

# Math for Quantum

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## 1 The complex numbers

By definition, we let  $i^2 = -1$ , and we define  $\mathbb{C}$  to be all numbers of the form  $a + ib$ , where  $a, b \in \mathbb{R}$ . The number  $a$  is called the **real part** of  $a + ib$ , and the number  $b$  is called the **imaginary part** of  $b$ . That is,  $\text{Re}(a + ib) = a$  and  $\text{Im}(a + ib) = b$ . We then define addition and multiplication on  $\mathbb{C}$  by

$$\begin{aligned} (a + ib) + (c + id) &= (a + c) + i(b + d), \\ (a + ib) \cdot (c + id) &= (ac - bd) + i(bc + ad). \end{aligned}$$

We define **complex conjugation** by

$$\overline{a + bi} = a - bi.$$

Note that  $\frac{1}{2}(z + \bar{z}) = \text{Re}(z)$  and  $\frac{1}{2}(z - \bar{z}) = \text{Im}(z)$ .

**Fact.** Addition and multiplication on  $\mathbb{C}$  are commutative and associative, and multiplication distributes over addition. Every nonzero complex number  $z$  has a unique multiplicative inverse, i.e. a complex number  $z^{-1}$  such that  $zz^{-1} = z^{-1}z = 1$ .

**Exercise 1.** Show that  $\bar{\bar{z}} = z$ , and  $\overline{z_1 + z_2} = \bar{z}_1 + \bar{z}_2$ , and  $\overline{z_1 \cdot z_2} = \bar{z}_1 \cdot \bar{z}_2$ .

For a complex number  $z$ , we say that  $z \in \mathbb{R}$  when the imaginary part of  $z$  is zero.

**Exercise 2.** Show that  $\bar{z} = z$  iff  $z \in \mathbb{R}$ , and  $\bar{z} = -z$  iff  $z = ib$  for some  $b \in \mathbb{R}$ .

**Exercise 3.** Show that  $\bar{z}z = a^2 + b^2$  when  $z = a + ib$ .

**Definition.** We define the **modulus** of a complex number by  $|z| = \sqrt{z\bar{z}}$ .

**Exercise 4.** Show that  $|z_1 \cdot z_2| = |z_1| \cdot |z_2|$ . Show that  $|z| = 0$  iff  $z = 0$ .

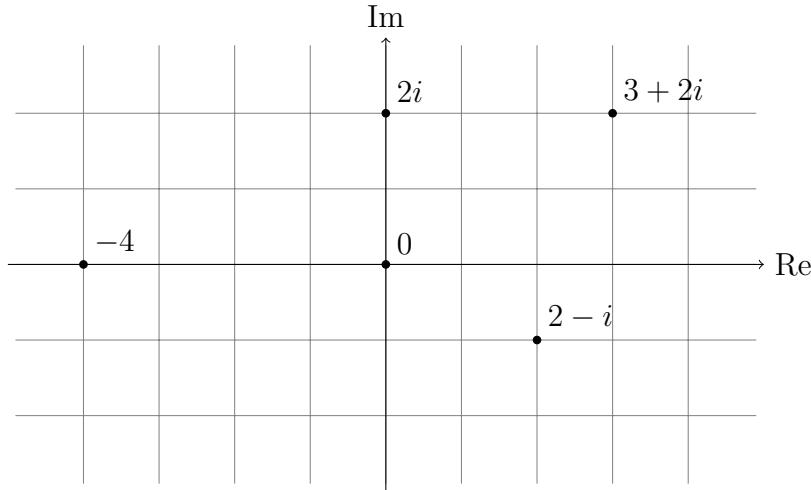
For any complex number  $z = a + bi$ , if we define  $\theta = \tan(\frac{b}{a})^{-1}$ , then

$$z = |z| (\cos \theta + i \sin \theta).$$

This is referred to as the **polar representation** of  $z$ , and  $\theta$  is called the **phase** of  $z$ . Defining  $r = |z|$  and  $e^{i\theta} = \cos \theta + i \sin \theta$ , we can write  $z = re^{i\theta}$ . In polar form, multiplication of complex numbers has the following convenient representation:

$$(r_1 e^{i\theta_1}) \cdot (r_2 e^{i\theta_2}) = (r_1 r_2) e^{i(\theta_1 + \theta_2)}.$$

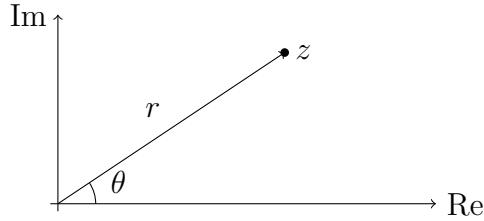
It is often convenient to represent the complex numbers on a plane, with the real numbers as the horizontal axis and the imaginary numbers as the vertical axis. In this case, the modulus  $r$  of  $re^{i\theta}$  corresponds to its distance from the origin, and the phase  $\theta$  corresponds to the angle that  $re^{i\theta}$  makes with the real axis. What's more, we have  $\overline{re^{i\theta}} = re^{i(-\theta)}$ , so that complex conjugation corresponds to reflection in the real axis.



**Definition.** Let  $\mathbb{T} = \{z \in \mathbb{C} \mid |z| = 1\}$ . We say that  $\mathbb{T}$  is the **complex unit circle**.

Each unit-length complex number is of the form  $e^{i\theta} = \cos \theta + i \sin \theta$ , and multiplication on  $\mathbb{T}$  can be represented as  $e^{i\theta_1} e^{i\theta_2} = e^{i(\theta_1 + \theta_2)}$ . Hence  $(e^{i\theta})^{-1} = e^{-i\theta}$ . We also have the famous **Euler identity**:

$$e^{i\pi} = \cos \pi + i \sin \pi = -1.$$



## 2 Vector spaces

A **complex vector space** is a set  $V$ , equipped with

- a special **zero** element  $\mathbf{0}$ ,
- an **addition** operation  $+$ :  $V \times V \rightarrow V$ , and
- a **scalar multiplication** operation  $\cdot$ :  $\mathbb{C} \times V \rightarrow V$ ,

such that for any  $u, v, w \in V$  and  $\alpha, \beta \in \mathbb{C}$ ,

$$\begin{aligned}
 u + v &= v + u \\
 u + (v + w) &= (u + v) + w \\
 v + \mathbf{0} &= v \\
 0 \cdot v &= \mathbf{0} \\
 1 \cdot v &= v \\
 \alpha \cdot (\beta \cdot v) &= (\alpha\beta) \cdot v \\
 \alpha \cdot (u + v) &= \alpha \cdot u + \alpha \cdot v \\
 (\alpha + \beta) \cdot v &= \alpha \cdot v + \beta \cdot v
 \end{aligned}$$

We will henceforth abbreviate scalar multiplication  $\alpha \cdot v$  by  $\alpha v$ .

**Exercise 5.** Show that  $v + (-1)v = \mathbf{0}$ .

**Definition.** A **basis** for  $V$  is a collection of vectors  $v_i \in V$  such that any  $v \in V$  can be expressed uniquely in the form  $v = \sum_i \alpha_i v_i$ . The **dimension** of vector space  $V$  is the smallest number  $n$  such that there exists a basis consisting of  $n$  vectors. We write  $\dim V = n$ .

**Example.**

**Definition.** A **subspace** of a vector space  $V$  is a nonempty subset  $W \subseteq V$  that is closed under addition and scalar multiplication. In other words,  $W$  is itself a vector space inside of  $V$  such that  $\dim W \leq \dim V$ . In the finite-dimensional case,  $\dim W = \dim V$  only if  $W = V$ .

**Definition.** Given a complex vector space  $V$ , an **inner product** is a binary operation  $\langle -, - \rangle : V \times V \rightarrow \mathbb{C}$  such that for any  $u, v, w \in V$  and  $c \in \mathbb{C}$ ,

$$\begin{aligned}\langle u, v \rangle &= \overline{\langle v, u \rangle} \\ \langle u, v + w \rangle &= \langle u, v \rangle + \langle u, w \rangle \\ \langle u, cv \rangle &= c\langle u, v \rangle \\ \langle u, u \rangle &\geq 0\end{aligned}$$

The **norm** of a vector  $v$  is defined to be  $\|v\| = \sqrt{\langle v, v \rangle}$ .

**Proposition 1** (Cauchy-Schwartz inequality).  $|\langle u, v \rangle| \leq \langle u, u \rangle \cdot \langle v, v \rangle$ .

**Definition.** Two vectors  $u, v \in V$  are said to be **orthogonal** if  $\langle u, v \rangle = 0$ .

**Definition.** An **orthonormal** basis for  $V$  is a basis consisting entirely of vectors of norm 1, which are all orthogonal to one another.

**Definition.** A **finite-dimensional Hilbert space** is a finite-dimensional complex vector space equipped with an inner product.

We use elements of a Hilbert space  $H$  to represent the so-called “pure” states of a quantum system. Unlike in the classical case, one cannot determine whether or not the system is in the state represented by  $v$  through a single measurement. The best one can do is find a measurement setup such that for any state  $v$ , the probability of a positive result is given by

$$\frac{|\langle v, w \rangle|^2}{\|v\|^2 \|w\|^2} \tag{1}$$

Hence, for any  $\alpha \in \mathbb{C}$  and  $\psi \in H$ , the states represented by  $\psi$  and  $\alpha\psi$  cannot be discriminated from one another; we therefore take them to be the same state. We use this freedom to stipulate that only the **unit norm** elements of a Hilbert space (those such that  $\|v\| = 1$ ) will be used to represent states. Under this stipulation, the probability of a positive result in the “ $\psi$ -test” is then simply given by  $|\langle \phi, \psi \rangle|^2$ . Note that even with this stipulation, there are still multiple elements of  $H$  representing the same state: e.g.,  $v$  and  $iv$ .

**Example.** Consider the two-dimensional vector space  $V$ . We can represent elements of  $V$  by ordered pairs (or by column vectors), and we can define an inner product on  $V$  by

$$\langle (a, b), (c, d) \rangle = a^*c + b^*d.$$

In this case, the pair  $(1, 0)$  and  $(0, 1)$  form an orthonormal basis for  $V$ .

### 3 Linear operators

**Definition.** Let  $V$  and  $W$  be vector spaces over  $\mathbb{C}$ , and let  $A : V \rightarrow W$  be a function. We say that  $A$  is **linear** just in case  $A(v + w) = Av + Aw$  and  $A(cv) = cAv$ , for all  $v, w \in V$  and all  $c \in \mathbb{C}$ . For the case where  $A : V \rightarrow V$ , we say that  $A$  is a **linear operator**.

You may well have seen linear operators before, but most likely under the name **matrices**. That's because matrices provide a particularly convenient way for representing linear operators on finite-dimensional vector spaces. Here an  $n \times n$  matrix (over  $\mathbb{C}$ ) is an array of  $n^2$  complex numbers. For the most part, it will suffice for us to look at the simplest non-trivial case, i.e.  $2 \times 2$  matrices. Let  $A$  be the matrix

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}.$$

Then the action of  $A$  on a vector  $\begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$  can be computed as follows:

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \begin{pmatrix} a_{11}b_1 + a_{12}b_2 \\ a_{21}b_1 + a_{22}b_2 \end{pmatrix}.$$

We've already defined a product of linear operators in terms of composition. You can then verify (by a straightforward calculation) that this product yields the standard matrix multiplication rule:

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix} = \begin{pmatrix} a_{11}b_{11} + a_{12}b_{21} & a_{11}b_{12} + a_{12}b_{22} \\ a_{21}b_{11} + a_{22}b_{21} & a_{21}b_{12} + a_{22}b_{22} \end{pmatrix}.$$

**Definition.** A **linear functional** on  $V$  is a linear map from  $V$  to  $\mathbb{C}$ .

Interestingly, the linear functionals on a vector space  $V$  themselves form another vector space  $V^*$ . In particular, given linear functionals  $\rho$  and  $\sigma$ , define  $\rho + \sigma$  pointwise, i.e.  $(\rho + \sigma)(v) = \rho(v) + \sigma(v)$ . Similarly, given a complex number  $c$ , define  $(c\rho)(v) = c\rho(v)$ . We call  $V^*$  the **dual space** of  $V$ .

**Theorem 1** (Riesz representation). *Let  $V$  be a finite-dimensional vector space, and let  $\rho : V \rightarrow \mathbb{C}$  be a linear functional. Then there is a unique vector  $w \in V$  such that  $\rho(v) = \langle w, v \rangle$ , for all  $v \in V$ .*

**Proposition 2.** Let  $V$  be a finite-dimensional vector space. If  $A : V \rightarrow W$  is linear, then there is a unique linear operator  $A^* : W \rightarrow V$  such that  $\langle w, Av \rangle = \langle A^*w, v \rangle$ , for all  $v \in V$  and  $w \in W$ .

**Fact.** In a matrix representation of linear operators, the matrix for the operator  $A^*$  is the conjugate transpose of the matrix for the operator  $A$ .

**Definition.** Let  $A$  be a linear operator. Then we define:

- $A$  is **normal** if  $A^*A = AA^*$ .
- $A$  is **self-adjoint** if  $A^* = A$ .
- $A$  is **unitary** if  $A^*A = AA^* = I$ .

**Example** (spin operators). Consider the two-dimensional vector space  $V$ , whose elements we will represent by column vectors. We define the function  $\sigma_x : V \rightarrow V$  to take  $(a, b)$  to  $\frac{1}{2}(a, -b)$ . Clearly  $\sigma_x$  is linear. In fact,  $\sigma_x$  can be represented as the matrix

$$\sigma_x = \frac{1}{2} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}.$$

It's easy to see then that  $\sigma_x^* = \sigma_x$ , and  $\sigma_x\sigma_x = 1$ . Hence,  $\sigma_x$  is both self-adjoint and unitary.

## 4 Composite systems