# Appendix A <br> Linear Algebra for Quantum Computation 

The purpose of this appendix is to compile the definitions, notations, and facts of linear algebra that are important for this book. This appendix also serves as a quick reference for the main operations in vector spaces, for instance, the inner and tensor products. Quantum computation inherited linear algebra from quantum mechanics as the supporting language for describing this area. Therefore, it is essential to have a solid knowledge of the basic results of linear algebra to understand quantum computation and quantum algorithms. If the reader does not have this base knowledge, we suggest reading some of the basic references recommended at the end of this appendix.

## A. 1 Vector Spaces

A vector space $V$ over the field of complex numbers $\mathbb{C}$ is a non-empty set of elements called vectors. In $V$, it is defined the operations of vector addition and multiplication of a vector by a scalar in $\mathbb{C}$. The addition operation is associative and commutative. It also obeys properties

- There is an element $\mathbf{0} \in V$, such that, for each $\mathbf{v} \in V, \mathbf{v}+\mathbf{0}=\mathbf{0}+\mathbf{v}=\mathbf{v}$ (existence of neutral element)
- For each $\mathbf{v} \in V$, there exists $\mathbf{u}=(-1) \mathbf{v}$ in $V$ such that $\mathbf{v}+\mathbf{u}=\mathbf{0}$ (existence of inverse element)
$\mathbf{0}$ is called zero vector. The scalar multiplication operation obeys properties
- $a .(b . \mathbf{v})=(a . b) . \mathbf{v}$ (associativity)
- $1 . \mathbf{v}=\mathbf{v}$ (1 is the neutral element of multiplication)
- $(a+b) . \mathbf{v}=a . \mathbf{v}+b . \mathbf{v}$ (distributivity of sum of scalars)
- $a .(\mathbf{v}+\mathbf{w})=a . \mathbf{v}+a . \mathbf{w}$ (distributivity in $V)$
where $\mathbf{v}, \mathbf{w} \in V$ and $a, b \in \mathbb{C}$.

A vector space can be infinite, but in most applications in quantum computation, finite vector spaces are used and are denoted by $\mathbb{C}^{n}$. In this case, the vectors have $n$ complex entries. In this book, we rarely use infinite spaces, and in these few cases, we are interested only in finite subspaces. In the context of quantum mechanics, infinite vector spaces are used more frequently than finite spaces.

A basis for $\mathbb{C}^{n}$ consists of exactly $n$ linearly independent vectors. If $\left\{\mathbf{v}_{1}, \ldots, \mathbf{v}_{n}\right\}$ is a basis for $\mathbb{C}^{n}$, then a generic vector $\mathbf{v}$ can be written as

$$
\mathbf{v}=\sum_{i=1}^{n} a_{i} \mathbf{v}_{i},
$$

where coefficients $a_{i}$ are complex numbers. The dimension of a vector space is the number of basis vectors.

## A. 2 Inner Product

The inner product is a binary operation $(\cdot, \cdot): V \times V \mapsto \mathbb{C}$, which obeys the following properties:

1. $(\cdot, \cdot)$ is linear in the second argument

$$
\left(\mathbf{v}, \sum_{i=1}^{n} a_{i} \mathbf{v}_{i}\right)=\sum_{i=1}^{n} a_{i}\left(\mathbf{v}, \mathbf{v}_{i}\right)
$$

2. $\left(\mathbf{v}_{1}, \mathbf{v}_{2}\right)=\left(\mathbf{v}_{2}, \mathbf{v}_{1}\right)^{*}$.
3. $(\mathbf{v}, \mathbf{v}) \geq 0$. The equality holds if and only if $\mathbf{v}=\mathbf{0}$.

In general, the inner product is not linear in the first argument. The property in question is called conjugate-linear.

There is more than one way to define an inner product on a vector space. In $\mathbb{C}^{n}$, the most used inner product is defined as follows: If

$$
\mathbf{v}=\left[\begin{array}{c}
a_{1} \\
\vdots \\
a_{n}
\end{array}\right], \quad \mathbf{w}=\left[\begin{array}{c}
b_{1} \\
\vdots \\
b_{n}
\end{array}\right],
$$

then

$$
(\mathbf{v}, \mathbf{w})=\sum_{i=1}^{n} a_{i}^{*} b_{i} .
$$

This expression is equivalent to the matrix product of the transpose-conjugate vector, which is usually denoted by $\mathbf{v}^{\dagger}$, by $\mathbf{w}$.

If an inner product is introduced in a vector space, we can define the notion of orthogonality. Two vectors are orthogonal if the inner product is zero. We also introduce the notion of norm using the inner product. The norm of $\mathbf{v}$, denoted by $\|\mathbf{v}\|$, is defined as

$$
\|\mathbf{v}\|=\sqrt{(\mathbf{v}, \mathbf{v})}
$$

A normalized vector or unit vector is a vector whose norm is equal to 1 . A basis is said orthonormal if all vectors are normalized and mutually orthogonal.

A finite vector space with an inner product is called a Hilbert space. In order to an infinite vector space be a Hilbert space, it must obey additional properties besides having an inner product. Since we will deal primarily with finite vector spaces, we use the term Hilbert space as a synonym for vector space with an inner product. A subspace $W$ of a finite Hilbert space $V$ is also a Hilbert space. The set of vectors orthogonal to all vectors of $W$ is the Hilbert space $W^{\perp}$ called orthogonal complement. $V$ is the direct sum of $W$ and $W^{\perp}$, that is, $V=W \oplus W^{\perp}$. An $N$-dimensional Hilbert space will be denoted by $\mathcal{H}^{N}$ to highlight its dimension. A Hilbert space associated with a system $A$ will be denoted by $\mathcal{H}_{A}$.

## A. 3 The Dirac Notation

In this review of linear algebra, we will systematically be using the Dirac or bra-ket notation, which was introduced by the English physicist Paul Dirac in the context of quantum mechanics to aid algebraic manipulations. This notation is very simple. Several notations are used for vectors, such as $\mathbf{v}$ and $\vec{v}$. The Dirac notation uses

$$
\mathbf{v} \equiv|v\rangle
$$

Up to this point, instead of using bold or putting an arrow over letter $v$, we put letter $v$ between a vertical bar and a right angle bracket. If we have an indexed basis, that is, $\left\{\mathbf{v}_{1}, \ldots, \mathbf{v}_{n}\right\}$, in the Dirac notation we use the form $\left\{\left|v_{1}\right\rangle, \ldots,\left|v_{n}\right\rangle\right\}$ or $\{|1\rangle, \ldots,|n\rangle\}$. Note that if we are using a single basis, letter $\mathbf{v}$ is unnecessary in principle. Computer scientists usually start counting from 0 . So, the first basis vector is usually called $\mathbf{v}_{0}$. In the Dirac notation we have

$$
\mathbf{v}_{0} \equiv|0\rangle
$$

Vector $|0\rangle$ is not the zero vector, it is only the first vector in a collection of vectors. In the Dirac notation, the zero vector is an exception, whose notation is not modified. Here we use the notation $\mathbf{0}$.

Suppose that vector $|v\rangle$ has the following entries in a basis

$$
|v\rangle=\left[\begin{array}{c}
a_{1} \\
\vdots \\
a_{n}
\end{array}\right] .
$$

The dual vector is denoted by $\langle v|$ and is defined by

$$
\langle v|=\left[a_{1}^{*} \cdots a_{n}^{*}\right] .
$$

Usual vectors and their duals can be seen as column and row matrices, respectively, for algebraic manipulation. The matrix product of $\langle v|$ by $|v\rangle$ is denoted by $\langle v \mid v\rangle$ and its value in terms of their entries is

$$
\langle v \mid v\rangle=\sum_{i=1}^{n} a_{i}^{*} a_{i}
$$

This is an example of an inner product, which is naturally defined via the Dirac notation. If $\left\{\left|v_{1}\right\rangle, \ldots,\left|v_{n}\right\rangle\right\}$ is an orthonormal basis, then

$$
\left\langle v_{i} \mid v_{j}\right\rangle=\delta_{i j}
$$

where $\delta_{i j}$ is the Kronecker delta. The norm of a vector in this notation is

$$
\||v\rangle \|=\sqrt{\langle v \mid v\rangle}
$$

We use the terminology ket for vector $|v\rangle$ and bra for dual vector $\langle v|$. Keeping consistency, we use the terminology bra-ket for $\langle v \mid v\rangle$.

It is also very common to meet the matrix product of $|v\rangle$ by $\langle v|$, denoted by $|v\rangle\langle v|$, known as the outer product, whose result is an $n \times n$ matrix

$$
\begin{aligned}
|v\rangle\langle v| & =\left[\begin{array}{c}
a_{1} \\
\vdots \\
a_{n}
\end{array}\right] \cdot\left[\begin{array}{lll}
a_{1}^{*} & \cdots & a_{n}^{*}
\end{array}\right] \\
& =\left[\begin{array}{ccc}
a_{1} a_{1}^{*} & \cdots & a_{1} a_{n}^{*} \\
\ddots & \\
a_{n} a_{1}^{*} & \cdots & a_{n} a_{n}^{*}
\end{array}\right] .
\end{aligned}
$$

The key to the Dirac notation is to always view kets as column matrices, bras as row matrices, and recognize that a sequence of bras and kets is a matrix product, hence associative, but non-commutative.

## A. 4 Computational Basis

The computational basis of $\mathbb{C}^{n}$, denoted by $\{|0\rangle, \ldots,|n-1\rangle\}$, is given by

$$
|0\rangle=\left[\begin{array}{c}
1 \\
0 \\
\vdots \\
0
\end{array}\right], \ldots,|n-1\rangle=\left[\begin{array}{c}
0 \\
0 \\
\vdots \\
1
\end{array}\right] .
$$

This basis is also known as canonical basis. A few times we will use the numbering of the computational basis beginning with $|1\rangle$ and ending with $|n\rangle$. In this book, when we use a small-caption Latin letter within a ket or bra, we are referring to the computational basis. Then, the following expression will always be valid

$$
\langle i \mid j\rangle=\delta_{i j}
$$

The normalized sum of all computational basis vectors defines vector

$$
|\mathrm{D}\rangle=\frac{1}{\sqrt{n}} \sum_{i=0}^{n-1}|i\rangle
$$

which we will call diagonal state. When $n=2$, the diagonal state is given by $|\mathrm{D}\rangle=|+\rangle$ where

Exercise A.1. Explicitly calculate the values of $|i\rangle\langle j|$ and

$$
\sum_{i=0}^{n-1}|i\rangle\langle i|
$$

in $\mathbb{C}^{3}$.

## A. 5 Qubit and the Bloch Sphere

The qubit is a unit vector in vector space $\mathbb{C}^{2}$. A generic qubit $|\psi\rangle$ is represented by

$$
|\psi\rangle=\alpha|0\rangle+\beta|1\rangle,
$$

where coefficients $\alpha$ and $\beta$ are complex numbers and obey the constraint

$$
|\alpha|^{2}+|\beta|^{2}=1
$$

The set $\{|0\rangle,|1\rangle\}$ is the computational basis of $\mathbb{C}^{2}$ and $\alpha, \beta$ are called amplitudes of state $|\psi\rangle$. The term state (or state vector) is used as a synonym for unit vector in a Hilbert space.

Fig. A. 1 Bloch Sphere. The state $|\psi\rangle$ of a qubit is represented by a point on the sphere


In principle, we need four real numbers to describe a qubit, two for $\alpha$ and two for $\beta$. The constraint $|\alpha|^{2}+|\beta|^{2}=1$ reduces to three numbers. In quantum mechanics, two vectors that differ from a global phase factor are considered equivalent. A global phase factor is a complex number of unit modulus multiplying the state. By eliminating this factor, a qubit can be described by two real numbers $\theta$ and $\phi$ as follows:

$$
|\psi\rangle=\cos \frac{\theta}{2}|0\rangle+\mathrm{e}^{\mathrm{i} \phi} \sin \frac{\theta}{2}|1\rangle,
$$

where $0 \leq \theta \leq \pi$ and $0 \leq \phi<2 \pi$. In the above notation, state $|\psi\rangle$ can be represented by a point on the surface of a sphere of unit radius, called Bloch sphere. Numbers $\theta$ and $\phi$ are spherical angles that locate the point that describes $|\psi\rangle$, as shown in Fig. A.1. The vector showed there is given by

$$
\left[\begin{array}{c}
\sin \theta \cos \phi \\
\sin \theta \sin \phi \\
\cos \theta
\end{array}\right] .
$$

When we disregard global phase factors, there is a one-to-one correspondence between the quantum states of a qubit and the points on the Bloch sphere. State $|0\rangle$ is in the north pole of the sphere, because it is obtained by taking $\theta=0$. State $|1\rangle$ is in the south pole. States
are the intersection points of the $x$-axis and the sphere, and states $(|0\rangle \pm i|1\rangle) / \sqrt{2}$ are the intersection points of the $y$-axis with the sphere.

The representation of classical bits in this context is given by the poles of the Bloch sphere and the representation of the probabilistic classical bit, that is, 0 with probability $p$ and 1 with probability $1-p$, is given by the point in $z$-axis with coordinate $2 p-1$. The interior of the Bloch sphere is used to describe the states of a qubit in the presence of decoherence.

Exercise A.2. Using the Dirac notation, show that opposite points in the Bloch sphere correspond to orthogonal states.

Exercise A.3. Suppose you know that a qubit is either is in state $|+\rangle$ with probability $p$ or in state $|-\rangle$ with probability $1-p$. If this is the best you know about the qubit's state, where in the Bloch sphere would you represent this qubit?
Exercise A.4. Does the outside of Bloch sphere play any role?

## A. 6 Linear Operators

Let $V$ and $W$ be vector spaces, $\left\{\left|v_{1}\right\rangle, \ldots,\left|v_{n}\right\rangle\right\}$ a basis for $V$, and $\mathcal{A}$ a function $\mathcal{A}: V \mapsto W$ that satisfies

$$
\mathcal{A}\left(\sum_{i} a_{i}\left|v_{i}\right\rangle\right)=\sum_{i} a_{i} \mathcal{A}\left(\left|v_{i}\right\rangle\right)
$$

for any complex numbers $a_{i}$. $\mathcal{A}$ is called a linear operator from $V$ to $W$. The term linear operator in $V$ means that both the domain and codomain of $\mathcal{A}$ is $V$. The composition of linear operators $\mathcal{A}: V_{1} \mapsto V_{2}$ and $\mathcal{B}: V_{2} \mapsto V_{3}$ is also a linear operator $\mathcal{C}: V_{1} \mapsto V_{3}$ obtained through the composition of their functions: $\mathcal{C}(|v\rangle)=$ $\mathcal{B}(\mathcal{A}(|v\rangle)$. The sum of two linear operators, both from $V$ to $W$, is naturally defined by formula $(\mathcal{A}+\mathcal{B})(|v\rangle)=\mathcal{A}(|v\rangle)+\mathcal{B}(|v\rangle)$.

The identity operator $\mathcal{I}$ in $V$ is a linear operator such that $\mathcal{I}(|v\rangle)=|v\rangle$ for all $|v\rangle \in V$. The null operator $\mathcal{O}$ in $V$ is a linear operator such that $\mathcal{O}(|v\rangle)=\mathbf{0}$ for all $|v\rangle \in V$.

The rank of a linear operator $\mathcal{A}$ in $V$ is the dimension of the image of $\mathcal{A}$. The kernel or nullspace of a linear operator $\mathcal{A}$ in $V$ is the set of all vectors $|v\rangle$ for which $\mathcal{A}(|v\rangle)=\mathbf{0}$. The dimension of the kernel is called the nullity of the operator. The rank-nullity theorem states that $\operatorname{rank} \mathcal{A}+$ nullity $\mathcal{A}=\operatorname{dim} V$.

## Fact

If we specify the action of a linear operator on a basis of vector space $V$, its action on any vector in $V$ can be straightforwardly determined.

## A. 7 Matrix Representation

Linear operators can be represented by matrices. Let $\mathcal{A}: V \mapsto W$ be a linear operator, $\left\{\left|v_{1}\right\rangle, \ldots,\left|v_{n}\right\rangle\right\}$ and $\left\{\left|w_{1}\right\rangle, \ldots,\left|w_{m}\right\rangle\right\}$ orthonormal bases for $V$ and $W$, respectively. A matrix representation of $\mathcal{A}$ is obtained by applying $\mathcal{A}$ to every vector in the basis of $V$ and expressing the result as a linear combination of basis vectors of $W$, as follows:

$$
\mathcal{A}\left(\left|v_{j}\right\rangle\right)=\sum_{i=1}^{m} a_{i j}\left|w_{i}\right\rangle,
$$

where index $j$ run from 1 to $n$. Therefore, $a_{i j}$ are entries of an $m \times n$ matrix, which we call $A$. In this case, expression $\mathcal{A}\left(\left|v_{j}\right|\right)$, which means function $\mathcal{A}$ applied to argument $\left|v_{j}\right\rangle$, is equivalent to the matrix product $A\left|v_{j}\right\rangle$. Using the outer product notation, we have

$$
A=\sum_{i=1}^{m} \sum_{j=1}^{n} a_{i j}\left|w_{i}\right\rangle\left\langle v_{j}\right| .
$$

Using the above equation and the fact that the basis of $V$ is orthonormal, we can verify that the matrix product of $A$ by $\left|v_{j}\right\rangle$ is equal to $\mathcal{A}\left(\left|v_{j}\right|\right)$. The key to this calculation is to use the associativity of matrix multiplication:

$$
\begin{aligned}
\left(\left|w_{i}\right\rangle\left\langle v_{j}\right|\right)\left|v_{k}\right\rangle & =\left|w_{i}\right\rangle\left(\left\langle v_{j} \mid v_{k}\right\rangle\right) \\
& =\delta_{j k}\left|w_{i}\right\rangle
\end{aligned}
$$

In particular, the matrix representation of the identity operator $\mathcal{I}$ in any orthonormal basis is the identity matrix $I$ and the matrix representation of the null operator $\mathcal{O}$ in any orthonormal basis is the zero matrix.

If the linear operator $\mathcal{C}$ is the composition of the linear operators $\mathcal{B}$ and $\mathcal{A}$, the matrix representation of $\mathcal{C}$ will be obtained by multiplying the matrix representation of $\mathcal{B}$ with that of $\mathcal{A}$, that is, $C=B A$.

When we fix orthonormal bases for the vector spaces, there is a one-toone correspondence between linear operators and matrices. In $\mathbb{C}^{n}$, we use the computational basis as a reference basis, so the terms linear operator and matrix are taken as synonyms. We will also use the term operator as a synonym for linear operator.

Exercise A.5. Suppose $B$ is an operator whose action on the computational basis of the $n$-dimensional vector space $V$ is

$$
B|j\rangle=\left|\psi_{j}\right\rangle
$$

where $\left|\psi_{j}\right\rangle$ are vectors in $V$ for all $j$.

1. Obtain the expression of $B$ using the outer product.
2. Show that $\left|\psi_{j}\right\rangle$ is the $j$-th column in the matrix representation of $B$.
3. Suppose that $B$ is the Hadamard operator

$$
H=\frac{1}{\sqrt{2}}\left[\begin{array}{rr}
1 & 1 \\
1 & -1
\end{array}\right] .
$$

Redo the previous items using operator $H$.

## A. 8 Diagonal Representation

Let $\mathcal{O}$ be an operator in $V$. If there exists an orthonormal basis $\left\{\left|v_{1}\right\rangle, \ldots,\left|v_{n}\right\rangle\right\}$ of $V$ such that

$$
O=\sum_{i=1}^{n} \lambda_{i}\left|v_{i}\right\rangle\left\langle v_{i}\right|,
$$

we say that $\mathcal{O}$ admits a diagonal representation or, equivalently, $\mathcal{O}$ is diagonalizable. The complex numbers $\lambda_{i}$ are the eigenvalues of $\mathcal{O}$ and $\left|v_{i}\right\rangle$ are the corresponding eigenvectors. Any multiple of an eigenvector is also an eigenvector. If two eigenvectors are associated with the same eigenvalue, then any linear combination of these eigenvectors is an eigenvector. The number of linearly independent eigenvectors associated with the same eigenvalue is the multiplicity of that eigenvalue.

If there are eigenvalues with multiplicity greater than one, the diagonal representation can be factored out as follows:

$$
O=\sum_{\lambda} \lambda P_{\lambda},
$$

where index $\lambda$ runs only on the distinct eigenvalues and $P_{\lambda}$ is the projector on the eigenspace of $\mathcal{O}$ associated with eigenvalue $\lambda$. If $\lambda$ has multiplicity $1, P_{\lambda}=$ $|v\rangle\langle v|$, where $|v\rangle$ is the unit eigenvector associated with $\lambda$. If $\lambda$ has multiplicity 2 and $\left|v_{1}\right\rangle,\left|v_{2}\right\rangle$ are linearly independent unit eigenvectors associated with $\lambda, P_{\lambda}=$ $\left|v_{1}\right\rangle\left\langle v_{1}\right|+\left|v_{2}\right\rangle\left\langle v_{2}\right|$ and so on. The projectors $P_{\lambda}$ satisfy

$$
\sum_{\lambda} P_{\lambda}=I
$$

An alternative way to define a diagonalizable operator is by requiring that $O$ is similar to a diagonal matrix. Matrices $O$ and $O^{\prime}$ are similar if $O^{\prime}=M^{-1} O M$ for some invertible matrix $M$. We have interest only in the case when $M$ is a unitary matrix. The term diagonalizable we use here is narrower than the one used in the literature, because we are demanding that $M$ be a unitary matrix.

Exercise A.6. Suppose that $O$ is a diagonalizable operator with eigenvalues $\pm 1$. Show that

$$
P_{ \pm 1}=\frac{I \pm O}{2}
$$

## A. 9 Completeness Relation

The completeness relation is so useful that it deserves to be highlighted. Let $\left\{\left|v_{1}\right\rangle\right.$, $\left.\ldots,\left|v_{n}\right\rangle\right\}$ be an orthonormal basis of $V$. Then,

$$
I=\sum_{i=1}^{n}\left|v_{i}\right\rangle\left\langle v_{i}\right| .
$$

The completeness relation is the diagonal representation of the identity matrix.
Exercise A.7. If $\left\{\left|v_{1}\right\rangle, \ldots,\left|v_{n}\right\rangle\right\}$ is an orthonormal basis, it is straightforward to verify the validity of equations

$$
A\left|v_{j}\right\rangle=\sum_{i=1}^{m} a_{i j}\left|w_{i}\right\rangle
$$

from equation

$$
A=\sum_{i=1}^{m} \sum_{j=1}^{n} a_{i j}\left|w_{i}\right\rangle\left\langle v_{j}\right| .
$$

Verify in the reverse direction using the completeness relation, that is, assuming that expressions $A\left|v_{j}\right\rangle$ are given for all $1 \leq j \leq n$, obtain $A$.

## A. 10 Cauchy-Schwarz Inequality

Let $V$ be a Hilbert space and $|v\rangle,|w\rangle \in V$. Then,

$$
|\langle v \mid w\rangle| \leq \sqrt{\langle v \mid v\rangle\langle w \mid w\rangle}
$$

A more explicit way of presenting the Cauchy-Schwarz inequality is

$$
\left|\sum_{i} v_{i} w_{i}\right|^{2} \leq\left(\sum_{i}\left|v_{i}\right|^{2}\right)\left(\sum_{i}\left|w_{i}\right|^{2}\right),
$$

which is obtained when we take $|v\rangle=\sum_{i} v_{i}^{*}|i\rangle$ and $|w\rangle=\sum_{i} w_{i}|i\rangle$.

## A. 11 Special Operators

Let $A$ be a linear operator in Hilbert space $V$. Then, there exists a unique linear operator $A^{\dagger}$ in $V$, called adjoint operator, that satisfies

$$
(|v\rangle, A|w\rangle)=\left(A^{\dagger}|v\rangle,|w\rangle\right),
$$

for all $|v\rangle,|w\rangle \in V$.
The matrix representation of $A^{\dagger}$ is the transpose-conjugate matrix $\left(A^{*}\right)^{\mathrm{T}}$. The main properties of the dagger or transpose-conjugate operation are

1. $(A B)^{\dagger}=B^{\dagger} A^{\dagger}$
2. $|v\rangle^{\dagger}=\langle v|$
3. $(A|v\rangle)^{\dagger}=\langle v| A^{\dagger}$
4. $(|w\rangle\langle v|)^{\dagger}=|v\rangle\langle w|$
5. $\left(A^{\dagger}\right)^{\dagger}=A$
6. $\left(\sum_{i} a_{i} A_{i}\right)^{\dagger}=\sum_{i} a_{i}^{*} A_{i}^{\dagger}$

The last property shows that the dagger operation is conjugate-linear when applied on a linear combination of operators.

## Normal Operator

An operator $A$ in $V$ is normal if $A^{\dagger} A=A A^{\dagger}$.

## Spectral Theorem

An operator $A$ in $V$ is diagonalizable if and only if $A$ is normal.

## Unitary Operator

An operator $U$ in $V$ is unitary if $U^{\dagger} U=U U^{\dagger}=I$.

## Facts About Unitary Operators

Unitary operators are normal, so they are diagonalizable with respect to an orthonormal basis. Eigenvectors of a unitary operator associated with different eigenvalues are orthogonal. The eigenvalues have unit modulus, that is, their form is $\mathrm{e}^{\mathrm{i} \alpha}$, where $\alpha$ is a real number. Unitary operators preserve the inner product, that is, the inner product of $U\left|v_{1}\right\rangle$ by $U\left|v_{2}\right\rangle$ is equal to the inner product of $\left|v_{1}\right\rangle$ by $\left|v_{2}\right\rangle$. The application of a unitary operator on a vector preserves its norm.

## Hermitian Operator

An operator $A$ in $V$ is Hermitian or self-adjoint if $A^{\dagger}=A$.

## Facts About Hermitian Operators

Hermitian operators are normal, so they are diagonalizable with respect to an orthonormal basis. Eigenvectors of a Hermitian operator associated with different eigenvalues are orthogonal. The eigenvalues of a Hermitian operator are real numbers. A real symmetric matrix is Hermitian.

## Orthogonal Projector

An operator $P$ in $V$ is an orthogonal projector if $P^{2}=P$ and $P^{\dagger}=P$.

## Facts About Orthogonal Projectors

The eigenvalues are equal to 0 or 1 . If $P$ is an orthogonal projector, then the orthogonal complement $I-P$ is also an orthogonal projector. Applying a projector on a vector either decreases its norm or maintains invariant. In this book, we use the term projector as a synonym for orthogonal projector. We will use the term non-orthogonal projector explicitly to distinguish this case. An example of a nonorthogonal projector on a qubit is $P=|1\rangle\langle+|$.

## Positive Operator

An operator $A$ in $V$ is said positive if $\langle v| A|v\rangle \geq 0$ for any $|v\rangle \in V$. If the inequality is strict for any nonzero vector in $V$, then the operator is said positive definite.

## Facts About Positive Operators

Positive operators are Hermitian. The eigenvalues are nonnegative real numbers.
Exercise A.8. Consider matrix

$$
M=\left[\begin{array}{ll}
1 & 0 \\
1 & 1
\end{array}\right]
$$

1. Show that $M$ is not normal.
2. Show that the eigenvectors of $M$ generate a one-dimensional space.

Exercise A.9. Consider matrix

$$
M=\left[\begin{array}{cc}
1 & 0 \\
1 & -1
\end{array}\right]
$$

1. Show that the eigenvalues of $M$ are $\pm 1$.
2. Show that $M$ is neither unitary nor Hermitian.
3. Show that the eigenvectors associated with distinct eigenvalues of $M$ are not orthogonal.
4. Show that $M$ has a diagonal representation.

## Exercise A.10.

1. Show that the product of two unitary operators is a unitary operator.
2. The sum of two unitary operators is necessarily a unitary operator? If not, give a counterexample.

## Exercise A.11.

1. Show that the sum of two Hermitian operators is a Hermitian operator.
2. The product of two Hermitian operators is necessarily a Hermitian operator? If not, give a counterexample.

Exercise A.12. Show that $A^{\dagger} A$ is a positive operator for any operator $A$.

## A. 12 Pauli Matrices

The Pauli matrices are

$$
\begin{aligned}
& \sigma_{0}=I=\left[\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right], \\
& \sigma_{1}=\sigma_{x}=X=\left[\begin{array}{ll}
0 & 1 \\
1 & 0
\end{array}\right], \\
& \sigma_{2}=\sigma_{y}=Y=\left[\begin{array}{cc}
0 & -\mathrm{i} \\
\mathrm{i} & 0
\end{array}\right], \\
& \sigma_{3}=\sigma_{z}=Z=\left[\begin{array}{cc}
1 & 0 \\
0 & -1
\end{array}\right] .
\end{aligned}
$$

These matrices are unitary and Hermitian, and hence their eigenvalues are equal to $\pm 1$. Putting in another way: $\sigma_{j}^{2}=I$ and $\sigma_{j}^{\dagger}=\sigma_{j}$ for $j=0, \ldots, 3$.

The following facts are extensively used:

$$
\begin{aligned}
X|0\rangle & =|1\rangle, \quad X|1\rangle
\end{aligned}=|0\rangle, ~ 子|0\rangle=|1\rangle=-|1\rangle .
$$

Pauli matrices form a basis for the vector space of $2 \times 2$ matrices. Therefore, a generic operator that acts on a qubit can be written as a linear combination of Pauli matrices.

Exercise A.13. Consider the representation of the state $|\psi\rangle$ of a qubit in the Bloch sphere. What is the representation of states $X|\psi\rangle, Y|\psi\rangle$, and $Z|\psi\rangle$ relative to $|\psi\rangle$ ? What is the geometric interpretation of the action of the Pauli matrices on the Bloch sphere?

## A. 13 Operator Functions

If we have an operator $A$ in $V$, we can ask whether it is possible to calculate $\sqrt{A}$, that is, to find an operator the square of which is $A$ ? In general, we can ask ourselves whether it makes sense to use an operator as an argument of a usual function, such as, exponential or logarithmic function. If operator $A$ is normal, it has a diagonal representation, that is, can be written in the form

$$
A=\sum_{i} a_{i}\left|v_{i}\right\rangle\left\langle v_{i}\right|
$$

where $a_{i}$ are the eigenvalues and the set $\left\{\left|v_{i}\right\rangle\right\}$ is an orthonormal basis of eigenvectors of $A$. We can extend the application of a function $f: \mathbb{C} \mapsto \mathbb{C}$ to $A$ as follows

$$
f(A)=\sum_{i} f\left(a_{i}\right)\left|v_{i}\right\rangle\left\langle v_{i}\right| .
$$

The result is an operator defined in the same vector space $V$ and it is independent of the choice of basis of $V$.

If the goal is to calculate $\sqrt{A}$, first $A$ must be diagonalized, that is, we must determine a unitary matrix $U$ such that $A=U D U^{\dagger}$, where $D$ is a diagonal matrix. Then, we use the fact that $\sqrt{A}=U \sqrt{D} U^{\dagger}$, where $\sqrt{D}$ is calculated by taking the square root of each diagonal element.

If $U$ is the evolution operator of an isolated quantum system that is initially in state $|\psi(0)\rangle$, the state at time $t$ is given by

$$
|\psi(t)\rangle=U^{t}|\psi(0)\rangle .
$$

The most efficient way to calculate state $|\psi(t)\rangle$ is to obtain the diagonal representation of the unitary operator $U$

$$
U=\sum_{i} \lambda_{i}\left|v_{i}\right\rangle\left\langle v_{i}\right|,
$$

and to calculate the $t$-th power $U$, that is,

$$
U^{t}=\sum_{i} \lambda_{i}^{t}\left|v_{i}\right\rangle\left\langle v_{i}\right| .
$$

The system state at time $t$ will be

$$
|\psi(t)\rangle=\sum_{i} \lambda_{i}^{t}\left\langle v_{i} \mid \psi(0)\right\rangle\left|v_{i}\right\rangle .
$$

The trace of a matrix is another type of operator function. In this case, the result of applying the trace function is a complex number defined as

$$
\operatorname{tr}(A)=\sum_{i} a_{i i}
$$

where $a_{i i}$ are the diagonal elements of $A$. In the Dirac notation

$$
\operatorname{tr}(A)=\sum_{i}\left\langle v_{i}\right| A\left|v_{i}\right\rangle
$$

where $\left\{\left|v_{1}\right\rangle, \ldots,\left|v_{n}\right\rangle\right\}$ is an orthonormal basis of $V$. The trace function satisfies the following properties:

1. $\operatorname{tr}(a A+b B)=a \operatorname{tr}(A)+b \operatorname{tr}(B)$, (linearity)
2. $\operatorname{tr}(A B)=\operatorname{tr}(B A)$,
3. $\operatorname{tr}(A B C)=\operatorname{tr}(C A B)$. (cyclic property)

The third property follows from the second one. Properties 2 and 3 are valid even when $A, B$, and $C$ are not square matrices.

The trace function is invariant under similarity transformations, that is, $\operatorname{tr}\left(M^{-1} A M\right)=\operatorname{tr}(A)$, where $M$ is an invertible matrix. This implies that the trace does not depend on the basis choice for the matrix representation of $A$.

A useful formula involving the trace of operators is

$$
\operatorname{tr}(A|\psi\rangle\langle\psi|)=\langle\psi| A|\psi\rangle
$$

for any $|\psi\rangle \in V$ and any $A$ in $V$. This formula can be easily proved using the cyclic property of the trace function.

Exercise A.14. Using the method of applying functions on matrices described in this section, find all matrices $M$ such that

$$
M^{2}=\left[\begin{array}{ll}
5 & 4 \\
4 & 5
\end{array}\right]
$$

## A. 14 Tensor Product

Let $V$ and $W$ be finite Hilbert spaces with basis $\left\{\left|v_{1}\right\rangle, \ldots,\left|v_{m}\right\rangle\right\}$ and $\left\{\left|w_{1}\right\rangle, \ldots\right.$, $\left.\left|w_{n}\right\rangle\right\}$, respectively. The tensor product of $V$ by $W$, denoted by $V \otimes W$, is an $m n-$ dimensional Hilbert space, for which set $\left\{\left|v_{1}\right\rangle \otimes\left|w_{1}\right\rangle,\left|v_{1}\right\rangle \otimes\left|w_{2}\right\rangle, \ldots,\left|v_{m}\right\rangle \otimes\left|w_{n}\right\rangle\right\}$ is a basis. The tensor product of a vector in $V$ by a vector in $W$, such as $|v\rangle \otimes|w\rangle$, also denoted by $|v\rangle|w\rangle$ or $|v, w\rangle$ or $|v w\rangle$, can be calculated explicitly via the Kronecker product, defined ahead. A generic vector in $V \otimes W$ is a linear combination of vectors $\left|v_{i}\right\rangle \otimes\left|w_{j}\right\rangle$, that is, if $|\psi\rangle \in V \otimes W$ then

$$
|\psi\rangle=\sum_{i=1}^{m} \sum_{j=1}^{n} a_{i j}\left|v_{i}\right\rangle \otimes\left|w_{j}\right\rangle
$$

The tensor product is bilinear, that is, linear in each argument:

1. $|v\rangle \otimes\left(a\left|w_{1}\right\rangle+b\left|w_{2}\right\rangle\right)=a|v\rangle \otimes\left|w_{1}\right\rangle+b|v\rangle \otimes\left|w_{2}\right\rangle$,
2. $\left(a\left|v_{1}\right\rangle+b\left|v_{2}\right\rangle\right) \otimes|w\rangle=a\left|v_{1}\right\rangle \otimes|w\rangle+b\left|v_{2}\right\rangle \otimes|w\rangle$.

A scalar can always be factored out to the beginning of the expression:

$$
a(|v\rangle \otimes|w\rangle)=(a|v\rangle) \otimes|w\rangle=|v\rangle \otimes(a|w\rangle)
$$

The tensor product of a linear operator $A$ in $V$ by $B$ in $W$, denoted by $A \otimes B$, is a linear operator in $V \otimes W$ defined by

$$
(A \otimes B)(|v\rangle \otimes|w\rangle)=(A|v\rangle) \otimes(B|w\rangle)
$$

A generic linear operator in $V \otimes W$ can be written as a linear combination of operators of the form $A \otimes B$, but an operator in $V \otimes W$ cannot be factored out in general. This definition can easily be extended to operators $A: V \mapsto V^{\prime}$ and $B: W \mapsto W^{\prime}$. In this case, the tensor product of these operators is of type $(A \otimes B):(V \otimes W) \mapsto\left(V^{\prime} \otimes W^{\prime}\right)$.

In quantum mechanics, it is very common to use operators in the form of external products, for example, $A=|v\rangle\langle v|$ and $B=|w\rangle\langle w|$. The tensor product of $A$ by $B$ can be represented by the following equivalent ways:

$$
\begin{aligned}
A \otimes B & =(|v\rangle\langle v|) \otimes(|w\rangle\langle w|) \\
& =|v\rangle\langle v| \otimes|w\rangle\langle w| \\
& =|v, w\rangle\langle v, w| .
\end{aligned}
$$

If $A_{1}, A_{2}$ are operators in $V$ and $B_{1}, B_{2}$ are operators in $W$, then the composition or the matrix product of the matrix representations obey the property

$$
\left(A_{1} \otimes B_{1}\right) \cdot\left(A_{2} \otimes B_{2}\right)=\left(A_{1} \cdot A_{2}\right) \otimes\left(B_{1} \cdot B_{2}\right)
$$

The inner product of $\left|v_{1}\right\rangle \otimes\left|w_{1}\right\rangle$ by $\left|v_{2}\right\rangle \otimes\left|w_{2}\right\rangle$ is defined as

$$
\left(\left|v_{1}\right\rangle \otimes\left|w_{1}\right\rangle,\left|v_{2}\right\rangle \otimes\left|w_{2}\right\rangle\right)=\left\langle v_{1} \mid v_{2}\right\rangle\left\langle w_{1} \mid w_{2}\right\rangle .
$$

The inner product of vectors written as a linear combination of basis vectors are calculated by applying the linear property in the second argument and the conjugatelinear property in the first argument of the inner product. For example,

$$
\left(\left(\sum_{i=1}^{n} a_{i}\left|v_{i}\right\rangle\right) \otimes\left|w_{1}\right\rangle,|v\rangle \otimes\left|w_{2}\right\rangle\right)=\left(\sum_{i=1}^{n} a_{i}^{*}\left\langle v_{i} \mid v\right\rangle\right)\left\langle w_{1} \mid w_{2}\right\rangle .
$$

The inner product definition implies that

$$
\||v\rangle \otimes|w\rangle\|=\||v\rangle\|\cdot\||w\rangle \| .
$$

In particular, the norm of the tensor product of unit-norm vectors is a unit-norm vector.

When we use matrix representations for operators, the tensor product can be calculated explicitly via the Kronecker product. Let $A$ be a $m \times n$ matrix and $B$ a $p \times q$ matrix. Then,

$$
A \otimes B=\left[\begin{array}{ccc}
a_{11} B & \cdots & a_{1 n} B \\
\ddots & \\
a_{m 1} B & \cdots & a_{m n} B
\end{array}\right]
$$

The dimension of the resulting matrix is $m p \times n q$. The Kronecker product can be used for matrices of any dimension, particularly for two vectors,

$$
\left[\begin{array}{l}
a_{1} \\
a_{2}
\end{array}\right] \otimes\left[\begin{array}{l}
b_{1} \\
b_{2}
\end{array}\right]=\left[\begin{array}{l}
a_{1}\left[\begin{array}{l}
b_{1} \\
b_{2}
\end{array}\right] \\
a_{2}\left[\begin{array}{l}
b_{1} \\
b_{2}
\end{array}\right]
\end{array}\right]=\left[\begin{array}{l}
a_{1} b_{1} \\
a_{1} b_{2} \\
a_{2} b_{1} \\
a_{2} b_{2}
\end{array}\right] .
$$

The tensor product is an associative and distributive operation, but noncommutative, that is, $|v\rangle \otimes|w\rangle \neq|w\rangle \otimes|v\rangle$ if $v \neq w$. Most operations on a tensor product are performed term by term, such as

$$
(A \otimes B)^{\dagger}=A^{\dagger} \otimes B^{\dagger}
$$

If both operators $A$ and $B$ are special operators of the same type, as the ones defined in Sect. A.11, then the tensor product $A \otimes B$ is also a special operator of the same type. For example, the tensor product of Hermitian operators is a Hermitian operator.

The trace of a Kronecker product of matrices is

$$
\operatorname{tr}(A \otimes B)=\operatorname{tr} A \operatorname{tr} B
$$

while the determinant is

$$
\operatorname{det}(A \otimes B)=(\operatorname{det} A)^{m}(\operatorname{det} B)^{n}
$$

where $n$ is the dimension of $A$ and $m$ of $B$.
The direct sum of a vector space $V$ with itself $n$ times is a particular case of the tensor product. In fact, a matrix $A \oplus \cdots \oplus A$ in $V \oplus \cdots \oplus V$ is equal to $I \otimes A$ for any $A$ in $V$, where $I$ is the $n \times n$ identity matrix. This shows that, somehow, the tensor product is defined from the direct sum of vector spaces, analogous to the product of numbers which is defined from the sum of numbers. However, the tensor product is richer than the simple repetition of the direct sum of vector spaces. Anyway, we can continue generalizing definitions: It is natural to define tensor potentiation, in fact, $V^{\otimes n}$ means $V \otimes \cdots \otimes V$ with $n$ terms.

If the diagonal state of the vector space $V$ is $|\mathrm{D}\rangle_{V}$ and of space $W$ is $|\mathrm{D}\rangle_{W}$, then the diagonal state of space $V \otimes W$ is $|\mathrm{D}\rangle_{V} \otimes|\mathrm{D}\rangle_{W}$. Therefore, the diagonal state of space $V^{\otimes n}$ is $|\mathrm{D}\rangle^{\otimes n}$.

Exercise A.15. Let $H$ be the Hadamard operator

$$
H=\frac{1}{\sqrt{2}}\left[\begin{array}{rr}
1 & 1 \\
1 & -1
\end{array}\right]
$$

Show that

$$
\langle i| H^{\otimes n}|j\rangle=\frac{(-1)^{i \cdot j}}{\sqrt{2^{n}}}
$$

where $n$ represents the number of qubits and $i \cdot j$ is the binary inner product, that is, $i \cdot j=i_{1} j_{1}+\cdots+i_{n} j_{n} \bmod 2$, where $\left(i_{1}, \ldots, i_{n}\right)$ and $\left(j_{1}, \ldots, j_{n}\right)$ are the binary decompositions of $i$ and $j$, respectively.

## A. 15 Registers

A register is a set of qubits treated as a composite system. In many quantum algorithms, the qubits are divided into two registers: one for the main calculation from where the result comes out and the other for the draft (calculations that will be erased). Suppose we have a register with two qubits. The computational basis is

$$
|0,0\rangle=\left[\begin{array}{l}
1 \\
0 \\
0 \\
0
\end{array}\right] \quad|0,1\rangle=\left[\begin{array}{l}
0 \\
1 \\
0 \\
0
\end{array}\right] \quad|1,0\rangle=\left[\begin{array}{l}
0 \\
0 \\
1 \\
0
\end{array}\right] \quad|1,1\rangle=\left[\begin{array}{l}
0 \\
0 \\
0 \\
1
\end{array}\right] .
$$

A generic state of this register is

$$
|\psi\rangle=\sum_{i=0}^{1} \sum_{j=0}^{1} a_{i j}|i, j\rangle
$$

where coefficients $a_{i j}$ are complex numbers that satisfy the constraint

$$
\left|a_{00}\right|^{2}+\left|a_{01}\right|^{2}+\left|a_{10}\right|^{2}+\left|a_{11}\right|^{2}=1
$$

To help generalizing to $n$ qubits, it is usual to compress the notation by converting binary-base representation to decimal-base. The computational basis for two-qubit register in decimal-base representation is $\{|0\rangle,|1\rangle,|2\rangle,|3\rangle\}$. In the binary-base representation, we can determine the number of qubits by counting the number of digits inside the ket, for example, $|011\rangle$ refers to three qubits. In the decimal-base representation, we cannot determine what is the number of qubits of the register. This information should come implicit. In this case, we can go back, write the numbers in the binary-base representation and explicitly retrieve the notation. In the compact notation, a generic state of a $n$-qubit register is

$$
|\psi\rangle=\sum_{i=0}^{2^{n}-1} a_{i}|i\rangle
$$

where coefficients $a_{i}$ are complex numbers that satisfy the constraint

$$
\sum_{i=0}^{2^{n}-1}\left|a_{i}\right|^{2}=1
$$

The diagonal state of a $n$-qubit register is the tensor product of the diagonal state of each qubit, that is, $|\mathrm{D}\rangle=|+\rangle^{\otimes n}$.
Exercise A.16. Let $f$ be a function with domain $\{0,1\}^{n}$ and codomain $\{0,1\}^{m}$. Consider a 2-register quantum computer with $n$ and $m$ qubits, respectively. Function $f$ can be implemented by using operator $U_{f}$ defined in the following way:

$$
U_{f}|x\rangle|y\rangle=|x\rangle|y \oplus f(x)\rangle,
$$

where $x$ has $n$ bits, $y$ has $m$ bits, and $\oplus$ is the binary sum (bitwise xor).

1. Show that $U_{f}$ is a unitary operator for any $f$.
2. If $n=m$ and $f$ is injective, show that $f$ can be implemented on a 1-register quantum computer with $n$ qubits.

## Further Reading

There are many good books about linear algebra. For an initial contact, we suggest [11, 12, 37, 72]; for a more advanced approach, we suggest [36]; for those who have mastered the basics and are only interested in the application of linear algebra on quantum computation, we suggest [64].

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