### Unitary, orthogonal, Hermitian matrices

#### 1 Unitary matrices

Take N orthonormal basis vectors in  $\mathbb{C}^N$ :  $\mathbf{u}_1, \dots, \mathbf{u}_N$ .

$$\mathbf{u}_{1} = (u_{11}, u_{21}, \dots, u_{N1})$$

$$\vdots$$

$$\mathbf{u}_{j} = (u_{1j}, u_{2j}, \dots, u_{Nj})$$

$$\vdots$$

$$\mathbf{u}_{N} = (u_{1N}, u_{2N}, \dots, u_{NN}).$$

Orthonormal means

$$(\mathbf{u}_j, \mathbf{u}_k) = \sum_{l=1}^{N} \overline{u}_{lj} \, u_{lk} = \delta_{jk} \tag{1}$$

Construct a  $N \times N$  matrix U by filling the j-th column with  $\mathbf{u}_j$ :

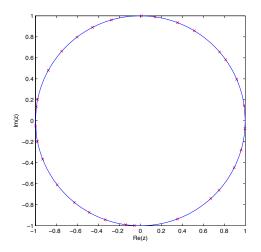
$$U = \begin{pmatrix} u_{11} & u_{12} & \cdots & u_{1N} \\ u_{21} & u_{22} & \cdots & u_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ u_{N1} & u_{N2} & \cdots & u_{NN} \end{pmatrix}$$

The matrix U is called unitary

The eigenvalues of U are complex numbers of modulo one, i.e.

$$e^{i\theta_1},\ldots,e^{i\theta_N}.$$

The eigenvalues of a unitary matrix lie on the unit circle in the complex plane.



A  $2 \times 2$  example of unitary matrix:

$$\begin{pmatrix} e^{i\phi_1}\cos\theta & -e^{-i(\chi-\phi_2)}\sin\theta\\ e^{i(\chi+\phi_1)}\sin\theta & e^{i\phi_2}\cos\theta \end{pmatrix}$$

You can check that this is constructed of orthonormal vectors as described above.

### 2 Hermitian inner product

The Hermitian product is a map  $\mathbb{C}^N \times \mathbb{C}^N \to \mathbb{C}$ , denoted by  $(\mathbf{v}, \mathbf{w})$ , with the following properties:

- 1.  $(\mathbf{v}, \mathbf{w}) = \overline{(\mathbf{w}, \mathbf{v})};$
- 2.  $(\mathbf{v}, \mathbf{w} + \mathbf{z}) = (\mathbf{v}, \mathbf{w}) + (\mathbf{v}, \mathbf{z});$
- 3.  $(\mathbf{v}, \alpha \mathbf{w}) = \alpha(\mathbf{v}, \mathbf{w}), \text{ from 1. } (\alpha \mathbf{v}, \mathbf{w}) = \overline{\alpha}(\mathbf{v}, \mathbf{w}) \quad \alpha \in \mathbb{C}$
- 4.  $(\mathbf{v}, \mathbf{v}) \geq 0$ , where the equality holds if and only if  $\mathbf{v} = \mathbf{0}$ .
- Two vectors are said to be orthogonal if  $(\mathbf{v}, \mathbf{w}) = 0$ .
- We define the *norm* of a vector to be  $|\mathbf{v}| = \sqrt{(\mathbf{v}, \mathbf{v})}$ . Example:

$$(\mathbf{v}, \mathbf{w}) = \sum_{j=1}^{N} \overline{v}_j w_j.$$

This is the inner product that we will always consider.

• The Hermitian conjugate or adjoint  $M^{\dagger}$  of a matrix M is defined by

$$(\mathbf{v}, M\mathbf{w}) = (M^{\dagger}\mathbf{v}, \mathbf{w}) \quad \forall \mathbf{v}, \mathbf{w} \in \mathbb{C}^{N}.$$

• Written in terms of components this formula becomes

$$\sum_{j,k=1}^{N} \overline{v}_{j} m_{jk} w_{k} = \sum_{j,k=1}^{N} \overline{m}^{\dagger}_{kj} \overline{v}_{j} w_{k}, \qquad (2)$$

which implies  $m_{jk} = \overline{m^{\dagger}}_{kj}$  or  $m_{jk}^{\dagger} = \overline{m}_{kj}$ .

• Sometimes  $M^{\dagger}$  is called *conjugate transpose* of M and is denoted by  $M^*$ .

In other words, if

$$M = \begin{pmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{pmatrix}$$

then

$$M^{\dagger} = \begin{pmatrix} \overline{m}_{11} & \overline{m}_{21} & \overline{m}_{31} \\ \overline{m}_{12} & \overline{m}_{22} & \overline{m}_{32} \\ \overline{m}_{13} & \overline{m}_{23} & \overline{m}_{33} \end{pmatrix}$$

For complex matrices the adjoint (conjugate transpose) takes the place of the transpose of a real matrix.

## 3 Hermitian inner product and orthogonal matrices:

Example:  $\mathbb{R}^3$ .

Let  $\mathbf{a} = (a_1, a_2, a_3)$  and  $\mathbf{b} = (b_1, b_2, b_3)$  two vectors in  $\mathbb{R}^3$ .

We have

$$\mathbf{a} \cdot \mathbf{b} = (\mathbf{a}, \mathbf{b}) = |\mathbf{a}| |\mathbf{b}| \cos \phi$$

Here  $\phi$  is the angle between **a** and **b**: see the illustration in Figure 1. You can check that this operation is linear in both arguments

Let  $\mathbf{i}$ ,  $\mathbf{j}$  and  $\mathbf{k}$  be the orthogonal unit vectors along the three axes  $x_1$ ,  $x_2$  and  $x_3$  respectively.

We have

$$(\mathbf{i}, \mathbf{j}) = (\mathbf{j}, \mathbf{k}) = (\mathbf{i}, \mathbf{k}) = 0$$
  
 $(\mathbf{i}, \mathbf{i}) = (\mathbf{j}, \mathbf{j}) = (\mathbf{k}, \mathbf{k}) = 1.$ 

If we write  $\mathbf{a} = a_1 \mathbf{i} + a_2 \mathbf{j} + a_3 \mathbf{k}$  and  $\mathbf{b} = b_1 \mathbf{i} + b_2 \mathbf{j} + b_3 \mathbf{k}$  we obtain

$$(\mathbf{a}, \mathbf{b}) = a_1 b_1 + a_2 b_2 + a_3 b_3. \tag{3}$$

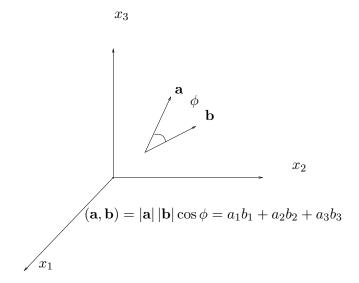


Figure 1:

Rigid rotations in space are achieved by  $3 \times 3$  orthogonal matrices, i.e.

$$\mathbf{a}' = O\mathbf{a}$$
 and  $\mathbf{b}' = O\mathbf{b}$ .

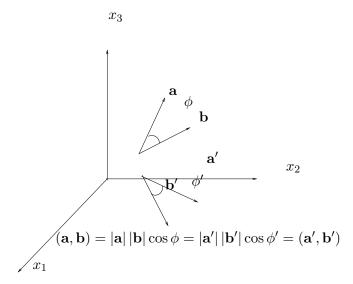


Figure 2:

Rigid rotations in space preserve length and angles: see Figure 2. This means that

$$(\mathbf{a}, \mathbf{b}) = (O\mathbf{a}, O\mathbf{b}) = (\mathbf{a}', \mathbf{b}').$$

Let  $\mathbf{x} = (x_1, \dots, x_N)$  and  $\mathbf{y} = (y_1, \dots, y_N)$  be two vectors in  $\mathbb{R}^N$ . The scalar product in  $\mathbb{R}^N$  is defined

$$(\mathbf{x}, \mathbf{y}) = \sum_{j=1}^{N} x_j y_j$$

**Definition:** An  $N \times N$  real matrix is called orthogonal if it preserves the scalar product in  $\mathbb{R}^N$ 

$$(O\mathbf{x}, O\mathbf{y}) = (\mathbf{x}, \mathbf{y}).$$

In particular O preserves the length of a vector, i.e.

$$|O\mathbf{x}|^2 = (O\mathbf{x}, O\mathbf{x}) = (\mathbf{x}, \mathbf{x}) = |\mathbf{x}|^2 = \sum_{j=1}^{N} x_i^2.$$

Let  $O = (o_{ij})$  be an  $N \times N$  orthogonal matrix. The *i*-th element  $(O\mathbf{x})_i$  of the vector  $O\mathbf{x}$  is

$$(O\mathbf{x})_i = \sum_{j=1}^N o_{ij} x_j$$

The definition in terms of matrix elements reads

$$(O\mathbf{x}, O\mathbf{y}) = \sum_{i,j,k=1}^{N} o_{ij} x_j o_{ik} y_k = \sum_{i,j,k=1}^{N} x_j o_{ij} o_{ik} y_k = \sum_{j=1}^{N} x_j y_j = (\mathbf{x}, \mathbf{y}).$$

This means

$$\sum_{i=1}^{N} o_{ij} o_{ik} = \delta_{jk}$$

Thus,  $\sum_{j,k=1}^{N} x_j \delta_{jk} y_k = \sum_{j=1}^{N} x_j y_j$ In other words, take the j-th and k-th column, multiply element by element and add along the rows.

The columns of an orthogonal matrix are a set of orthonormal vectors.

Now, let  $O^t = (o_{ij}^t)$  be the **transpose matrix** of  $O = (o_{ij})$ , i.e.  $o_{ij}^t = o_{ji}$ But because O preserves the scalar product

$$\sum_{i=1}^{N} o_{ij} o_{ik} = \delta_{jk} = \sum_{i=1}^{N} o_{ji}^{t} o_{ik}.$$

Or equivalently  $O^t O = I$ . In other words  $O^t = O^{-1}$ Note also that

$$O^t O = I$$
$$OO^t O = O$$

Therefore  $OO^t = I$ .

An equivalent definition of an orthogonal matrix is

$$OO^t = O^tO = I. (4)$$

In terms of matrix elements equation (4) reads

$$\sum_{i=1}^{N} o_{ij} o_{ik} = \sum_{i=1}^{N} o_{ji} o_{ki} = \delta_{jk}.$$

Both the sets of rows and columns of an orthogonal matrix are orthonormal bases.

To construct the  $2 \times 2$  orthogonal that corresponds to a anti-clockwise rotation of an angle  $\theta$  in  $\mathbb{R}^2$  (the two-dimensional real plane), let **i** and **j** be the unit vectors along the two axes  $x_1, x_2$  of a right-handed reference frame. Then

$$\mathbf{i} \mapsto \cos \theta \mathbf{i} + \sin \theta \mathbf{j}$$
  
 $\mathbf{j} \mapsto -\sin \theta \mathbf{i} + \cos \theta \mathbf{j}$ 

In components the above equations read

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} \mapsto \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} 0 \\ 1 \end{pmatrix} \mapsto \begin{pmatrix} -\sin \theta \\ \cos \theta \end{pmatrix}$$

Therefore the matrix is

$$O = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}.$$

#### 4 Hermitian matrices

•  $H = (h_{jk})$  is called Hermitian or self-adjoint if

$$H^{\dagger} = H \text{ or } h_{jk} = \overline{h}_{kj}.$$
 (5)

- If the elements of H are real, then trivially  $H = H^t$  or  $h_{jk} = h_{kj}$ , i.e. H is symmetric.
- 1. The eigenvalues of a Hermitian matrix are real.

Suppose **x** is an eigenvector of H with eigenvalue  $\lambda \neq 0$ , then

$$(\mathbf{x}, H\mathbf{x}) = (\mathbf{x}, \lambda\mathbf{x}) = \lambda(\mathbf{x}, \mathbf{x}) = (H\mathbf{x}, \mathbf{x}) = (\lambda\mathbf{x}, \mathbf{x}) = \overline{\lambda}(\mathbf{x}, \mathbf{x}).$$

Since  $(\mathbf{x}, \mathbf{x})$  is real and positive  $\lambda = \overline{\lambda}$ . Note: this proof does not use the fact that the scalar product is defined in a complex linear space. Therefore, it holds for symmetric matrices too. That is, the eigenvalues of a symmetric matrix are real.

2. Eigenvectors of a Hermitian matrix corresponding to different eigenvalues are orthogonal.

Let  $H\mathbf{x} = \lambda \mathbf{x}$  and  $H\mathbf{y} = \mu \mathbf{y}$ . Then

$$(\mathbf{x}, H\mathbf{y}) = \mu(\mathbf{x}, \mathbf{y}) \text{ and } (H\mathbf{x}, \mathbf{y}) = \lambda(\mathbf{x}, \mathbf{y}).$$
 (6)

Since  $(\mathbf{x}, H\mathbf{y}) = (H\mathbf{x}, \mathbf{y})$ , by subtracting both sides of these equations we obtain

$$0 = (\mu - \lambda)(\mathbf{x}, \mathbf{y}). \tag{7}$$

Since  $\lambda \neq \mu$  it must be that  $(\mathbf{x}, \mathbf{y}) = 0$ .

3. A  $N \times N$  Hermitian matrix can be diagonalized and admits a set of N orthogonal eigenvectors.

Note: Facts 2 and 3 hold for symmetric matrices too.

A matrix H is called anti-Hermitian if  $H=-H^{\dagger}$ . If H is Hermitian iH is anti-Hermitian.

### 5 More on unitary matrices

Unitary matrices are the generalization to complex matrices of orthogonal matrices.

**Definition:** A  $N \times N$  complex matrix  $U = (u_{jk})$  is called *unitary* if it preserves the Hermitian inner product:

$$(U\mathbf{v}, U\mathbf{w}) = (\mathbf{v}, \mathbf{w}) \tag{8}$$

Immediate consequences:

1.  $(U\mathbf{v}, U\mathbf{w}) = (\mathbf{v}, U^{\dagger}U\mathbf{w}) = (\mathbf{v}, \mathbf{w}) \Rightarrow UU^{\dagger} = U^{\dagger}U = I$ . In components

$$\sum_{k=1}^{N} \overline{u}_{jk} u_{lk} = \sum_{k=1}^{N} \overline{u}_{kj} u_{kl} = \delta_{jl}.$$

Equivalently, the rows (columns) of a unitary matrix are sets of orthonormal vectors, ie. they are bases in  $\mathbb{C}^N$ .

- 2. Hermitian (symmetric) matrices are diagonalized by unitary (orthogonal) matrices. The eigenvectors  $\{\mathbf{u}_j\}_{j=1,\dots,N}$  of a Hermitian (symmetric) matrix are orthogonal. If they are normalized so that  $|\mathbf{x}_j| = 1$ , then they are an orthonormal basis in  $\mathbb{C}^N$  ( $\mathbb{R}^N$ ). By definition, the matrix U whose columns are the vectors  $\mathbf{u}_j$  is unitary (orthogonal).
- 3. The eigenvalues of a unitary matrix are complex numbers with absolute value one.

If **x** is an eigenvector with eigenvalue  $\lambda$ , then  $(U\mathbf{x}, U\mathbf{x}) = (\lambda \mathbf{x}, \lambda \mathbf{x}) = |\lambda|^2 (\mathbf{x}, \mathbf{x}) = (\mathbf{x}, \mathbf{x})$ , therefore  $|\lambda|^2 = 1$  (This is true for orthogonal matrices too, but note that the eigenvalues of an orthogonal matrix are not real.)

4. The product of two unitary matrices is a unitary matrix.

Let U and V be unitary matrices, then  $(VU\mathbf{v}, VU\mathbf{w}) = (U\mathbf{v}, U\mathbf{w}) = (\mathbf{v}, \mathbf{w})$ . Therefore, by definition VU is unitary. Note that  $(VU)^{\dagger} = U^{\dagger}V^{\dagger}$ .

5. A unitary matrix maps orthonormal bases into orthonormal bases.

$$(\mathbf{v}, \mathbf{w}) = 0 \Rightarrow (U\mathbf{v}, U\mathbf{w}) = 0$$

- 6. Eigenvectors of a unitary matrix corresponding to different eigenvalues are orthogonal.
- 7. A  $N \times N$  unitary matrix can be diagonalized and admits a set of N orthogonal eigenvectors.

The statements 4, 5, 6, 7 are true for orthogonal matrices too.

8. Unitary matrices are diagonalized by unitary matrices, i.e.  $B^{\dagger}UB = \Lambda$ , with  $\Lambda = \text{diag}(e^{i\theta_1}, \dots, e^{i\theta_N})$ , i.e.

$$\Lambda = \begin{pmatrix} e^{i\theta_1} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & e^{i\theta_N} \end{pmatrix},$$
(9)

Note that B and  $\Lambda$  are unitary matrices too.

**Important:** This statement is not true for orthogonal matrices, <u>i.e.</u> orthogonal matrices are not diagonalized by orthogonal matrices.

# 6 Groups

**Definition:** A set G equipped with a multiplication law, is said to be a group if

- 1.  $g_1, g_2 \in G$  implies that  $g_3 = g_1g_2$  belongs to G too.
- 2.  $(g_1g_2)g_3 = g_1(g_2g_3)$
- 3. There exists an element  $e \in G$  such that eg = ge = g for every element  $g \in G$ .
- 4. For every element  $g \in G$  there must be an inverse  $g^{-1} \in G$  such that  $gg^{-1} = g^{-1}g = e$ .

#### **Examples:**

- $\bullet$  The permutations of N letters form a group.
- Let  $GL(N, \mathbb{C})$  the set of all complex matrices M such that  $\det M \neq 0$ .  $GL(N, \mathbb{C})$  equipped with the usual matrix multiplication is a group.(Remember that  $M^{-1}$  exists if and only if  $\det(M) \neq 0$ .)

#### 6.1 The Unitary Group

The set of  $N \times N$  unitary matrices form a group denoted by  $\mathrm{U}(N)$ . It is a subgroup of  $\mathrm{GL}(N,\mathbb{C})$ .

Checks:

- 1. If U and V are unitary so is UV.
- 2. Associative: if U, V, W are unitary, then (UV)W = U(VW)
- 3. The identity matrix  $\begin{pmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{pmatrix}$  is trivially unitary.
- 4. If U is unitary, so is  $U^{\dagger}$  and  $U^{-1} = U^{\dagger}$ .