorem 2.14. If W_1 and W_2 are finite dimensional subspaces of V (not necessarily finite dimensional), then $W_1 + W_2$ is finite dimensional and

$$\dim(W_1 + W_2) = \dim W_1 + \dim W_2 - \dim(W_1 \cap W_2)$$

Proof. $W_1 \cap W_2$ is a finite subsapce of V and thus there are $\alpha = \{\alpha_1, \ldots, \alpha_a\}$ which form a basis for $W_1 \cap W_2$. Furthermore, since $W_1 \cap W_2 \subset W_1, W_2$ then there are vectors $\beta = \{\beta_1, \ldots, \beta_b\}$ and $\gamma = \{\gamma_1, \ldots, \gamma_c\}$ such that $\alpha \cup \beta$ is a basis for W_1 and $\alpha \cup \gamma$ is a basis for W_2 . We claim that $\alpha \cup \beta \cup \gamma$ is a basis for $W_1 + W_2$. For all $\xi_1 \in W_1$ and $\xi_2 \in W_2$, $\xi_1 + \xi_2 \in W_1 + W_2$. Writing $\xi_1 = x_1\alpha_1 + \cdots + x_a\alpha_a + y_1\beta_1 + \cdots + y_b\beta_b$ and $\xi_2 = x_1'\alpha_1 + \ldots x_a'\alpha_a + z_1\gamma_1 + \cdots + z_c\gamma_c$ we get

$$\xi_1 + \xi_2 = \sum_{i=1}^a (x_i + x_i')\alpha_i + \sum_{i=1}^b y_i\beta_i + \sum_{i=1}^c z_i\gamma_i \in \operatorname{span}(\alpha \cup \beta \cup \gamma)$$

and thus $W_1 + W_2 \subset \operatorname{span}(\alpha \cup \beta \cup \gamma)$. Cleary $\operatorname{span}(\alpha \cup \beta \cup \gamma) \subset W_1 + W_2$ and thus $W_1 + W_2 = \operatorname{span}(\alpha \cup \beta \cup \gamma)$. Now we need to show that $\alpha \cup \beta \cup \gamma$ is independent. Consider

$$\sum_{i=1}^{a} x_i \alpha_i + \sum_{i=1}^{b} y_i \beta_i + \sum_{i=1}^{c} z_i \gamma_i = 0$$

$$\implies \sum_{i=1}^{a} x_i \alpha_i + \sum_{i=1}^{b} y_i \beta_i = \sum_{i=1}^{c} z_i \gamma_i$$

and thus $\sum_{i=1}^c z_i \gamma_i \in W_1$ and as a result $\sum_{i=1}^c z_i \gamma_i \in W_1 \cap W_2$ therefore, writing $\sum_{i=1}^c z_i \gamma_i = \sum_{i=1}^a x_i' \alpha_i$

$$\sum_{i=1}^{a} x_i \alpha_i + \sum_{i=1}^{b} y_i \beta_i = -\sum_{i=1}^{a} x_i' \alpha_i$$

$$\implies \sum_{i=1}^{a} (x_i + x_i') \alpha_i + \sum_{i=1}^{b} y_i \beta_i = 0$$

and since $\alpha \cup \beta$ is a basis for W_1 then $y_i = 0$. hence

$$\sum_{i=1}^{a} x_i \alpha_i + \sum_{i=1}^{c} z_i \gamma_i = 0$$

which implies $\alpha_i = 0, \gamma_i = 0$ as $\alpha \cup \gamma$ is a basis for W_2 .