

Vector spaces and linear algebra

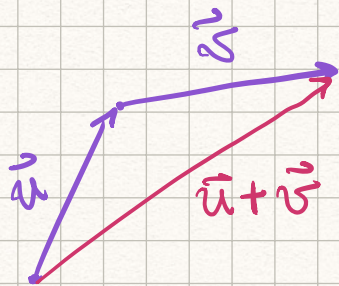
John von Neumann \rightarrow unification of Q.M. maths
 \hookrightarrow Mathematical rules of Q.M.

Vector space

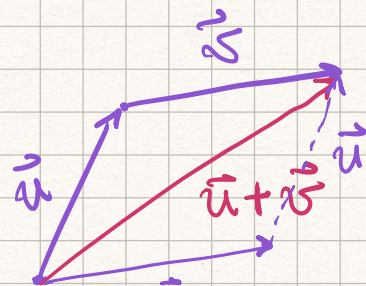
- Algebraic structure

\rightarrow we are used to geometrical vector

$$\vec{u}, \vec{v}; \Rightarrow \vec{v} + \vec{u}$$



Tail-to-head method



Parallelogram method

In quantum physics we use the Dirac notation

$|\dots\rangle$: vector: "a ket"

"..." what labels our vector

A vector space consists of a set of vectors

$$|\alpha\rangle, |\beta\rangle, |\chi\rangle, \dots$$

and some scalars ("numbers"), $a, b, c, \dots \in \mathbb{C}$

$a, b, c, \dots \in \mathbb{C}$: a, b, c, \dots belong to the set of complex numbers

Algebra \Rightarrow vector addition:

$$|\alpha\rangle + |\beta\rangle = |\kappa\rangle \quad \text{is also in the vector space}$$

Scalar multiplication

$$a|\alpha\rangle = |\kappa\rangle \quad \text{is also in the vector space}$$

Linear combination

$$a|\alpha\rangle + b|\beta\rangle + c|\gamma\rangle + \dots$$

$|\lambda\rangle$ is linearly independent from the set $|\alpha\rangle, |\beta\rangle, |\gamma\rangle, \dots$

If we cannot write

$$|\lambda\rangle \neq a|\alpha\rangle + b|\beta\rangle + c|\gamma\rangle \dots$$

A set of vectors is linearly independent if no vector in the set is a linear combination of the others

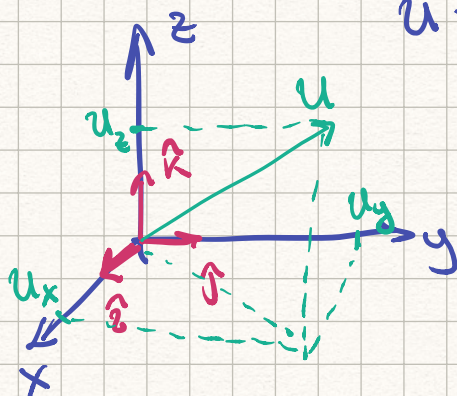
A set of vectors **spans** the vector space if every vector (in the space) can be written as a linear superposition of the vectors in the set

A set of linearly independent vectors that spans the space is a **basis**

The number of vectors in a basis is the **dimension** of the space

Example:

In 3D



$$\vec{u} = u_x \hat{i} + u_y \hat{j} + u_z \hat{k}$$

with $u_x, u_y, u_z \in \mathbb{C}$.
 $\{\hat{i}, \hat{j}, \hat{k}\}$ is the **basis** of the 3D cartesian space, which has dimension 3

Finite-dimensional space

All vectors can be written as a finite sum of other vectors (on the basis)

Basis: $\{|e_1\rangle, |e_2\rangle, \dots, |e_n\rangle\}$

dimension is n

Every vector $|\alpha\rangle$ can be written as

$$|\alpha\rangle = a_1 |e_1\rangle + a_2 |e_2\rangle + \dots + a_n |e_n\rangle$$

The scalars a_1, \dots, a_n are the **components** or **coordinates** of $|\alpha\rangle$ in the basis of $|e_i\rangle$

$$|\alpha\rangle = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix}$$

A vector can be seen as a column of numbers

$$\vec{u} = \begin{pmatrix} u_x \\ u_y \\ u_z \end{pmatrix}$$

These are a particular case of the **algebra of Tables**

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & & & & \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{pmatrix}$$

• Two-index object
Matrix (pl. matrices)

rectangular ($m \neq n$) or squared ($m = n$) matrix

Matrices also follow their algebra of addition (adding components one by one) and scalar multiplication (rescaling all elements)

Operation on Matrices

Transposition

The transpose (A^T : "A transposed") interchanges rows and columns

$$A^T = \begin{pmatrix} a_{11} & a_{21} & a_{31} & \dots & a_{m1} \\ a_{12} & a_{22} & a_{32} & \dots & a_{m2} \\ a_{13} & a_{23} & a_{33} & \dots & a_{m3} \\ \vdots & & & & \\ a_{1n} & a_{2n} & a_{3n} & \dots & a_{mn} \end{pmatrix} \quad \begin{array}{l} \text{A } m \times n \text{ matrix} \\ \text{yields } A^T \text{ which} \\ \text{is } n \times m \end{array}$$

If A is squared ($m=n$), the determinant

$$\det(A) = |A| = \begin{vmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & & & & \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{vmatrix}$$

Example:

$$\det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}$$

Note that $|x\rangle^T$

$$|x\rangle = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix} \Rightarrow |x\rangle^T = (a_1, a_2, \dots, a_n)$$

When working with complex-valued vector fields, it's better to also take the complex conjugate when transposing

Apply the "transpose-conjugate" operation

• Hermitian conjugate \Rightarrow Hermite

• Adjoint

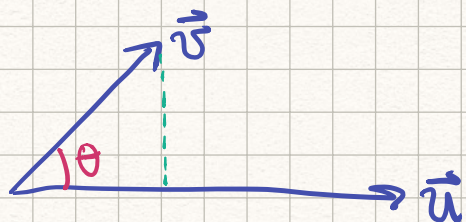
We use the symbol A^\dagger : "A dagger"

$$A^\dagger \equiv (A^T)^*$$

$$|\alpha\rangle^\dagger = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix}^{T*} = (a_1^*, a_2^*, \dots, a_n^*) \equiv \langle \alpha| \quad \text{"the bra of } |\alpha\rangle \text{"}$$
$$|\alpha\rangle^\dagger \equiv \langle \alpha|$$

Scalar product

$\vec{u} \cdot \vec{v} = uv \cos\theta$: shadow that \vec{v} casts upon \vec{u} .



$$\vec{u} \cdot \vec{v} = u_x v_x + u_y v_y + u_z v_z$$

For abstract vectors the scalar product is called inner product

$\langle \beta | \alpha \rangle$: scalar

$$\langle \beta | = (b_1^*, b_2^*, \dots, b_n^*)$$

$$\langle \beta | \alpha \rangle = (b_1^*, b_2^*, \dots, b_n^*) \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix}$$

$$= b_1^* a_1 + b_2^* a_2 + \dots + b_n^* a_n \quad \text{: scalar}$$

If we take

$$\begin{aligned}\langle \alpha | \alpha \rangle &= a_1^* a_1 + a_2^* a_2 + \dots + a_n^* a_n \\ &= |a_1|^2 + |a_2|^2 + \dots + |a_n|^2 : \text{Real number}\end{aligned}$$

The norm (length) of the vector:

$$\| \alpha \| \equiv \sqrt{\langle \alpha | \alpha \rangle}$$

Two vectors that have zero scalar product are **orthogonal**

Vectors in a basis should

- be normalized: $\| e_i \| = 1$
- be mutually orthogonal: $\langle e_i | e_j \rangle = \delta_{ij} = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$

δ_{ij} : Kronecker symbol

Applications or transformations

Functions that act on vectors

Let's assume a linear function T

$$T(a|\alpha\rangle + b|\beta\rangle) = aT|\alpha\rangle + bT|\beta\rangle$$

We have full knowledge of T if we know the effect that T has on every element of the basis

$$T|e_1\rangle = T_{11}|e_1\rangle + T_{21}|e_2\rangle + \dots + T_{n1}|e_n\rangle$$

$$T|e_2\rangle = T_{12}|e_1\rangle + T_{22}|e_2\rangle + \dots + T_{n2}|e_n\rangle$$

\vdots

$$T|e_n\rangle = T_{1n}|e_1\rangle + T_{2n}|e_2\rangle + \dots + T_{nn}|e_n\rangle$$

For any vector $|\alpha\rangle$ we compute $T|\alpha\rangle$ by decomposing $|\alpha\rangle$ in the elem. of the basis

$$|\alpha\rangle = \sum_{i=1}^n a_i |e_i\rangle = a_1 |e_1\rangle + a_2 |e_2\rangle + \dots + a_n |e_n\rangle$$

$$T|\alpha\rangle = \sum_{i=1}^n a_i T|e_i\rangle$$

$$\begin{aligned} T|e_i\rangle &= T_{1i}|e_1\rangle + T_{2i}|e_2\rangle + \dots + T_{ni}|e_n\rangle \\ &= \sum_{k=1}^n T_{ki}|e_k\rangle \end{aligned}$$

$$\begin{aligned} T|\alpha\rangle &= \sum_{i=1}^n a_i \sum_{k=1}^n T_{ki}|e_k\rangle \\ &= \sum_{k=1}^n \left(\sum_{i=1}^n T_{ki} a_i \right) |e_k\rangle = \begin{pmatrix} \sum_{i=1}^n T_{1i} a_i \\ \sum_{i=1}^n T_{2i} a_i \\ \vdots \\ \sum_{i=1}^n T_{ni} a_i \end{pmatrix} = T|\alpha\rangle \end{aligned}$$

The transformation is reduced to a matrix

$$T = \begin{pmatrix} T_{11} & T_{12} & \dots & T_{1n} \\ T_{21} & T_{22} & \dots & T_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ T_{n1} & T_{n2} & \dots & T_{nn} \end{pmatrix}$$

Hermitian transformation

$$T = T^\dagger : T \text{ is Hermitian}$$

Unitary : $T^{-1} = T^\dagger : T \text{ is unitary}$

Eigenvector

Don't get rotated in space, only rescaled

$$T|\alpha\rangle = \lambda|\alpha\rangle \quad : \text{ with } \lambda \text{ a scalar}$$

$|\alpha\rangle$ is an eigenvector of T with eigenvalue λ

$$T|\alpha\rangle = \lambda\mathbb{1}|\alpha\rangle$$

$$T|\alpha\rangle - \lambda\mathbb{1}|\alpha\rangle = 0$$

$|\alpha\rangle$ is nonzero

$$(T - \lambda\mathbb{1})|\alpha\rangle = 0 \quad \Rightarrow \quad T - \lambda\mathbb{1} \text{ has to be singular not invertible}$$

$$\det(T - \lambda\mathbb{1}) = \begin{vmatrix} T_{11} - \lambda & T_{12} & \dots & T_{1n} \\ T_{21} & T_{22} - \lambda & \dots & T_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ T_{n1} & T_{n2} & \dots & T_{nn} - \lambda \end{vmatrix} = 0$$

polynomial equation for λ of order n .

For Hermitian matrices

- Eigenvalues of a Hermitian transformation are real number
- Eigenvectors of a Hermitian transformation belonging to distinct eigenvalues are orthogonal
- The eigenvector of a Hermitian transformation span the space

Infinite-dimensional spaces

- Countable (like integers)
- Continuous (like the set of real numbers)

Vectors can be functions \Rightarrow function spaces

Upgrade

$$\sum \rightarrow \int$$

We could write the scalar product in the following way (between two vector functions)

$$\langle f | g \rangle = \int f^*(x) g(x) dx$$

Transformations will be done through **linear operator**. For instance

$$\frac{d}{dx}, \frac{d^2}{dx^2}, \text{ or simply: } x$$