

**4.1 Vector spaces and subspaces**

- Know the definition of a vector space and how to decide whether a given set is a vector space or not.
- Know the definition of a subspace of a vector space and how to decide whether a given subset of a vector space is a subspace or not.
- For given vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  in a vector space  $\mathcal{V}$  know the definition of a linear combination and the span, denoted by  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$ .
- Know that  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$  is a subspace of  $\mathcal{V}$ .

**4.2 Null spaces, column spaces, and linear transformations**

- For a given  $m \times n$  matrix  $A$  know the definitions of  $\text{Nul } A$ ,  $\text{Col } A$  and  $\text{Row } A$  and how to find bases for these subspaces. (See the post of October 4, Section 4.3 and Section 4.5.)

**4.3 Linearly independent sets; bases**

- Know the definition of linearly independent vectors and a basis for a subspace.
- Know how to decide whether given vectors from  $\mathbb{R}^n$  are linearly independent or not.
- Know how to prove that the monomials  $\mathbf{q}_0(t) = 1$ ,  $\mathbf{q}_1(t) = t$ ,  $\mathbf{q}_2(t) = t^2$ , are linearly independent.
- Know how to prove that the functions  $\mathbf{c}(t) = \cos t$ ,  $\mathbf{s}(t) = \sin t$  are linearly independent.

**4.4 Coordinate systems.**

- Let  $\mathcal{V}$  be a vector space and let  $\mathcal{B}$  be a basis for  $\mathcal{V}$ . Know: the unique representation theorem, the definition of a coordinate mapping, the meaning of the symbol  $[\mathbf{v}]_{\mathcal{B}}$  for a given vector  $\mathbf{v}$  in  $\mathcal{V}$ .
- The importance of the matrix

$$P_{\mathcal{B}} = [\mathbf{b}_1 \cdots \mathbf{b}_n]$$

for a given basis  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  for  $\mathbb{R}^n$  (this is a special change-of-coordinate matrix, more in Section 4.7)

- Theorem 8: Given a basis  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  of a vector space  $\mathcal{V}$ , the coordinate mapping  $\mathbf{v} \mapsto [\mathbf{v}]_{\mathcal{B}}$  is a one-to-one linear transformation from  $\mathcal{V}$  onto  $\mathbb{R}^n$ . In other words, the coordinate mapping  $\mathbf{v} \mapsto [\mathbf{v}]_{\mathcal{B}}$  is a linear bijection from  $\mathcal{V}$  to  $\mathbb{R}^n$ .
- Know how a coordinate mapping for polynomials works, Examples 5 and 6
- Know Exercises 10, 11, 13

**4.5 The dimension of a vector space**

- The definition of a finite dimensional vector space and the definition of the dimension of a finite dimensional vector space

**4.6 Rank**

- Know the definition of the rank of a matrix, denoted by  $\text{rank } A$ .
- Let  $m$  and  $n$  be positive integers. For a given  $m \times n$  matrix  $A$  know the relationship between the following nonnegative integers:

$$\dim(\text{Nul } A), \quad \dim(\text{Col } A), \quad \dim(\text{Row } A), \quad \text{rank } A, \quad m, \quad n$$

**4.7 Change of bases (in fact: Change of coordinates).**

- Know that the matrix  $M$  with the property  $[\mathbf{v}]_{\mathcal{C}} = M[\mathbf{v}]_{\mathcal{B}}$  is called the **change-of-coordinate matrix from  $\mathcal{B}$  to  $\mathcal{C}$** . It is denoted  ${}_{\mathcal{C} \leftarrow \mathcal{B}} P$  and it is calculated as

$${}_{\mathcal{C} \leftarrow \mathcal{B}} P = \begin{bmatrix} [\mathbf{b}_1]_{\mathcal{C}} & \cdots & [\mathbf{b}_n]_{\mathcal{C}} \end{bmatrix};$$

here  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  and  $\mathcal{C}$  are two bases of an  $n$ -dimensional vector space  $\mathcal{V}$  and  $[\mathbf{v}]_{\mathcal{B}} \in \mathbb{R}^n$  denotes the coordinate vector of  $\mathbf{v} \in \mathcal{V}$  relative to the basis  $\mathcal{B}$ .

- $\left({}_{\mathcal{C} \leftarrow \mathcal{B}} P\right)^{-1} = {}_{\mathcal{B} \leftarrow \mathcal{C}} P$
- Know that there is a special basis of  $\mathbb{R}^n$ , called the **standard basis**, which consists of the columns of the identity matrix  $I_n$ . It is denoted by  $\mathcal{E}$ . This notation comes from the fact that the vectors in this basis are commonly denoted by  $\mathbf{e}_1, \dots, \mathbf{e}_n$ .
- Know that for a basis  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  for  $\mathbb{R}^n$  (this is a special case) we have

$${}_{\mathcal{E} \leftarrow \mathcal{B}} P = [\mathbf{b}_1 \cdots \mathbf{b}_n] = P_{\mathcal{B}}$$

(the matrix  $P_{\mathcal{B}}$  was introduced in Section 4.4); this basically says that  ${}_{\mathcal{E} \leftarrow \mathcal{B}} P$  is a very friendly matrix.

- Know  ${}_{\mathcal{C} \leftarrow \mathcal{B}} P = \left({}_{\mathcal{C} \leftarrow \mathcal{E}} P\right) \left({}_{\mathcal{E} \leftarrow \mathcal{B}} P\right) = \left({}_{\mathcal{E} \leftarrow \mathcal{C}} P\right)^{-1} {}_{\mathcal{E} \leftarrow \mathcal{B}} P = [\mathbf{c}_1 \cdots \mathbf{c}_n]^{-1} [\mathbf{b}_1 \cdots \mathbf{b}_n]$  where  $\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}$  is another basis for  $\mathbb{R}^n$ ; this basically says that  ${}_{\mathcal{C} \leftarrow \mathcal{B}} P$  can be written as a product of a very friendly matrix and the inverse of another very friendly matrix
- Know Exercises 4 - 10, 13, 14.

### 5.1 Eigenvectors and eigenvalues.

- Know the definition of an eigenvector and eigenvalue. It is a little tricky. Pay attention.
- Know the definition of an eigenspace and how to find an eigenspace corresponding to a given eigenvalue.
- Know that the eigenvalues of a triangular matrix are the entries on its main diagonal.
- **Theorem.** Eigenvectors corresponding to distinct eigenvalues are linearly independent. (Or, in a more formal mathematical language: Let  $A$  be an  $n \times n$  matrix, let  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m \in \mathbb{R}^n$  and  $\lambda_1, \lambda_2, \dots, \lambda_m \in \mathbb{R}$ . If  $A\mathbf{v}_k = \lambda_k \mathbf{v}_k$ ,  $\mathbf{v}_k \neq \mathbf{0}$  and  $\lambda_j \neq \lambda_k$  for all  $j, k = 1, 2, \dots, m$ , then  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  are linearly independent.

### 5.2 The characteristic equation.

- Know that  $\lambda$  is an eigenvalue of an  $n \times n$  matrix  $A$  if and only if  $\det(A - \lambda I) = 0$  (this is the characteristic equation for  $A$ )
- Know how to calculate  $\det(A - \lambda I)$  for  $2 \times 2$  and  $3 \times 3$  matrices, how to find eigenvalues and corresponding eigenvectors. Exercises 1-8, but do more and find eigenvectors as well. See the post of October 10.

### 5.3 Diagonalization.

- **Theorem.** (The diagonalization theorem) An  $n \times n$  matrix  $A$  is diagonalizable if and only if  $A$  has  $n$  linearly independent eigenvectors.
- Know how to decide whether a given  $2 \times 2$  and  $3 \times 3$  matrix  $A$ , is diagonalizable or not; if it is diagonalizable, how to find invertible matrix  $P$  and diagonal matrix  $D$  such that  $A = PDP^{-1}$ . See the post of October 10.
- Know how to decide whether a triangular matrix is diagonalizable or not. Consider the matrix  $A$  in Exercise 18 in Section 5.2 and find  $h$  such that the matrix  $A$  is diagonalizable.

### 5.4 Eigenvectors and linear transformations.

- Know that the matrix  $M$  with the property  $[T\mathbf{v}]_{\mathcal{C}} = M[\mathbf{v}]_{\mathcal{B}}$  is called the **matrix of  $T$  relative to the bases  $\mathcal{B}$  and  $\mathcal{C}$**  and it is calculated as

$$M = \begin{bmatrix} [T\mathbf{b}_1]_{\mathcal{C}} & \cdots & [T\mathbf{b}_n]_{\mathcal{C}} \end{bmatrix},$$

where  $T$  be a linear transformation from an  $m$ -dimensional vector space  $\mathcal{V}$  to an  $n$ -dimensional vector space  $\mathcal{W}$ ,  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_m\}$  is a basis for  $\mathcal{V}$  and  $\mathcal{C}$  is a basis for  $\mathcal{W}$ .

- Know that the matrix  $M$  with the property  $[T\mathbf{v}]_{\mathcal{B}} = M[\mathbf{v}]_{\mathcal{B}}$  is called the **matrix of  $T$  relative to the basis  $\mathcal{B}$** , it is denoted by  $[T]_{\mathcal{B}}$  and it is calculated as

$$[T]_{\mathcal{B}} = \begin{bmatrix} [T\mathbf{b}_1]_{\mathcal{B}} & \cdots & [T\mathbf{b}_n]_{\mathcal{B}} \end{bmatrix},$$

where  $T : \mathcal{V} \rightarrow \mathcal{V}$  is a linear transformation on an  $n$ -dimensional vector space  $\mathcal{V}$  and  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  is a basis for  $\mathcal{V}$ .

- **Theorem.** (The diagonal matrix representation) Let  $A, D, P$  be  $n \times n$  matrices, where  $P$  is invertible,  $D$  is diagonal and  $A = PDP^{-1}$ . If  $\mathcal{B}$  is the basis for  $\mathbb{R}^n$  which consists of the columns of  $P$  and if  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  is given by  $T\mathbf{v} = A\mathbf{v}$  for all  $\mathbf{v} \in \mathbb{R}^n$ , then  $[T]_{\mathcal{B}} = D$ .

## 5.5 Complex eigenvalues.

- Know how to calculate complex eigenvalues and corresponding eigenvectors of  $2 \times 2$  and  $3 \times 3$  real matrices
- Know that the most important class of  $2 \times 2$  matrices with complex eigenvalues are rotation matrices; the matrix of the counterclockwise rotation about the origin by the angle  $\theta$  measured in radians relative to the standard basis for  $\mathbb{R}^2$  is

$$R_{\theta} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix};$$

the  $2 \times 2$  matrix  $R_{\theta}$  is called the **rotation matrix**.

- **Theorem.** (A “hiding rotation” theorem.) Let  $A$  be a real  $2 \times 2$  matrix with a nonreal eigenvalue. Then there exist an invertible  $2 \times 2$  matrix  $P$ , a positive scalar  $\alpha$  and a rotation matrix  $R_{\theta}$  such that  $A = \alpha P R_{\theta} P^{-1}$ .
- In relation to the previous item see the post on October 15.

## 6.1 Inner product, length, and orthogonality.

- Know the definition of the dot product in  $\mathbb{R}^n$ , its basic properties and calculations involving it. For

$$\mathbf{u} = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} \in \mathbb{R}^n, \quad \mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} \in \mathbb{R}^n, \quad \mathbf{u} \cdot \mathbf{v} = \sum_{k=1}^n u_k v_k = \mathbf{u}^{\top} \mathbf{v} = \mathbf{v}^{\top} \mathbf{u}.$$

- Know the definition, the basic properties of the length of a vector in  $\mathbb{R}^n$ , its properties and calculations involving it.
- Know the definition of the distance in  $\mathbb{R}^n$  and calculations involving it.
- Know the definition of orthogonality in  $\mathbb{R}^n$  and calculations involving it.
- Know the statement and the proof of the linear algebra version of the **Pythagorean theorem**.
- Know the definition and the basic properties of the orthogonal complement in  $\mathbb{R}^n$ .
- Know that for a given  $m \times n$  matrix  $A$  we have  $(\text{Row } A)^{\perp} = \text{Nul } A$  and  $(\text{Col } A)^{\perp} = \text{Nul}(A^{\top})$ .
- Know the geometric interpretation of the dot product in  $\mathbb{R}^2$  and  $\mathbb{R}^3$ :

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \vartheta, \tag{1}$$

where  $\mathbf{u}$  and  $\mathbf{v}$  are vectors in  $\mathbb{R}^2$  or in  $\mathbb{R}^3$ ,  $\vartheta$  is the angle at the vertex  $O$  in the triangle  $OAB$  with  $O$  being the origin,  $A$  being the endpoint of  $\mathbf{u}$  and  $B$  the endpoint of  $\mathbf{v}$ . (You should know the proof of formula (1).)