

# Linear Algebra Cheat Sheet for OR

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## Abstract

This document contains a list of all linear algebra results that I needed for my Operations Research coursework at UIUC. The target audience of this document is Operations Research students. It can be helpful to [page](#) these results into your brain before a linear-algebra-heavy exam/project, and proving these results yourself (except those marked DWAP) is a good exercise.

Notation:

- DWAP abbreviates “Don’t Worry About the Proof”.
- For any integer  $n \geq 0$ , define  $[n] := \{1, 2, \dots, n\}$ .

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## 1 Vector Spaces

### 1.1 Field

**Definition 1** (Field). *A field is a set  $F$  equipped with two operations  $+$  and  $\times$  that satisfy some special properties. (you don’t need to know the properties, unless you want to be a pure mathematician, in which case you can find them [here](#)).*

*Every field has two special elements, called the additive identity (usually denoted by  $0$ ), and the multiplicative identity (usually denoted by  $1$ ).*

**Theorem 1** (DWAP). *The following are fields:  $\mathbb{Q}$  (the set of rational numbers),  $\mathbb{R}$  (the set of real numbers),  $\mathbb{C}$  (the set of complex numbers).*

**Theorem 2** (DWAP).  *$\mathbb{Z}_p := \{0, 1, \dots, p-1\}$  is a field when  $p$  is prime, where  $+$  and  $\times$  are addition and multiplication modulo  $p$ .*

## 1.2 Vector Space

**Definition 2.** *A vector space  $V$  over a field  $F$  is a set with a vector addition operation ( $V \times V \mapsto V$ ) and a scalar multiplication operation ( $F \times V \mapsto V$ ) which satisfies some special properties. (you don't need to know the properties, unless you want to be a pure mathematician, in which case you can find them [here](#)).*

*Every vector space has a special vector, called the additive identity, denoted by  $\mathbf{0}$ . The elements of  $V$  are called vectors. The elements of  $F$  are called scalars.*

**Definition 3** (Subspace). *Let  $V$  be a vector space.  $U$  is called a subspace of  $V$  if  $U \subseteq V$  and  $U$  is also a vector space.*

**Theorem 3** (DWAP). *The set of all polynomials over field  $F$  forms a vector space.*

**Theorem 4** (DWAP). *For a field  $F$ ,  $F^d$  is a vector space.*

**Definition 4** (Linear and affine combinations). *Let  $V$  be a vector space over field  $F$ . Let  $X := \{x^{(i)} : i \in [k]\}$ , where  $x^{(i)} \in V$ . Let  $y = \sum_{i=1}^k \alpha_i x^{(i)}$ , where  $\alpha_i \in F$ .*

- *$y$  is called a linear combination of  $X$ .*
- *If  $\sum_{i=1}^k \alpha_i = 1$ , then  $y$  is called an affine combination of  $X$ .*

**Definition 5** (Span).  *$\text{span}(X)$  is defined as the set of all linear combinations of  $X$ . For sets  $X$  and  $Y$  of vectors,  $X$  is called a spanning set of  $Y$  if  $Y \subseteq \text{span}(X)$ .*

**Lemma 5.** *If  $X$  spans  $S$ , then  $X$  also spans  $\text{span}(S)$ .*

**Theorem 6** (DWAP). *Let  $X$  be a finite subset of  $F^d$ , where  $F$  is a field. Then  $\text{span}(X)$  is a vector space.*

## 1.3 Linear Independence and Basis

**Definition 6** (Linear independence). *A set  $\{x_1, x_2, \dots, x_n\}$  of vectors over field  $F$  is called linearly independent iff*

$$\forall (\alpha_1, \dots, \alpha_n) \in F^n, \left( \sum_{i=1}^n \alpha_i x_i = 0 \implies (\alpha_i = 0 \forall i \in [n]) \right).$$

**Lemma 7** (Incrementing a linearly independent set). *Let  $X$  be a linearly independent set of vectors and  $y$  be a vector. If  $y \notin \text{span}(X)$ , then  $X \cup \{y\}$  is linearly independent.*

**Lemma 8** (Decrementing a linearly dependent set). *Let  $X$  be a linearly dependent set of vectors. Then  $\exists x \in X$  such that  $\text{span}(X) = \text{span}(X - \{x\})$ .*

**Theorem 9** (DWAP). *Let  $X$  be a spanning set of vector space  $V$ . If  $Y \subseteq V$  and  $|Y| > |X|$ , then  $Y$  is linearly dependent.*

**Definition 7** (Basis). *Let  $S$  be a subset of vector space  $V$ . Then  $X \subseteq S$  is a basis of  $S$  iff (the following definitions are equivalent):*

- $X$  is linearly independent and spans  $S$ .
- $X$  is the largest linearly independent subset of  $S$ .
- $X$  is a maximal linearly independent subset of  $S$ .
- $X$  is the smallest spanning subset of  $S$ .
- $X$  is a minimal spanning subset of  $S$ .

Equivalence of these definitions can be proven using Theorem 9 and Lemmas 7 and 8.

**Lemma 10.** *If  $X$  is a basis of  $S$ , then  $X$  is also a basis of  $\text{span}(S)$ .*

**Lemma 11.** *Let  $F$  be a field. Let  $e^{(i)} \in F^d$  be a vector whose  $i^{\text{th}}$  coordinate is 1 and other coordinates are 0. Then  $E := \{e^{(i)} : i \in [d]\}$  is a basis of  $F^d$ . ( $E$  is called the standard basis of  $F^d$ .)*

**Theorem 12.** *All bases of  $S$  have the same size. This size is called the rank of  $S$  (denoted as  $\text{rank}(S)$ ). If  $S$  is a vector space, it's called the dimension of  $S$  (denoted as  $\text{dim}(S)$ ).*

**Theorem 13.** *Let  $X$  be a set of  $\text{rank}(S)$  vectors. Then*

$$X \text{ is a basis of } S \iff X \text{ is linearly independent} \iff X \text{ spans } S.$$

**Theorem 14** (Coordinatization). *Let  $B := \{b^{(1)}, b^{(2)}, \dots, b^{(k)}\}$  be a basis of vector space  $V$ . Then  $\forall x \in V$  there is a unique tuple  $(\alpha_1, \alpha_2, \dots, \alpha_k)$  such that  $x = \sum_{i=1}^k \alpha_i b^{(i)}$ .*

## 1.4 Elementary Operations

**Definition 8.** *For a sequence  $X := [x_1, \dots, x_n]$  of vectors (over field  $F$ ), an elementary operation is one of the following:*

- For  $i \neq j$  and  $\alpha \in F$ , replace  $x_i$  by  $x_i + \alpha x_j$ .
- For  $i \in [n]$  and  $\alpha \in F - \{0\}$ , replace  $x_i$  by  $\alpha x_i$ .
- For  $i \neq j$ , swap  $x_i$  and  $x_j$ .

**Lemma 15** (Reversibility). *Let  $X$  be a set of vectors. Let  $Y$  be the vectors obtained by applying an elementary operation on  $X$ . Then  $X$  can be obtained by applying an elementary operation to  $Y$ .*

**Lemma 16.** *For a set  $X$  of vectors, applying elementary operations doesn't change  $\text{span}(X)$  or  $\text{rank}(X)$ .*

## 1.5 Affine Independence

**Definition 9.** A set  $\{x_1, x_2, \dots, x_n\}$  of vectors over field  $F$  is called *affinely independent* iff

$$\forall(\alpha_1, \dots, \alpha_n) \in F^n, \left( \left( \sum_{i=1}^n \alpha_i = 0 \text{ and } \sum_{i=1}^n \alpha_i x_i = 0 \right) \implies (\alpha_i = 0 \forall i \in [n]) \right).$$

**Theorem 17.** The set  $\{x^{(i)} : i \in [n]\}$  of vectors is *affinely independent* iff  $\{x^{(i)} - x^{(n)} : i \in [n-1]\}$  is *linearly independent*.

**Theorem 18.** The set  $\{x^{(i)} : i \in [n]\}$  of vectors from  $F^d$  is *affinely independent* iff  $\{(x^{(i)}, 1) : i \in [n]\}$  is *linearly independent*.

## 2 Matrices

**Lemma 19** (Matrix of elementary rowops). For a sequence  $S$  of elementary operations, there is a unique matrix  $R$  such that applying  $S$  to rows of any matrix  $A$  gives us  $RA$ .

**Definition 10** (row space, column space). Let  $F$  be a field and  $A \in F^{m \times n}$  be a matrix, Let  $\text{rows}(A)$  be the set of all row vectors of  $A$ , and  $\text{cols}(A)$  be the set of all column vectors of  $A$ . Then

- $\text{rowSpace}(A) := \text{span}(\text{rows}(A))$ ,
- $\text{colSpace}(A) := \text{span}(\text{cols}(A))$ ,
- $\text{rank}(A) := \text{rowRank}(A) := \text{rank}(\text{rows}(A))$ ,
- $\text{colRank}(A) := \text{rank}(\text{cols}(A))$ .

**Theorem 20** (DWAP). For any matrix  $A$ ,  $\text{rowRank}(A) = \text{colRank}(A)$ .

**Definition 11** (Nullspace and nullity). For a matrix  $A \in F^{m \times n}$ ,  $\text{nullSpace}(A) := \{x \in F^n : Ax = 0\}$  and  $\text{nullity}(A) := \dim(\text{nullSpace}(A))$ .

**Theorem 21** (Rank-nullity theorem, DWAP).  $\text{rank}(A) + \text{nullity}(A) = |\text{cols}(A)|$ .

*Proof sketch.* We can show that row space and nullspace are not affected by elementary row operations on  $A$ . Hence, we can assume that  $A$  is in Reduced-Row Echelon Form. There are  $\text{rank}(A)$  pivot columns in  $A$ . Given any value of non-pivot variables, we can compute the value of pivot variables such that  $Ax = 0$ . Hence,  $\text{nullity}(A) = |\text{cols}(A)| - \text{rank}(A)$ .  $\square$

**Theorem 22** (DWAP). If  $V$  is a subspace of  $F^d$ , then  $\exists A$  such that  $V = \text{nullSpace}(A)$ .

**Theorem 23** (DWAP). *Basic results on matrix multiplication:*

- Matrix multiplication is associative, i.e.,  $(AB)C = A(BC)$ .
- $(AB)^T = B^T A^T$ .
- $|AB| = |A||B|$  if  $A$  and  $B$  are square ( $|A|$  is the determinant of  $A$ ).
- $(AB)^{-1} = B^{-1}A^{-1}$  if  $A$  and  $B$  are invertible.

**Theorem 24** (Matrix singularity, DWAP). *Let  $A \in F^{n \times n}$ . The following are equivalent:*

- $\text{rank}(A) = n$ .
- $0$  is the unique solution to  $Ax = 0$ .
- $|A| \neq 0$  ( $|A|$  is the determinant of  $A$ ).
- $A$  is invertible, i.e.,  $\exists B \in F^{n \times n}$  such that  $AB = BA = I$ .

**Theorem 25.** *Let  $A \in F^{n \times n}$ . Then*

$$(A^{-1})[i, j] = \frac{(-1)^{i+j}}{|A|} |A[[n] - \{j\}, [n] - \{i\}]|.$$

**Corollary 25.1.**

$$\begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix}^{-1} = \frac{1}{a_{1,1}a_{2,2} - a_{1,2}a_{2,1}} \begin{bmatrix} a_{2,2} & -a_{1,2} \\ -a_{2,1} & a_{1,1} \end{bmatrix}.$$

### 3 Miscellaneous

**Definition 12** ( $p$ -norms). *For  $x \in \mathbb{R}^d$ ,*

$$\|x\|_p := \left( \sum_{i=1}^d |x_i|^p \right)^{1/p} \quad \|x\|_\infty := \max_{i=1}^d |x_i| \quad \|x\| := \|x\|_2$$

**Definition 13** (Linear combinations). *Let  $V$  be a vector space over field  $\mathbb{R}$ . Let  $X := \{x^{(i)} : i \in [k]\}$ , where  $x^{(i)} \in V$ . Let  $y = \sum_{i=1}^k \alpha_i x^{(i)}$ , where  $\alpha_i \in F$ .*

- $y$  is called a linear combination of  $X$ .
- If  $\alpha_i \geq 0$  for all  $i \in [k]$ , then  $y$  is called a non-negative linear combination of  $X$ .
- If  $\sum_{i=1}^k \alpha_i = 1$ , then  $y$  is called an affine combination of  $X$ .
- A non-negative affine combination is called a convex combination.

**Theorem 26** (Cauchy-Schwarz inequality).  $\forall x, y \in \mathbb{R}^d, |x^T y| \leq \|x\| \|y\|$ .