

Adjoint Continued:

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$$4) (A^t)^t = A$$

Proof: $\langle x | (A^t)^t y \rangle \stackrel{\text{def. } A^t}{=} \langle A^t x | y \rangle \stackrel{\text{def. of } *}{=} \langle y | A^t x \rangle \stackrel{\text{def. of } t}{=} \langle Ay | x \rangle^* = \langle x | Ay \rangle$

$\therefore (A^t)^t = A$

$$5) (AB)^t = B^t A^t$$

Proof: $\langle x | (AB)^t y \rangle \stackrel{\text{def. } t}{=} \langle ABx | y \rangle = \langle A(Bx) | y \rangle = \langle Bx | A^t y \rangle$
 $= \langle x | B^t A^t y \rangle \Rightarrow (AB)^t = B^t A^t$

Theorem: A is a l.f. on V . Then $A=0 \Leftrightarrow \langle x | Ay \rangle = 0 \forall x, y$.

Proof: a) $A=0 \Rightarrow \langle x | Ay \rangle = 0$ trivial

b) If $\langle x | Ay \rangle = 0 \forall x, y$ let $x = Ay \Rightarrow \langle Ay | Ay \rangle = 0$

But by the axioms for an ~~inner~~ inner product space $\Rightarrow \langle Ay | Ay \rangle = 0$

But if $Ay = 0$ for all $y \Rightarrow$ ~~inner~~ $A=0$ ✓

Theorem: Given a $\left\{ \begin{array}{l} \text{self adjoint} \\ \text{arbitrary} \end{array} \right\}$ l.f. A on V where $V = \left\{ \begin{array}{l} \text{inner product} \\ \text{unitary} \end{array} \right\}$

space then $A=0 \Leftrightarrow \langle x | Ax \rangle = 0$.

Proof: not proved in class.

Theorem: If $A=A^t$ & $B=B^t$ then $AB=(AB)^t$ or $BA=(BA)^t$ only if $[A, B]=0$.

Proof: $(AB)^t = B^t A^t = BA$
 $\stackrel{=} {=} BA$
 $\stackrel{=} {=} AB$ } only if $[A, B]=0$

Definition: If $A^\dagger = -A$ A is skew-symmetric or skew-Hermitian

(F147)

Note: $A = \underbrace{\frac{1}{2}(A+A^\dagger)}_{\text{self-adjoint}} + \underbrace{\frac{1}{2}(A-A^\dagger)}_{\text{skew self-adjoint}}$

Theorem: A is a l.t. on V ; A is Hermitian iff $\langle x|Ax \rangle$ is real

Proof: a) If $A = A^\dagger \Rightarrow \langle x|Ax \rangle = \langle A^\dagger x|x \rangle = \langle x|A^\dagger x \rangle^* = \langle x|Ax \rangle^*$

\uparrow def. A^\dagger \uparrow def. $*$ \uparrow $A = A^\dagger$

$\therefore A = A^\dagger \Rightarrow \langle x|Ax \rangle$ is real.

b) If $\langle x|Ax \rangle$ is always real ($\forall x$) $\Rightarrow \langle x|Ax \rangle = \langle x|Ax \rangle^* = \langle Ax|x \rangle = \langle x|A^\dagger x \rangle$

\uparrow assumed \uparrow def. $*$ \uparrow def. \dagger

$\therefore \langle x|Ax \rangle = \langle x|Ax \rangle^* \Rightarrow A = A^\dagger$

Notation: $\langle x|Ax \rangle \equiv$ expectation value of A in the state x .

In QM $\langle x|Ax \rangle \equiv \langle x|A|x \rangle$ is the value obtained by measuring the eigenvalue of A . Since this is a physical quantity $\langle x|Ax \rangle$ must be real.

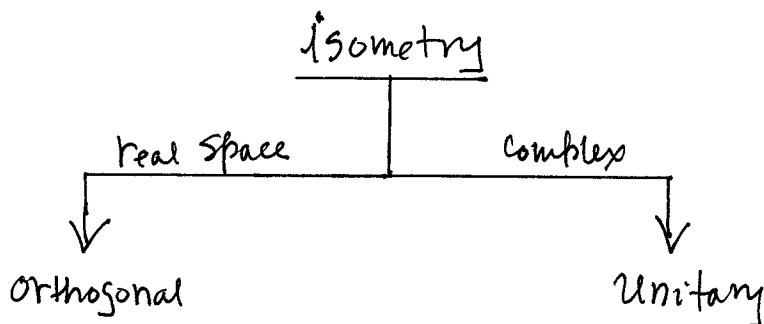
Hence, observables in QM are the (real) eigenvalues of Hermitian operators.

ISOMETRIES: ORTHOGONAL & UNITARY TRANSFORMATIONS

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Definition: A is an isometry of $A^+ = A^{-1}$ (iso = same -metry = measure)

$$A^+ = A^{-1} \Rightarrow AA^+ = I = A^+A$$



SUMMARY OF TRANSFORMATIONS

$$(A)_{ij} = \alpha_{ij}$$

	Self Adjoint Transf. ($A^+ = A$)	Isometric Transf. ($A^+ = A^{-1}$)
Unitary (complex) Space	Hermitian (A) $(A^+)_{ij} = \alpha_{ji}^*$	Unitary $\sum_k \alpha_{ik} \alpha_{jk}^* = \delta_{ij}$
Euclidean (real) Space	Symmetric $(A^+)_{ij} = \alpha_{ji}$	Orthogonal $\sum_k \alpha_{ik} \alpha_{jk} = \delta_{ij}$

Theorem on Isometric Transformations:

The following 3 conditions are equivalent statements about unitary transformations:

a) $U^+U = I$ (or $U^+ = U^{-1}$)

b) $\langle Ux | Uy \rangle = \langle x | y \rangle$

c) $|Ux| = |x|$

Unitary Transformations (cont'd):

Proof: We show (1) \Rightarrow (2) \Rightarrow (3) \Rightarrow (1)

$$(1) \Rightarrow (2) \quad U^t U = I \Rightarrow \langle Ux | Uy \rangle = \langle x | \underbrace{U^t U}_{I} y \rangle = \langle x | y \rangle \quad \checkmark \quad (1)$$

$$(2) \Rightarrow (3) \quad \langle Ux | Uy \rangle = \langle x | y \rangle \Rightarrow \langle Ux | Ux \rangle = \langle x | x \rangle \Rightarrow |Ux|^2 = |x|^2 \quad (*)$$
$$\Rightarrow |Ux| = |x| \quad \checkmark \quad (2)$$

$$(3) \Rightarrow (1) \quad \left. \begin{aligned} |Ux|^2 &= \langle Ux | Ux \rangle = \langle x | U^t U x \rangle \\ &\stackrel{(*)}{=} \langle x | x \rangle \end{aligned} \right\} \Rightarrow U^t U = I \quad (3)$$

Isometric Transformations preserve the lengths of vectors, and scalar products, and hence they also preserve the angles between vectors:

In an abstract (i.e. any) vector space

$$\cos \theta \equiv \frac{\langle x | y \rangle}{|x| \cdot |y|} \xrightarrow{U} \frac{\langle Ux | Uy \rangle}{|Ux| \cdot |Uy|} = \frac{\langle x | y \rangle}{|x| \cdot |y|} \quad \checkmark \quad (4)$$

Side Comment: If $\{x_i\}$ is a CNB basis so is $\{Ux_i\}$.

Unitarity Constraints: $(U)_{ij} = u_{ij} \quad (U^t)_{ij} = u_{ji}^* \Rightarrow \quad (5)$

$$(I)_{ik} = (UU^t)_{ik} = \sum_j u_{ij} (U^t)_{jk} = \sum_j u_{ij} u_{kj}^* \quad \left\{ \begin{array}{l} \sum_j u_{ij} u_{kj}^* = \delta_{ik} \end{array} \right. \quad (6)$$

\hookrightarrow these are the familiar orthogonality relations when $u_{ij} = \text{real}$.

Check: $2 \times 2 \quad U = \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}$

$$i=1 \quad k=1 \Rightarrow \delta_{11} = 1 \stackrel{?}{=} \sum_j u_{1j} u_{1j}^* = u_{11} u_{11}^* + u_{12} u_{12}^* = \cos^2 \theta + \sin^2 \theta = 1 \quad \checkmark$$

$$i=1 \quad k=2 \Rightarrow \delta_{12} = 0 \stackrel{?}{=} \sum_j u_{1j} u_{2j}^* = u_{11} u_{21}^* + u_{12} u_{22}^* = \cos \theta (-\sin \theta) + \sin \theta \cos \theta = 0 \quad \checkmark \quad (7)$$

Theorem: If A is orthogonal then $\det A = \pm 1$

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Proof: $AA^T = I \Rightarrow \det(AA^T) = (\det A)(\det A^T) = \det I = 1$ (8)

$\Rightarrow = (\det A^{-1}) = \det A$
 $\therefore (\det A)^2 = 1 \Rightarrow \boxed{\det A = \pm 1}$ (9)

Note that if A is orthogonal $\Rightarrow A^T = A^{-1}$ and $\det A^T = \det A$ (recall volume argument)

$\det A = +1 \Rightarrow$ rotation

$\det A = -1 \Rightarrow$ inversion

Theorem: If U is unitary then $|\det U| = 1 \Leftrightarrow \det U = e^{i\theta}$

Proof: Assume U is diagonalized so that $U = \begin{pmatrix} u_{11} & & 0 \\ & u_{22} & \\ 0 & & u_{33} \end{pmatrix}$ (10)

$U^{-1} = \begin{pmatrix} 1/u_{11} & & 0 \\ & 1/u_{22} & \\ 0 & & 1/u_{33} \end{pmatrix} = U^T$ (11)

$(\det U)(\det U^T) = (u_{11}u_{22}u_{33}) \frac{1}{u_{11}u_{22}u_{33}} = 1$ (12)

However, by definition the elements of U^T are $\begin{pmatrix} u_{11}^* & & 0 \\ & u_{22}^* & \\ 0 & & u_{33}^* \end{pmatrix}$ (13)

Hence $U^T = U^{-1} \Rightarrow u_{11}^* = 1/u_{11}$ etc $\Rightarrow |u_{11}|^2 = 1$
 $\Rightarrow u_{11} = e^{i\theta_1}$ etc.

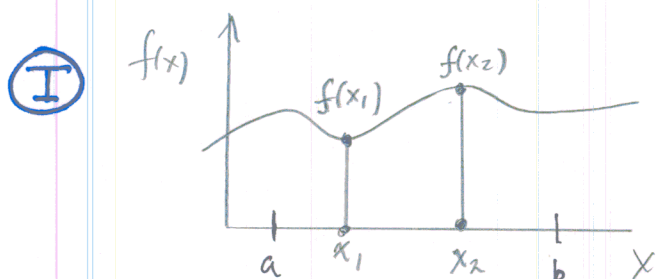
U must then have the form

$U = \begin{pmatrix} e^{i\theta_1} & & 0 \\ & e^{i\theta_2} & \\ 0 & & e^{i\theta_3} \end{pmatrix} \Rightarrow \det U = e^{i(\theta_1 + \theta_2 + \theta_3)} \equiv e^{i\theta}$ (14)
 $\Rightarrow |\det U| = 1$

INTRODUCTION TO HILBERT SPACE

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TWO (RELATED) VIEWS OF A FUNCTION AS A VECTOR IN AN n -DIM SPACE



In this way a function is represented in the interval $[a, b]$ by an n set of values $(f(x_a), f(x_1), \dots, f(x_b))$ which give the "projections" of $f(x)$ on the "axes" (x_a, x_1, \dots, x_b) . This is similar to specifying the components of an n -dimensional vector by giving its components along the n -axes:

$$\vec{V} = (V_1, V_2, \dots, V_n).$$

Ⓙ We can also represent a function via a Taylor series in $[a, b]$

$$f(x) = \sum_{n=0}^{\infty} a_n x^n \quad \Rightarrow \quad f(x) = [a_0, a_1, \dots, a_n, \dots]$$

A function which is piecewise continuous (continuous in every subinterval in $[a, b]$)

These two are related: A function which is piecewise continuous can be expanded as in Ⓙ via a Taylor series [More later!]

SCALAR PRODUCT: For 2 functions $f(x), g(x)$

$$\langle f | g \rangle \equiv \int_a^b dx f(x) g(x) \quad (1)$$

NORM OF A FUNCTION $N[f] \equiv \langle f | f \rangle = \int_a^b dx |f(x)|^2 \quad (2)$

SIDE COMMENT ON SCALAR PRODUCT:

179.1

The definition $\langle f|g \rangle = \int_a^b dx f^*(x)g(x)$ is the natural extension of the usual finite dimensional result:

$$\langle a|b \rangle = \vec{a} \cdot \vec{b} = \sum_{i=1}^N a_i^* b_i = a_1^* b_1 + a_2^* b_2 + \dots + a_N^* b_N$$

Let $a_i \rightarrow a(x_i), \dots, a_N \rightarrow a(x_N)$; $b_i \rightarrow b(x_i), b_N \rightarrow b(x_N)$

Then $\langle a|b \rangle \rightarrow \sum_i a_i^* b(x_i) \rightarrow \int dx a^*(x) b(x)$ ✓

The only additional ingredient needed is the interval over which the integral is evaluated.

HILBERT SPACE:

The set of functions for which the norm is finite constitutes Hilbert space. Sometimes to ensure that the norm is finite, especially when the integration limits are ∞ , a weighting function $w(x)$ is introduced:

$$\langle f | g \rangle \rightarrow \int_a^b dx f^*(x) g(x) w(x) \quad (3)$$

For example: $w(x) \sim e^{-x}$ (Laguerre) or $e^{-x^2/2}$ (Hermite)

Reisz-Fischer Theorem: (not proved here)

Functions with $N[f] < \infty$ are complete: A Hilbert space is then a complete linear vector space with a complex scalar product.

DIGRESSION ON "∞" [G.F. Cantor - Theory of Transfinite Numbers]

- Denumerable ∞: \aleph_0 ["Aleph-Null"] $\left. \begin{matrix} 1, 2, 3, \dots \\ 2, 4, 6, \dots \end{matrix} \right\}$ KNOW CANTOR PROOF!! SEE "WORLD OF MATH" or GATMOW "ONE, TWO, THREE... INFINITY"
- Continuum: \mathbb{C} [points on a line]

(I) \Rightarrow Continuum infinity to specify points

(II) \Rightarrow Denumerable infinity " " "

However, for a piecewise continuous function a denumerable ∞ is enough.

Separable Hilbert Space: one having at least one denumerable basis. Otherwise the Hilbert space is non separable:

NOTATION IN HILBERT SPACE

F180.1

(a) $\langle f|g\rangle = 0 \Rightarrow f, g$ are orthogonal in $[a, b]$

(b) $\langle f|f\rangle = 1 \Rightarrow f$ is normalized in $[a, b]$

(c) $\langle f_i|f_j\rangle = \delta_{ij} \Rightarrow f_i, f_j$ are orthonormal in $[a, b]$

Example: $f_n(x) = \frac{1}{\sqrt{2\pi}} e^{inx}$; $n = 0, \pm 1, \pm 2, \dots$

The set $\{f_n(x)\}$ is orthonormal in $[-\pi, \pi]$. To see this:

$$\langle f_n|f_m\rangle = \int_{-\pi}^{\pi} f_n^*(x) f_m(x) dx = \left(\frac{1}{\sqrt{2\pi}}\right)^2 \int_{-\pi}^{\pi} (e^{inx})^* e^{imx} dx \quad (1)$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} dx e^{i(m-n)x} = \frac{1}{2\pi i(m-n)} e^{i(m-n)x} \Big|_{-\pi}^{\pi} = 0 \text{ if } m \neq n \quad (2)$$

However, when $n=m$ (2) gives

$$\langle f_n|f_n\rangle = \frac{1}{2\pi} \int_{-\pi}^{\pi} dx \cdot 1 = 1 \quad \checkmark \quad (3)$$

Hence altogether $\langle f_n|f_m\rangle = \delta_{mn} \quad \checkmark$

When a (non-negative) weight function $w(x)$ is used (as previously) then

$$\langle f_n|f_m\rangle \equiv \int_a^b dx f_n^*(x) f_m(x) w(x) = \delta_{mn} \Rightarrow \text{orthonormality}$$