

# 3.1 Linear Algebra

## Vector spaces

Instead of visualizing a complete set of orthonormal basis functions,  $f_\alpha(r)$ , consider a set of basis vectors which 'spans a space',  $\{|\alpha\rangle\} \rightarrow$  e.g.,  $|x\rangle, |y\rangle, |z\rangle$

plus a set of scalars,  $\{a\}$ , which can be multipliers

basis  $\Rightarrow$  a set of 'linearly independent' vectors which spans said space

span  $\Rightarrow$  every vector (in this space) can be written as a linear combination of said vectors

vector addition is commutative and associative

scalar multiplication is distributive and associative

within basis  $\{|e_i\rangle\}$ , an arbitrary vector,  $|\alpha\rangle$ , can be expressed as  $a_i |e_i\rangle$ , sum implied

## Inner products

'inner product' operation:  $\langle \alpha | \beta \rangle$

'norm':  $\| \alpha \| \equiv \sqrt{\langle \alpha | \alpha \rangle}$ , a scalar

norm = 1  $\Rightarrow$  'normalized'

$\langle \alpha_i | \alpha_j \rangle \propto \delta_{ij} \Rightarrow$  'orthogonal'

both properties  $\Rightarrow$  'orthonormal set'

Schwarz inequality:  $|\langle \alpha | \beta \rangle|^2 \leq \langle \alpha | \alpha \rangle \langle \beta | \beta \rangle$

## 'Linear' transformations

e.g., in three-space,

(a) rotate every vector by  $A^\circ$  about  $|z\rangle$ , or

(b) reflect every vector in  $xy$  plane

each vector becomes a new vector (which *could* be identical)

operation:  $|e'_i\rangle = \hat{T} |e_i\rangle = |e_j\rangle T_{ji}$ , sum implied

matrix element  $T_{ij} = \{\mathbf{T}\}_{ij} \equiv \langle e_i | \hat{T} | e_j \rangle$

operator  $\hat{T} \equiv |e_i\rangle T_{ij} \langle e_j| \Leftrightarrow$  matrix  $\mathbf{T}$

matrix addition

matrix multiplication:  $\mathbf{U} = \mathbf{ST} \Leftrightarrow U_{ik} = S_{ij} T_{jk}$ ,  
sum implied

define components of  $|\alpha\rangle$  as a column matrix

transformation rule can be written as  $\mathbf{a}' = \mathbf{T}\mathbf{a}$

transpose:  $\{\tilde{\mathbf{T}}\}_{ij} \equiv \{\mathbf{T}\}_{ji}$

transpose of a column vector is a row vector

symmetric:  $\tilde{\mathbf{T}} = \mathbf{T}$

antisymmetric:  $\tilde{\mathbf{T}} = -\mathbf{T}$

conjugate:  $\{\mathbf{T}^*\}_{ij} \equiv \{\mathbf{T}\}_{ij}^*$

real:  $\mathbf{T}^* = \mathbf{T}$

imaginary:  $\mathbf{T}^* = -\mathbf{T}$

Hermitian conjugate (or adjoint):  $\{\mathbf{T}^\dagger\}_{ij} \equiv \{\mathbf{T}\}_{ji}^*$

a *square* matrix is Hermitian (or self-adjoint) iff  
it is equal to its Hermitian conjugate, *i.e.*,

$$\mathbf{T}^\dagger = \mathbf{T}$$

skew-Hermitian:  $\mathbf{T}^\dagger = -\mathbf{T}$

in matrix form,  $\langle \alpha | \beta \rangle = \mathbf{a}^\dagger \mathbf{b}$

in general, matrix multiplication is *not* commutative; the difference is called the commutator:  $[\mathbf{S}, \mathbf{T}] \equiv \mathbf{ST} - \mathbf{TS}$

$$(\tilde{\mathbf{S}}\tilde{\mathbf{T}}) = \tilde{\mathbf{T}}\tilde{\mathbf{S}}$$

$$(\mathbf{ST})^\dagger = \mathbf{T}^\dagger\mathbf{S}^\dagger$$

unit matrix is  $\mathbf{1}$ , where  $\{\mathbf{1}\}_{ij} = \delta_{ij}$

inverse,  $\mathbf{T}^{-1}$ :  $\mathbf{T}^{-1}\mathbf{T} = \mathbf{TT}^{-1} \equiv \mathbf{1}$

inverse exists iff determinant  $\neq 0$

$\mathbf{T}^{-1} = (\det\mathbf{T})^{-1}\tilde{\mathbf{C}}$ , where  $\mathbf{C}$  is matrix of cofactors (see text)

$$(\mathbf{ST})^{-1} = \mathbf{T}^{-1}\mathbf{S}^{-1}$$

matrix  $\mathbf{U}$  is unitary iff  $\mathbf{U}^{-1} = \mathbf{U}^\dagger$

consider basis,  $\{e_i\}$ , and a new basis spanning the same space,  $\{f_i\}$ :

these bases must be related by a linear transformation,  $\mathbf{S} \ni |e_j\rangle = |f_i\rangle S_{ij}$

suppose we have vector  $|\alpha\rangle$ , expressed using basis  $\{e_i\}$ :  $|\alpha\rangle = a_i^e |e_i\rangle$

then  $|\alpha\rangle = a_i^e |f_j\rangle S_{ji}$ , or  $a_j^f = S_{ji} a_i^e$ , or  $\mathbf{a}^f = \mathbf{S} \mathbf{a}^e$

remember  $\hat{T}$ : in old basis,  $\mathbf{a}'^e = \mathbf{T}^e \mathbf{a}^e$

in new basis,  $\mathbf{a}'^f = \mathbf{S} \mathbf{a}'^e = \mathbf{S} (\mathbf{T}^e \mathbf{a}^e) = \mathbf{S} \mathbf{T}^e \mathbf{S}^{-1} \mathbf{a}^f$

$\therefore \mathbf{T}^f = \mathbf{S} \mathbf{T}^e \mathbf{S}^{-1}$ ; or  $\mathbf{T}^f$  and  $\mathbf{T}^e$  are 'similar'

also, both bases are orthonormal iff  $\mathbf{S}$  is unitary

furthermore,  $\det(\mathbf{T}^f) = \det(\mathbf{T}^e)$  and  
 $\text{Tr}(\mathbf{T}^f) = \text{Tr}(\mathbf{T}^e)$

## Eigenvectors and Eigenvalues

Consider a particular linear transformation, a rotation by  $\theta$  about some axis in three-space:

all vectors will travel around a cone about the axis of rotation—except those vectors *on* that axis, which are unchanged

those special vectors which transform into simple multiples of themselves,  $\hat{T} |\alpha\rangle = \lambda |\alpha\rangle$ , are called *eigenvectors* of the transformation, while the (complex)  $\lambda$  are called *eigenvalues*

With respect to a *particular* basis, the eigenstates of  $\hat{T}$  can be found from  $\mathbf{T}\mathbf{a} = \lambda\mathbf{a}$ , or  $(\mathbf{T} - \lambda\mathbf{1})\mathbf{a} = \mathbf{0}$ :

if  $\mathbf{a} \neq \mathbf{0}$ , then  $(\mathbf{T} - \lambda\mathbf{1})^{-1}$  must be singular and  $\det(\mathbf{T} - \lambda\mathbf{1}) = 0$

this last condition leads to all the eigenvalues and eigenvectors of  $\hat{T}$  *in that basis*

If  $\mathbf{T}$  is diagonalizable, then there is a similarity matrix which transforms it into diagonal form  
 $= \mathbf{S}\mathbf{T}\mathbf{S}^{-1}$

# Hermitian Transformations

Consider the basis  $\{i\}$ , which spans the space of interest:

$$\langle \alpha | \hat{T} | \beta \rangle = \langle \alpha | i \rangle \langle i | \hat{T} | j \rangle \langle j | \beta \rangle = \alpha_i^* T_{ij} \beta_j = (\alpha_i T_{ij}^*)^* \beta_j = (\alpha_i \{\mathbf{T}^\dagger\}_{ji})^* \beta_j = (\{\mathbf{T}^\dagger\}_{ji} \alpha_i)^* \beta_j$$

*i.e.*,  $\hat{T}^\dagger$  is that transformation which, when applied to the first member of an inner product, gives the same result as if  $\hat{T}$  itself had been applied to the second vector.

Four properties:

1. any Hermitian matrix can be diagonalized by a similarity transformation;
2. the eigenvalues of a Hermitian transformation are real;
3. the eigenvectors of a Hermitian transformation belonging to distinct eigenvalues are orthogonal; the other eigenvectors can be made orthogonal using procedure of Gram-Schmidt;
4. the eigenvectors of a Hermitian transformation span the space.  
**NB**, #4 is true only for finite-dimensional spaces



## 3.2 Function Spaces

It is possible to define certain classes of functions so that they constitute a vector space:

e.g., all polynomials on  $x \in (-1, 1)$  of degree  $< N$ ,  $P(N)$ , or  
all odd functions that  $= 0$  at  $x = 1$ , or  
all periodic functions with period  $\pi$ .

It is necessary that they span the space and convenient that they be orthonormal.

Linear operators (such as  $\hat{x}$  or  $\hat{D} \equiv d/dx$ ) behave as linear transformations if they carry functions in this space into other members of the space.

Note that the operator  $\hat{x}$  is *not* linear in  $P(N)$  since it is able to promote the  $(N - 1)^{\text{th}}$ -order polynomial into the  $N^{\text{th}}$ -order polynomial, which is *not* a member of  $P(N)$ .

But,  $\hat{x}$  is linear in  $P(\infty)$

$\hat{x}$  is also Hermitian in  $P(\infty)$

Note, however, that it has *no* eigenfunctions in  $P(\infty)$ !

In fact, it can be shown that the eigenfunctions of  $\hat{x}$  are Dirac delta functions

In general, in infinite-dimensional spaces some Hermitian operators have complete sets of eigenvectors, some have incomplete sets, and others have no eigenvectors at all (in that space).

Unfortunately, the completeness property is *essential* in quantum mechanical applications. This will be discussed more shortly.

## Hilbert spaces

A *complete inner product space* is called a Hilbert space.

$P(\infty)$  can be *completed* by adding the remainder of the square-integrable functions on the interval  $x = (-1, 1)$ :  $L_2(-1, 1)$ .

We are concerned with the particular Hilbert space,  $L_2(-\infty, \infty)$ .

The eigenfunctions of the Hermitian operators  $i\hat{D} = id/dx$  and  $\hat{x} = x$  are  $f_\lambda(x) = A_\lambda e^{-i\lambda x}$  and  $g_\lambda(x) = B_\lambda \delta(x - \lambda)$ , respectively.

Every real number is an eigenvalue of  $i\hat{D}$ , and every real number is an eigenvalue of  $\hat{x}$ . The set of eigenvalues of an operator is called its spectrum;  $i\hat{D}$  and  $\hat{x}$  are operators with continuous spectra.

Unfortunately, these eigenfunctions are not square integrable on  $x = (-1, 1)$ , and therefore they *do not lie in Hilbert space*.

It is customary to ‘orthonormalize’ these (fundamentally unnormalizable) functions to the Dirac delta function:

$$\langle f_\lambda | f_\mu \rangle = \delta(\lambda - \mu), A_\lambda = (2\pi)^{-1/2} \text{ and} \\ \langle g_\lambda | g_\mu \rangle = \delta(\lambda - \mu), B_\lambda = 1$$

Importantly, these two *complete* sets of ‘nearly orthonormalized’ functions are *useful* in quantum mechanics.

Their continuous eigenvalue spectra necessitate changing the sum over discrete eigenstates into an integral over continuous eigenstates:

$$\hat{1} = \int_{-\infty}^{\infty} d\lambda |f_\lambda\rangle\langle f_\lambda| = \int_{-\infty}^{\infty} d\lambda |g_\lambda\rangle\langle g_\lambda|$$

## 3.3 Statistical Interpretation

Postulates:

1. The state of a particle is represented by a normalized vector,  $|\Psi\rangle$ , in the Hilbert space  $L_2$ .
2. Observables  $Q(x, p, t)$  are represented by  $\hat{Q}(x, (\hbar/i)(\partial/\partial x), t)$ ; the expectation value of  $Q$  in the state  $\Psi$  is

$$\langle Q \rangle_{\Psi} = \int dx \Psi(x, t)^* \hat{Q} \Psi(x, t) = \langle \Psi | \hat{Q} | \Psi \rangle$$

3. If you measure  $Q$  for particle in state  $\Psi$ , you *must* get one of the eigenvalues of  $\hat{Q}$ . The probability of getting a particular eigenvalue  $\lambda$  is equal to absolute square of the  $\lambda$  component of  $\Psi$  when expressed in orthonormal basis of eigenvectors.

Clearly, the eigenfunctions of any operator representing an observable *must* form a complete set.

$\hat{Q} |e_n\rangle = \lambda_n |e_n\rangle$ ,  $\langle e_n | e_m \rangle = \delta_{nm}$ , and  
 $|\Psi\rangle = c_n |e_n\rangle$  imply that the probability of  
getting one particular eigenvalue  $\lambda_n$   
 $= |c_n|^2 = |\langle e_n | \Psi \rangle|^2$

## 3.4 Uncertainty Principle

For any two observables,  $\sigma_A^2 \sigma_B^2 \geq (\langle [\hat{A}, \hat{B}] \rangle / 2i)^2$ .

Observables such that their operators do not commute are called *incompatible*; it is not possible to find a complete basis set in which both operators are diagonal.

Conversely, observables such that their operators commute *can* be diagonalized simultaneously, leading to a complete set of common eigenfunctions.

Moreover, in those cases where the operators do not commute, it is possible to define a minimum uncertainty wavepacket.

The energy-time uncertainty principle relates the time it takes *any* observable of a system to change appreciably,  $\Delta t = \frac{\sigma_Q}{|d\langle Q \rangle / dt|}$ , to the corresponding uncertainty in a measurement of that system's energy:  $\Delta E \Delta t \geq \hbar/2$ .

**NB**, this principle *does not* imply that energy conservation can be “suspended” to the extent of  $\Delta E$  for a time  $\Delta t$ .